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Abstract: Background

Generation and phenotyping of mutant mouse models continues to increase along with the search for the most efficient phenotyping tests. Here we asked if a combination of different locomotor tests is necessary for comprehensive locomotor phenotyping, or if a large data set from an automated gait analysis with the CatWalk system would suffice.

New Method

First we endeavored to meaningfully reduce the large CatWalk data set by Principal Component Analysis (PCA) to decide on the most relevant parameters. We analyzed the influence of sex, body weight, genetic background and age. Then a combination of different locomotor tests was analyzed to investigate the possibility of redundancy between tests.

Result

The extracted 10 components describe 80% of the total variance in the CatWalk, characterizing different aspects of gait. With these, effects of CatWalk version, sex, body weight, age and genetic background were detected. In addition, the PCA on a combination of locomotor tests suggests that these are independent without significant redundancy in their locomotor measures.

Comparison with existing methods The PCA has permitted the refinement of the highly dimensional CatWalk (and other tests) data set for the extraction of individual component scores and subsequent analysis.

Conclusion The outcome of the PCA suggests the possibility to focus on measures of the front and hind paws, and one measure of coordination in future experiments to detect phenotypic differences. Furthermore, although the CatWalk is sensitive for detecting locomotor phenotypes pertaining to gait, it is necessary to include other tests for comprehensive locomotor phenotyping.

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- The authors also certify that formal approval to conduct the experiments described has been obtained from the human subjects review board of their institution and could be provided upon request.
- If the studies deal with animal experiments, the authors certify that they were carried out in accordance with the National Institute of Health Guide for the Care and Use of Laboratory Animals (NIH Publications No. 80-23) revised 1996 or the UK Animals (Scientific Procedures) Act 1986 and associated guidelines, or the European Communities Council Directive of 24 November 1986 (86/609/EEC).
- The authors also certify that formal approval to conduct the experiments described has been obtained from the animal subjects review board of their institution and could be provided upon request.
- The authors further attest that all efforts were made to minimize the number of animals used and their suffering.
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Point-by-point response to reviewer's suggestions

Ref.: Ms. No. JNEUMETH-D-16-00606 Analysis of Locomotor Behavior in the German Mouse Clinic Journal of Neuroscience Methods

Dear Dr. Crunelli,

Dear Reviewers,

Thank you very much for the very helpful critical review of our submitted manuscript. As we received encouraging responses we are happy to address the remaining concerns by restructuring/rewriting different parts of the manuscript and including more information and details. Please find attached the revised version of our manuscript and below the point-by-point response to the reviewer's questions.

Having made these changes we now hope that the manuscript can be accepted for publication in the Special Issue: Measuring Behavior 2016.

With kind regards,

Annemarie Zimprich

Reviewers' comments:

Reviewer #1:

The present manuscript submitted by Zimprich and collaborators reports a statistical model dedicated to analyse and reduce by Principal Component Analysis a large set of data coming from a multitude of parameters recorded by currents video tracing software such as Catwalk into component which could describe mice motor phenotype.

1)

I found the idea to focus on large set of data relatively interesting and challenging, and the manuscript submitted by Zimprich and collaborators is perfectly in line with this. However, the manuscript is quite difficult to follow in those terms. The way the article is written and it corpus is exclusively linked to statistical correlations and descriptions which make the reading quite difficult to follow. Moreover, the absence of behavioural raw data in parallel to the statistical analysis make difficult as well to attribute a judgement about the quality of the data analysis and the model itself. It would have been nice to have side by side, the Raw Data showing differences between groups in terms of motor differences and the analysis obtained via PCA.

We improved the traceability by rearranging parts and included original data of parameters to illustrate the major points (please see also answer to question 10)).

2)

Moreover, I would advise the author to make the manuscript read by a native English speaker, the quality of the writing is for me not good enough for Journal of Neuroscience Methods.

We now have had a native English speaker critically reading and correcting the manuscript.

Comments:

INTRODUCTION

3)

Regarding the introduction section, my first comment is in regard of the term locomotor phenotype. It might be good to have a clear definition in this section of "locomotor phenotype" which wants to be detected with the Catwalk. Indeed, this term is a bit confusing because this task has been designed mainly to assess motor performance but not proper locomotion or activity which is something quite different.

We inserted a definition of "locomotor phenotype": "In this paper we focus on locomotor phenotypes, whereby the definition of "locomotor phenotype" is used very broadly. We include gait phenotypes (as analyzed by the CW), as well as activity (as in the Open Field (OF), home cage (HC) and SHIRPA) and motor ability phenotypes (as in the Grip Strength (GS), Rotarod (RR) and Vertical Pole (VP) test) into this term." (page 3, line 29ff)

4)

Because the methods and result section of the manuscript rely mainly on statistics and PCA it would have been good to mentioned literature which use similar sort of analyses for behaviour.

We quoted literature where a PCA was run on behavioral data (compare page 3, line 19ff) but now elaborate on this further in the Introduction and Discussion. We also included the citation from Ohl et al, which Reviewer #2 mentioned (also see question 26).

5)

- The second paragraph of the introduction is a little bit confusing which is probably related to writing issues. Because it is stated that in the manuscript the focus is carried on locomotor phenotype associated with neurological disease such as Parkinson's disease, Amyotrophic Lateral Sclerosis and Attention Deficit Hyperactivity Syndrome, I expected to see the PCA analysis carried on different mouse model used for these pathologies.

To circumvent this confusion we excluded this sentence from the introduction.

METHODS

6)

- The method part for me needs to be written again or at least better organized, it appear really messy. It would be good to have it extended by information about the animals, a summary table with how many animals has been used for each separate analysis, the age, and genotype would help.

We completely restructured the Methods and added a table in the Supporting Information for details of animal numbers per analysis as suggested (see Table S1).

7)

- I understood that the manuscript has been written to demonstrate the power of PCA for analysing a large pool of data, but it would be good to have for each behavioural test used a description of the different behavioural apparatus with the parameters used being explained or at least defined in a supplementary table.

We included brief descriptions of how the different behavioral tests were performed in the Methods and added tables listing the parameters used in the statistical analysis in the Supporting Information (Table S2 and S7).

8)

- It would good to know as well the type of protocol which has been used for the tasks, for example are the 1499 mice used for the statistical analyses of the 113 parameters of the CW been assessed in the same exact protocol?

Yes, all mice were subjected to the same CW testing protocol. The animals were all transferred to the testing room at least 30 minutes prior to testing. All animals were tested in a darkened room. Two to 4 continuous uninterrupted runs per animal were used for analysis. Only the apparatus version changes, i.e. from 7.1 to XT. The main difference between the versions is the time-related resolution of the cameras. In more detail the CW XT version has a lid above the runway, which has a red light integrated and the animals run towards a so called "goal box" underneath which their home cage is situated. The CW 7.1 version does not have a lid nor a "goal box".

9)

- In the statistical part, it would nice to have a little definition of what a z transformation and an oblimin rotation are. I understood that the z transformation has been chosen to normalise a set of different parameters in order to compare its together as well as oblimin rotation has been chosen to regroup positive correlation in component, but I think a little definition of each in the method part would be a nice plus.

We elucidate more on the Z-transformation and the oblique rotation in the Methods (page 6, line 19ff).

RESULTS

10)

- Globally the results part is really hard to follow. The fact that all paragraphs are a succession of statistical descriptions make the results part difficult to follow and to rely on actual mouse behaviour.

We included original data to illustrate the statements made in the paragraphs (for example Fig. 2, page 10; Fig. 3, page 12, Fig. 4, page 13 Fig. 6 and 7, pages 17 and 20).

11)

- I am very surprised to see such an enormous difference between CW 7.1 and XT in terms of correlation (Figure 1). Is the difference the same tendency when raw data for parameters are analysed directly? Are the performances of Males of Females mice analysed without z transformation for the XT or 7.1 versions comparable to the data after normalisation?

Yes, when comparing the original data of males and females by the two CW versions we do see the same pattern within the parameters as in the Radar Chart

12)

- The heat map presented in Figure 3 doesn't have a proper figure legend and it is very difficult to understand all the clusters and even to see differences between genotype and groups of mice.

We rearranged the heat map and included a proper legend and explanation (now Fig. 5).

13)

- The authors mentioned in the introduction as a second objective, if a combination of tests measuring locomotor

behaviour is useful for getting a comprehensive phenotyping, or if there are redundancies between tests. This is a very good point to demonstrate indeed. But if I understood right, especially on the table 2 (p13), none of the parameters related to OF or VP could have been clustered with CW parameters. It seems to indicate that each test separately are measuring specific parameters for specific phenotypes which I found interesting and in accordance about what behavioural task has been designed which is to measure specific component of the behaviour. However, how this result can it be interpreted? I am not sure that this point which is for me important is discussed in the discussion part.

We agree with the reviewer. As the single components are always made up by parameters from a single test, we propose that the different tests are not replaceable with each other. As stated above, this confirms the original idea that each test measures distinct aspects of behavior. Nonetheless it is not self-evident that this is actually the case. It would have been possible that activity measures, such as the run duration of the CW and the speed or distance travelled in the OF would cluster together as they measure the same dimension. Also with the IMPC data set we see this separation of tests in the components. Again it would have been entirely possible that the activity measures of SHIRPA, OF and HC would cluster together in one component. Apparently this is not the case, suggesting that other (environmental and/or procedural) factors come into play and alter activity levels according to the test situation. In the discussion we elaborate on this point (page 23, line 23ff).

14)

- Regarding the table 2, why the component presented for the CW data are not similar to the one presented in the table 1 especially when the same number of parameters has been included (113)? Why, the original "spatial dimension" and "temporal dimension" presented in table 1 are respectively named "print dimension" and "step cycle" in table 2? Moreover, why "Stride length" which was originally in Table 1 included in the spatial dimension is in step-cycle dimension in table 2? Does it mean that for different set of data analysed the components can appear to be different? Could this point be critical in terms of reproducibility for an analytic model?

Depending on the input (parameters, tests), the outcome of components can vary. With the first analysis we included 112 parameters only of the CW; in the second analysis we included 147 parameters from the CW (#113), OF (#31) and VP (#3). In both analyses we have the cut off of the number of components used at 80% explained variance. Due to this there are smaller differences of what parameter has its highest loading in what componentand therefor the characterization/naming of the component. Nevertheless, the general dimensions are preservedwhen comparing only the CW components- i.e. spatial and temporal dimensions as well as coordination and its variation. The component "Stride Length" in the first analysis is correlated with C 02 "Paw Contact", a component of the temporal dimension. Therefore it is not too surprising that in the second analysis the" Stride Length" parameters are associated with C 5 "Stand and Stride Length" - again a component of the temporal dimension. Some CW parameters are not independent of each other- i.e. the stride length is influenced by run duration which can also influence stand and swing phase. As such the spatial dimension of the stride length heavily depends on the temporal dimension. We do not think that this is a critical point, as all dimensions of the CW are represented in both analyses. Depending on the input of data (number of parameters and tests) and the processing of this data, the output components can change. Nonetheless we do think that the reproducibility (in terms of taking the same data and re-running the analysis) as well as the replicability (taking a new data set -same tests and parameters assumed) of our data is given.

To strengthen the similarities between the two CW analysis we changed the naming of the dimensions in table 2 (Step Cycle \rightarrow Temporal dimension).

DISCUSSION

15)

- Furthermore, in the discussion part, it is status: "We went on to see how the parameters of several different tests would group into principal components. We did this with a subset of the animals used above for the CW

analysis..." Does it mean that the same animal has been assessed twice with the model? And if it is the case, why was it not possible to get the same correlations and then obtain different component between table 1 and 2?

Yes, the second analysis on CW, OF and VP data was done by including a subset of animals from the first analysis (CW only), because not all animals of the first analysis were also tested on OF and VP.

For the question about the difference between table 1 and 2 please refer to the answer above.

16)

- In the first paragraph of the discussion, it is mentioned that the mice age range for the analysis of the 7.1 version was between 3 and 28 months and for the XT version between 3 and 15 month. Why not having considered in the analysis the same age range for the comparison? This point makes the analysis of the difference between Catwalk 7.1 and XT quite confusing then.

As a starting point we wanted as many animals as possible for the analysis, also because we did not expect a difference between the two versions. As we did see differences between the versions in the analysis, we thereupon analyzed only animals in the same age range (and similar genetic backgrounds) and could still pick up the differences between the versions (see page 22, line 5ff). Therefore we decided to stay with the first approach- by including all animals. This also has the benefit, that analysis on age, body weight and background strain can be done with decent animal numbers.

In conclusion, I can see and understand that the manuscript submitted by Zimprich and collaborator represent an astronomical quantity of analyses and work but is not for now convincing enough to attribute a positive point of view on it yet. The idea for me is really interesting and might be followed but the quality of the manuscript needs to be improved. I am convinced that presented differently the model proposed by Zimprich and collaborator can be really nice and useful.

Additional comments: 17)

17)

- Figure legends are missing it is mainly title instead of proper legends.

We included proper figure legends

18)

- "OF and vertical pole testing was already described mouse elsewhere [10]", the sentence is incorrect or some word are missing. (Page 4, Line 10)

We corrected the sentence.

Reviewer #2: To the Authors

19)

Automated systems for quantifying behaviour may produce such a large number of primary and secondary diagnostic parameters that they can become difficult to evaluate. The authors show that principle component analysis is a useful method to reduce parameters to a much smaller set (10 in this case). Despite the reduction the derived parameters can still be allocated to clearly separable general functions. These were the spatial, temporal,

coordination and variation aspects of locomotion. While this is a useful approach it is not directly new. Other studies have used it before and the current study should reference and discuss this more on a general level.

We included more details/references. As we do not want to discuss PCA on a general level, but in combination with CW, we only focused on papers, where a PCA was described in reproducible detail and run on a behavioral data set. (page 3, line 22ff)

20)

The authors found significant differences between two versions of the Cat Walk apparatus. This information is of little interest to the general reader and I recommend reducing it to a few side remarks.

As the two CW versions have a major impact on our analysis of the data, we do need to explain the major differences. Also still CW 7.1 versions are being used and data published and for reasons of comparability across studies the information we give might be useful. Also refer to Chen et al (DOI: 10.1186/1743-0003-11-62), who found differences between the versions and could only detect subtle changes with the XT version but not the CW 7 in a sciatic nerve injury model. Besides, the potential impact of equipment versions, i.e. equipment specifications, on results, particularly on absolute values, unfortunately still keeps being overlooked in the biomedical field, although it might explain parts of the currently debated "reproducibility crisis". For these reasons we would like to keep this part as it is.

21)

While the authors point out that the derived principal components capture essential general parameters it should be better and more critically discussed to what extent the primary measurements are contained in the derived parameters.

We are not sure if we understand the question correctly. For characterizing the components we used the parameters, which have their highest loading (above |0.5|) in this component. The loadings of the individual parameters can be checked by consulting the structure matrix (Table S3, Supporting Information).

22)

I was surprised to see that two different measures of home cage activity (from the home cage and from the open field) apparently were little correlated so that they resulted in different principal components. On one hand this should be better discussed on the other hand it somewhat questioned for me the wider appliccability of this approach.

Yes, we were also surprised to see that the activity measures of the HC and OF (and SHIRPA) were not included in the same component. We now elaborate on this point in the discussion (see page 23, line 37). Please note that the 20 min open field test does not measure home cage activity, but spontaneous locomotion and exploratory activity in a novel environment.

We do not think that this result questions the applicability of the current approach, but it shows that the different test paradigms have a great influence on the parameters measured- even if the parameters are claimed to measure the same quality (in this case activity). Our results illustrate that the circumstances under which activity measures are recorded determine what they reflect. For a further discussion please refer to question 13.

23)

The major result of the study seems to be that results from mouse strains differ. What I find lacking is to what extent the analysis helped in distinguishing for the specific studies performed the experimental animals from the control animals. This is the major question that a researcher wants to answer when performing these measurements.

With our analysis we suggest that for distinguishing between mutant and wild type the measures of FP and HP (for the temporal and spatial dimensions) would be enough and either the couplings or phase dispersion would suffice for a first analysis, thereby reducing the amount of parameters from 113 to 44 (page 21, line 10ff; page 8, line 29ff). The CW can clearly detect differences between control and experimental animals. In the 25 mutant mouse lines used in our analysis 21 of them show a genotype-specific phenotype. This high phenotype- rate in mutant mouse lines subjected to the CW is due to a hypothesis-driven selection of mouse lines (page 23, line 18ff).

To get a comprehensive picture of the "locomotor phenotype" of the mouse line analyzed it is necessary to include a few tests. For instance the CW could be included to analyze the gait, OF, SHRIPA and HC for the activity measures under different circumstances as well as RR, VP and GS for motor performance and muscle strength. All these tests add to the different aspects of locomotor phenotypes (page 23, line 23ff).

In respect to large-scale phenotyping, where the issue of refining and reducing of animals and tests comes into play, we can clearly state that, at least for "locomotor phenotypes", all of the above mentioned tests have their validity and that there are no redundancies between them.

Specific comments 24) OF open field -> spell out on first use: OF, HC, RH?, others?

We corrected this.

25)

The manuscript would significantly benefit from having a professional English language editor go through it. Since this is available for around 200-400 EUR the authors should not shy that cost.

We now have had a native English speaker critically reading and correcting the manuscript.

26)

This study proposes the use of Principal Component Analysis as a "new method" to reduce dimonsionality and complexity in data sets gathered for behavioural phenotyping by automated instruments. However, PCA is a standard method that has been used in other studies. Thus, I think a general comparison on the use of PCA for behavioural phenotyping by comparing the general merit of this approach with other such studies is warranted. E.g. Vannoni E, Voikar V, Colacicco G, Sánchez MA, Lipp HP, Wolfer DP. 2014. Spontaneous behavior in the social homecage discriminates strains, lesions and mutations in mice. J Neurosci Methods. 30;234:26-37. doi: 10.1016/j.jneumeth.2014.04.026. OR Ohl F, Roedel A, Binder E, Holsboer F. 2003. Impact of high and low anxiety on cognitive performance in a modified hole board test in C57BL/6 and DBA/2 mice. Eur J Neurosci. 17:128-136.

We did not want to indicate that PCA is a new method, but want to show that it is a sound method to reduce dimensions on highly dimensional data sets, such as the CW; thus we edited this part of the Introduction (page 3, line 22 ff) and Discussion (page 20, line 23 ff). We included the paper from Ohl et al to the mentioned literature (Vannoni et al are already named). We are not aware of any publication refining CW parameters by PCA.

27)

The fact that the two Cat Walk versions give different results is of interest mainly to the authors who know and have used both and maybe to the manufacturer Noldus but is not really of general interest in the context of the method "PCA" to a wider audience. Thus, the sections devoted to the cat walk comparison should be reduced to a

few side remarks. I think it would be sufficient to point out that there is a difference but then concentrate on just one version, e.g. the recent version of the cat walk. Maybe much of the comparative data can be put into a supplement outside the main article.

We do think that the power of this analysis comes from the high amount of animal numbers differing in genotype, sex, body weight and age. Only with this high amount of animals it is possible to analyze age or body weight effects in such detail. Please also see our reply to question 20.

28)

"Running a PCA on data sets to extract individual component scores is a useful tool in interpreting large data sets. " -> rather "highly dimensional / multidimensional" data sets or similar

We changed this expression.

29)

"it is necessary to include other locomotor tests for a comprehensive phenotyping." -> surely, you do not mean "general behavioural phenotyping" but "locomotor phenotyping". Also: since the only other measure that you really suggest with evidence is "activity" this could maybe mentioned explicitly instead of just "other locomotor tests".

Yes, that is correct, we do mean locomotor phenotyping. We changed the sentence to: "Furthermore, although the CatWalk is sensitive for detecting locomotor phenotypes pertaining to gait, it is necessary to include other tests for comprehensive locomotor phenotyping." (page 2, line 22f). As we use the term "locomotor phenotype" in a very broad way (Reviewer 1 suggested to include a definition- we did so; page 3, line 29ff), which includes gait, activity as well as motor performance, we would prefer keeping it the way it is and not change it to "activity".

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Journal of Neuroscience Methods

Special Issue "SI: Measuring Behavior 2016"

Analysis of Locomotor Behavior in the German Mouse Clinic

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Highlights

- Assessment of comprehensive locomotor phenotyping strategy
- Refinement of highly dimensional data sets by Principal Component Analysis
- Influence of equipment version, sex, body weight, age and genetic background
- No redundancies detected between different locomotor tests

1 Journal of Neuroscience Methods

2 Special Issue "SI: Measuring Behavior 2016"

3

4 Analysis of Locomotor Behavior in the German Mouse Clinic

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1 Abstract

2 Background

3 Generation and phenotyping of mutant mouse models continues to increase along with the search for

4 the most efficient phenotyping tests. Here we asked if a combination of different locomotor tests is

5 necessary for comprehensive locomotor phenotyping, or if a large data set from an automated gait

6 analysis with the CatWalk system would suffice.

7 New Method

8 First we endeavored to meaningfully reduce the large CatWalk data set by Principal Component Analysis 9 (PCA) to decide on the most relevant parameters. We analyzed the influence of sex, body weight, 10 genetic background and age. Then a combination of different locomotor tests was analyzed to 11 investigate the possibility of redundancy between tests.

12 Result

13 The extracted 10 components describe 80% of the total variance in the CatWalk, characterizing different

14 aspects of gait. With these, effects of CatWalk version, sex, body weight, age and genetic background

15 were detected. In addition, the PCA on a combination of locomotor tests suggests that these are

16 independent without significant redundancy in their locomotor measures.

17 Comparison with existing methods

18 The PCA has permitted the refinement of the highly dimensional CatWalk (and other tests) data set for

19 the extraction of individual component scores and subsequent analysis.

20 Conclusion

21 The outcome of the PCA suggests the possibility to focus on measures of the front and hind paws, and

22 one measure of coordination in future experiments to detect phenotypic differences. Furthermore,

23 although the CatWalk is sensitive for detecting locomotor phenotypes pertaining to gait, it is necessary

24 to include other tests for comprehensive locomotor phenotyping.

25

- 26 Key words: CatWalk; Gait; Principal Component Analysis; Mouse; Locomotion; Activity; Phenotyping
- 27

28 Introduction

Recent advances in genome editing technology have revolutionized and accelerated the generation of mutant mouse lines for modeling human disease. This technological cataclysm has brought with it an upsurge in demand for comprehensive behavioral phenotyping that is reflected in the rising number of large-scale phenotyping centers and consortia [1,2]. To meet the demand, new sophisticated phenotyping devices have been developed to provide an automatic and detailed characterization of mice. The goal of automation is not only to speed up data acquisition and to make it more objective and replicable, but also to reduce the burden on the animal through less handling. Ideally, a few automated systems would be sufficient to comprehensively phenotype a mouse model. The questions repeatedly arising in this context are how valid are individual tests; which test is the most informative and how many tests are really needed. It would also be beneficial to know if parameters of one test could predict the utility of another test for more detailed phenotypic characterization. Such knowledge would enhance the efficacy of phenotyping strategies, which have to be outlined in license applications for experiments on animals, thus saving time, animals and money.

7 Here we explore these questions with respect to locomotor phenotyping. Would it be sufficient, for 8 example, to use the Catwalk (CW) from Noldus; an automated gait analysis system with video based 9 paw tracking [3,4]? Previously the assessment of gait parameters involved footprint analysis, where the 10 paws were dipped into ink and the animal walked across a paper. Approximately 20 parameters were 11 measured from these prints, however with the automated CW system, around 230 parameters are 12 measured. With this dramatic increase in parameters/data points data analysis has become more 13 complex. Traditionally only single parameters or a few selected ones were analyzed. But it is hard to 14 formulate a clear picture of the phenotype when dealing with a large amount of parameters for one 15 test. Thus it is necessary to decide which of the parameters give most information and is the best for 16 detecting disease-relevant phenotypes. Alternatively, a compression of the data may be more 17 appropriate to garner a comprehensive overview. As such reducing the number of parameters under 18 consideration (aka dimensions) to a few components (i.e. artificial variables) by Principal Component 19 Analysis (PCA) improves data handling without significant information loss [5-9]. The subsequent analysis then runs on the extracted components that substitute for the more numerous original 20 21 parameters.

Running a PCA on behavioral data is not new; several groups have used it for analyzing data [5-9]. Yet running a PCA on a multidimensional data set (a high amount of animals and parameters, i.e. over 1000 cases and over 100 parameters, respectively) is not frequently done. With a few exceptions animal numbers are moderate (below 100 animals) and parameters analyzed small (below 50). To answer the above mentioned questions we apply a PCA to a multidimensional data set as collected by the CW, which, to our knowledge, has not been done so far. By including a fairly large sample size (over 1000 cases) we expect to produce reliable results, as factor instability is less likely to occur in a larger sample.

29 In this paper we focus on locomotor phenotypes, whereby the definition of a "locomotor phenotype" is 30 used very broadly. We include gait phenotypes (as analyzed by the CW), as well as activity (as in the 31 Open Field (OF), home cage (HC) and SHIRPA) and motor ability phenotypes (as in the Grip Strength 32 (GS), Rotarod (RR) and Vertical Pole (VP) test) into this term. Gait phenotypes describe how the animal 33 moves, activity phenotypes describe how frequently the animal moves, and motor ability phenotypes 34 describe in this case muscle strength, balance and coordination. To test for a disease-relevant locomotor 35 phenotype generally one to several different tests are applied. Here we assess if the CW alone, with its 36 large amount of parameters, can be sufficient to detect locomotor deficits and if a reduction in 37 dimensions is appropriate to detect meaningful components. With historical data from our lab we ran a 38 PCA to reduce the dimensions of the CW data. Furthermore, we characterized the resulting components 39 and analyzed them with different statistical methods to evaluate the impact of different versions of the 40 CW system (we have used two different systems in our lab), sex, body weight, age and genetics.

Likewise we ask if a combination of tests measuring locomotor behavior is necessary for comprehensive locomotor phenotyping, or if there are redundancies between tests. To answer these questions we included two data sets. In the first set we took historical data from our lab including parameters from the CW, OF and VP test. The second data set comprises publicly available data from the International Mouse Phenotyping Consortium (IMPC; [10]) website (https://www.mousephenotype.org) selecting different measures of motor performance (horizontal as well as vertical activity, muscle strength and coordination), which included parameters measured in the OF, SHIRPA, RR, GS and in the HC during indirect calorimetry at the German Mouse Clinic. With both data sets a PCA was run to see how the different parameters are distributed among the extracted components, thereby assessing possible

- 8 redundancies between different tests.
- 9

10

11 Methods:

12 Behavior:

For all tests mice are transferred to the testing room at a minimum of 30 minutes prior to testing. Mutants are always concurrently tested with respective wild types and males prior to females. After each mouse the test apparatus is cleaned and disinfected. Different mutant lines were used in the behavioral assessments. For sample sizes by sex and genotype please refer to Table S1 in the Supporting Information. All experiments were approved by the government of Upper Bavaria, Germany.

18

19 CatWalk:

20 Mice were tested on two versions of the CW system (Noldus Information Technology, Wageningen, 21 Netherlands): the CW 7.1 version and the CW XT 10.5 version (hereafter referred to as CW XT). On both 22 systems the mouse traverses an elevated glass walkway that is bordered by Plexiglas walls in a darkened 23 room. A camera situated underneath the middle of the walkway tracks the illuminated footprints, which 24 are then analyzed with the CW software. For each animal the mean of 2 to 4 uninterrupted and 25 continuous runs (each included approx. 4-6 step cycles) were calculated and used for the following analysis. The major difference between the two versions is the camera: for the older 7.1 version the 26 27 camera captures the footprints at a rate of up to 50Hz and for the XT version a high speed camera is 28 used, capturing at a rate of up to 100Hz. The CW system measures various aspects of paw contacts with 29 the floor in a dynamic way and automatically calculates a broad number of spatial and temporal gait 30 parameters in several categories. These include i) parameters related to individual paw prints, such as 31 the width and length, together with a calculation for the front and hind paws (FP and HP, respectively); 32 ii) parameters related to the position of paw prints with respect to each other, for example the stride 33 length and the base of support (BOS), which measures the width between the two FPs and the two HPs 34 respectively; iii) parameters related to time-based relationships between paw pairs (couplings and phase 35 dispersion) and their variation, as well as step patterns. For a more detailed description of the CatWalk 36 method see [3]. Table S2 lists the selected parameters for the statistical analysis.

- 37
- 38 Open Field and Vertical Pole:

OF and VP testing was already described elsewhere [11]. In brief, mice are placed in a square OF arena surrounded by infra-red beams (ActiMot, TSE, Bad Homburg, Germany) and behavior is recorded for 20 minutes. Several parameters are analyzed for the whole 20 minutes and four of them (distance travelled, number of rearings, time in center [%] and distance in center [%]) are also analyzed in 5

42 minute bins (Table S2). The VP is a 50cm high, taped pole (diameter 1cm) where the mouse is placed

1 head upwards at the top. Time to turn and for complete descending is recorded. The time from turning

- till complete descent is calculated ("TimeDown", Table S2). Animals receive 2-3 training trials and 3-5
 test trials with 5-10 minutes inter-trial intervals.
- 4

5 <u>IMPC phenotyping</u>:

- 6 Mice from the IMPC consortium undergo a standardized phenotypic screening that involves several
- 7 different biological and clinical aspects- including behavior and neurological endpoints. In this study we
- 8 included parameters from the OF, SHIRPA, RR, GS as well as the indirect calorimetry (see Table S7) from
- 9 the German Mouse Clinic only. For the generation of behavioral data for IMPC lines please see
- 10 https://www.mousephenotype.org/impress/procedures/14 on the IMPC webpage. In brief, OF testing is
- 11 performed as mentioned above.
- 12 Rotarod:
- 13 Motor Performance is tested with a commercially available Rotarod apparatus (Bioseb, Chaville, France).
- 14 The test phase consists of three trials separated by 15 minute inter-trial intervals. The mice are placed
- on the rotating rod and the apparatus accelerates from 4 to 40rpm in 300 seconds. The latency to fall is
- 16 recorded as well as the number of passive rotations.
- 17 SHIRPA:
- 18 The primary observation screen is a modification of the Irwin procedure [12]. The mouse is placed in a
- 19 clear cylinder to observe tremors and body positioning. Then the mouse is transferred into an arena
- 20 (420 x 260 x 180mm) in which a Perspex sheet on the floor is marked with 15 squares. Here several
- other parameters are taken; one being locomotor activity which is recorded in the first 30 seconds by
- 22 counting the number of squares being crossed.
- 23 Grip Strength:
- A grip strength meter (Bioseb, Chaville, France) is used. The mouse is lowered towards the grid, so that
- the FP cling onto it. By slowly pulling back the mouse the maximum strength is recorded. Then the
- 26 mouse is lowered so that both FP and HP attach; again the maximum grip strength is recorded. For both
- 27 measurements the mean of three measurements are calculated. Body weight is recorded. For the PCA
- the grip/body weight ratio is used.
- 29 Home Cage activity:
- 30 While the animals are tested for indirect calorimetry (PhenoMaster System, TSE Systems, Germany),
- 31 activity data (horizontal and vertical) and speed are also recorded. Here animals are singly housed for at
- 32 least 21h.
- 33
- 34
- 35 Data sets:

36 Animals were only included in the analysis, when all parameters of all tests were present.

- 37
- 38 CW data:
- 39 Data included 1499 cases and 112 parameters. We included data from two CW versions (i.e. the 7.1 and
- 40 the XT version) integrating only parameters which both CW versions collect (see Table S2).
- 41
- 42 CW-OF-VP data:
- 43 For the investigation of redundancies between tests we took a data set consisting of animals of the "CW
- data", which were also tested in OF and VP. In total 1057 cases and 147 parameters from the CW, OF
- 45 and the VP test were collected (Table S1 and S2).
- 46
- 47 IMPC data:

1 We also wanted to investigate an independent data set for redundancies between tests assessing 2 locomotor phenotypes: For this we took a data set, which included 1327 animals and 20 parameters 3 from the OF, SHIRPA, RR, GS and HC activity during indirect calorimetry (Table S1 and S7). Data was 4 taken from our internal database but it is also accessible for the public via 5 http://www.mousephenotype.org/. As we could not include all animals measured (due to excessive 6 animal numbers) we decided to pick mouse lines upon extremes in activity based on 5 parameters (OF-7 distance, OF-rearings, SHIRPA-locomotion, HC-distance and HC-rearings). The mouse lines were selected based on the effect size (Cohen's D (d= (mean_(mutants)-mean_(controls))/ v((SD²_(mutants) +SD²_(controls))/2))) - all 8 9 lines below the 5th and above the 95th percentile were chosen.

- 10
- 11

12 Statistics:

All statistical analysis was performed with PASW Statistics 18 (Version 18.0.0; SPSS Inc., Chicago, USA), if
 not mentioned otherwise, and a p-value ≤0.05 was considered as statistically significant.

15

16 *Principal Component Analysis:*

17 We conducted three PCAs: a) with CW data, b) with CW, OF and VP and c) with IMPC data.

18 Generally the handling of data for PCA was as follows:

All parameters of the individual data sets were standardized to a mean of 0 and a standard deviation of 1 via Z-transformation, to be able to compare between parameters. Generally, if the scales between parameters differ greatly and no transformation is being done, then parameters with a large scale can

22 dominate the outcome. Thus, in an exploratory analysis it is recommended to standardize the data set.

23 The sampling adequacy was confirmed by the Kaiser-Meyer-Olkin (KMO) measure and the Bartlett- Test.

As we do not know if our components are dependent or independent of each other, we chose the

25 oblique rotation (via the oblimin method) thereby allowing correlation between components. For the

extraction of components two criterions were set; firstly, only those with eigenvalues greater than 1
were chosen and secondly, the scree plot was consulted or a cut off at 80% of explained variance was

used. As the scree plot becomes difficult to interpret with a high amount of parameters (i.e. for the first

two analyses; CW data: 112 parameters and CW-OF-VP data: 147 parameters), we used for the first two

30 data sets the cut off at 80% of explained variance. For the third data set we could confirm the extracted

31 components by consulting the scree plot, as here fewer parameters were measured (20 parameters; for

32 the scree plot see Fig. S1). Individual component scores were calculated. Components were

33 characterized according to the highest loadings (>|0.5|) of individual parameters.

34 35 (

CW data: 36 With this data set we conducted further analysis to investigate effects of sex, CW version, body weight, 37 age and genotype. For analyzing sex and CW version effects a two-way ANOVA was applied. For the 38 correlation with body weight the Pearson correlation coefficient was used, including only animals where 39 body weight was measured within a week of CW testing. Age effects were evaluated by the Pearson 40 correlation applied to all animals experiencing the CW for the first time and including only wild type 41 animals. A linear regression model was used for analyzing age, sex and body weight influences on 42 components. The medians of the individual component scores of each group were clustered in a 43 hierarchical agglomerative cluster analysis using R (Version 3.0.2, R Foundation for Statistical 44 Computing, Vienna, Austria) to investigate similarities between the lines, the corresponding genotypes and sex (e.g. Group 1: male wild types of mouse line A; Group 2: male mutants of mouse line A). The 45 46 Ward method was used as a linkage criterion and the Euclidean distance as metric. This method allows

47 the classification of groups on the basis of trait similarities.

1 2	
3	Results
4	

5 PCA on historical CW data

6 <u>Characterization of the ten principal components revealed the possibility to focus on a reduced</u> 7 <u>parameter-set in future analysis</u>

8 We firstly ran a PCA only on the CW data set to reduce dimensions and to analyze the impact of sex, 9 body weight, age and genotype. Historical data of 1499 cases was used for the PCA analyzing CW data. 10 As there was a change of the CW version (from 7.1 to XT) in our institute we took data from both, 11 analyzing only parameters that were measured by the two. After the omission of one parameter due to 12 low correlations in the correlation matrix ("StepSequence_Rb"), we analyzed 112 parameters. The 13 sampling adequacy of the data was done via the KMO criterion and Bartlett's test and found to be 14 appropriate (KMO= 0.810; Bartlett- Test: Chi²= 522783.471, df= 6216, p<0.001). Ten components were 15 extracted via oblique rotation describing 79.5% of the total variance. In other words, we were looking 16 for the components that accounted for the most variation and hence the most likely to reveal 17 differences between the cases and these were the ten revealed. The loadings, which can be described as 18 the correlation coefficient between the parameter and the components, are depicted in the structure 19 matrix table (see Table S3). Components were characterized according to the highest loadings (above 20 [0.5]) of the individual parameters (see Table 1).

- 21
- 22
- 23 Table 1: CW PCA: Component characterization according to the major parameters. The components are

arranged due to their representation of different gait dimensions. The explained variance is given for
each component and for the dimensions.

		Component name	Major parameters (with loadings > 0.5)	Explain varianc		
Spatial dimension	C 01	Print Dimensions	Print Length, Print Width, Print Area, Maximum Contact Area	23.2%	26%	
Spa dime	C 08	Stride Length	Stride Length	2.7%	20%	
le n	C 02	Paw Contact	Stand Duration, Duty Cycle, Run Duration, Support Diagonal, Support Single	14.7%		
Temporal dimension	C 04	Swing Phase	Swing Duration, Swing Speed	8.7%	27%	
	C 07	Turning point	Max Contact At	3.4%		

Interlimb coordination	C 05	Coordination of diagonal and ipsilateral pairs	Coupling and Phase Dispersion of the diagonal and ipsilateral pairs, Alternate step patterns	6.1%	11%	
Inte coord	C 06	Coordination of girdle pairs	Couplings and Phase Dispersion of the girdle pairs, Cruciate step patterns	4.5%		
rlimb	C 03	Variation incl RH	Variation in coordination of paw pairs including the RH, Regularity Index	12.2%		
Variation in interlimb coordination	on in inte ordinatio 60 O	Variation excl RH for diagonal and ipsilateral pairs	Variation in coordination of diagonal and ipsilateral pairs not including the RH	2.1%	16%	
Variati coo	C 10	Variation for FP girdle pairs and Print Position	Variation in coordination of FP girdle pairs, Print Position	1.9%		

1

RH- right hind paw; FP- front paws

The extracted components can be attributed to different dimensions of gait performance, such as
 spatial and temporal aspects as well as interlimb coordination and its variation.

Component C 01 *Print Dimensions* (describing 23% of the total variance) contains nearly all
spatial print dimensions, which includes the print length and width as well as the area. Component C 08 *Stride Length* (describing 3% of the total variance) accounts for the rest of the spatial measures.
Together these components explain 26% of the total variance.

8 The temporal dimensions are represented by the components C 02 *Paw Contact* (which includes 9 the stand duration, duty cycle as well as the support and run duration, describing 15% of the total 10 variance), C 04 *Swing Phase* (which includes the swing duration and swing speed, describing 8% of the 11 total variance) and C 07 *Turning Point* (which contains the turning point measured by the "Maximum 12 contact at %", describing 3% of the total variance), which explain together 27% of the total variance.

The interlimb coordination is represented by two components, C 05 *Coordination of diagonal* and ipsilateral pairs (including the alternate step patterns (see also Table S2), describing 6% of the total variance) and C 06 *Coordination of girdle pairs* (including the cruciate step patterns, describing 5% of the total variance) explaining together 11% of the total variance.

Another three components, C 03 *Variation including RH* (right hind paw) (contains also the regularity index, describing 12% of the total variance), C 09 *Variation excluding RH for diagonal and ipsilateral pairs* (describing 2% of the total variance), and C 10 *Variation for FP (front paw) girdle pairs and Print Position* (describing 2% of the total variance) represent the variation in the interlimb coordination explaining together 16% of the total variance.

The consequence of allowing for correlations between components by the oblique rotation can be seen in the component correlation matrix (see Table S4). We see slight correlations between some components. The strongest correlation occurs between two components describing the variation; C 03 *Variation including RH* and C 09 *Variation excluding RH for diagonal and ipsilateral pairs* (r= -0.490). There is also a correlation between C 01 *Print Dimensions* and C 10 *Variation FP girdle pairs and Print Position* (r=-0.306), and a weaker one between C 02 *Paw Contact* and C 08 *Stride Length* (r=-0.252).

For the spatial and temporal parameters we have measures from all four paws as well as the combinations for the FPs and HPs (hind paws) respectively (see Table S2 and Fig. 2). All six measures of the same parameter always coincide within the same component (see Table S3); suggesting that analysis of only FP and HP could suffice in future analysis (in contrast to analyzing all single paws). For two 1 parameters we have only measurements of the FP and HP, and left and right side, respectively; these

are the base of support (BOS) (FP and HP) and the print position (left and right side). The print position

3 for left and right side integrated in the same component (C 10). Interestingly, the BOS for FP and HP are

within different components, i.e. the BOS FP has its major loading to C 05 *Coordination of the diagonal and ipsilateral pairs* (-0.469) but also loads onto C 02 *Paw Contact* (0.420) and C 08 *Stride Length* (-

6 0.374) with a similar weight; the BOS HP has its highest loading on C 04 Swing Phase (0.339). Another

7 noteworthy point is that the BOS of FP and HP do not correlate $(r_{(1498)}=0.046)$ and as such have to be

8 regarded as independent parameters, which is also illustrated by the different integration into a

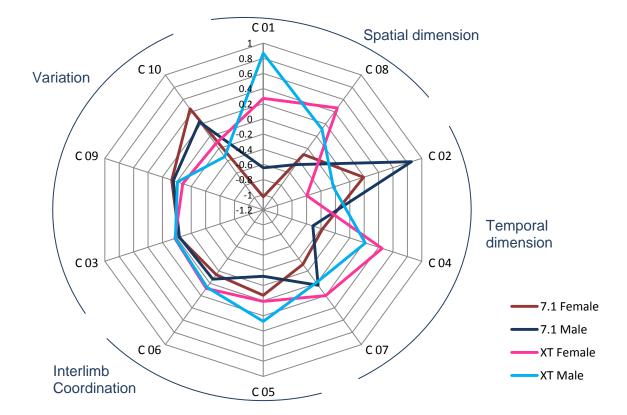
9 component and the differential pattern of loadings onto other components.

10

11 <u>Highly significant differences between two CW versions</u>

12 Over the course of time newer versions of a test apparatus might be installed in a lab and the question 13 arises of how comparable these versions are. This is especially important when comparing mouse lines 14 and experiments before and after such a change has occurred. As we switched from the CW 7.1 to the 15 XT version, we employed the data set to analyze CW version effects and sex effects by a two-way 16 ANOVA. The results showed interactions between sex and CW version for C 01 Print Dimensions, 02 Paw 17 Contact, 05 Coordination of diagonal and ipsilateral pairs and 07 Turning Point ($F_{(1,1495)}$ =9.109, p=0.003; F_(1,1495)=11.119, p=0.001; F_(1,1495)=25.248, p<0.001; F_(1,1495)=28.247, p<0.001, respectively). We saw highly 18 19 significant differences between the CW versions in all components except C 03 Variation incl RH, which 20 is not significant (C 01 F_(1,1495)=1579.028, p<0.001; C 02 F_(1,1495)=443.631, p<0.001; C 04 F_(1,1495)=259.273, 21 p<0.001; C 05 F_(1,1495)=43.622, p<0.001; C 06 F_(1,1495)=11.564, p=0.001; C 07 F_(1,1495)=20.636, p<0.001; C 08 22 F_(1,1495)=186.445, p<0.001; C 09 F_(1,1495)=3.952, p=0.047; C 10 F_(1,1495)=128.866, p<0.001). Fig. 1 depicts 23 these differences in a Radar Chart. Thus we continued by separately analyzing the data of the two 24 versions. To further illustrate this difference between the two versions we plotted the original data of 25 the C 01 Print Dimensions parameter "Print Area" and for C 08 Stride Length "Stride Length" (Fig. 2). We 26 see a larger print area and a longer stride length in the CW XT version compared to the 7.1 version. 27 These results show that a highly significant difference between the two CW versions exists and that 28 caution has to be taken when comparing data of the two.

29



1

2 Fig. 1: Radar Chart depicting the differences between CW version and sex (shown are the means per

- 3 group; $n_{(7.1,M)}$ =344; $n_{(7.1,F)}$ =283; $n_{(XT,M)}$ =464; $n_{(XT,F)}$ =408). Notice that components are grouped according to
- 4 their representation of different gait dimensions.
- 5

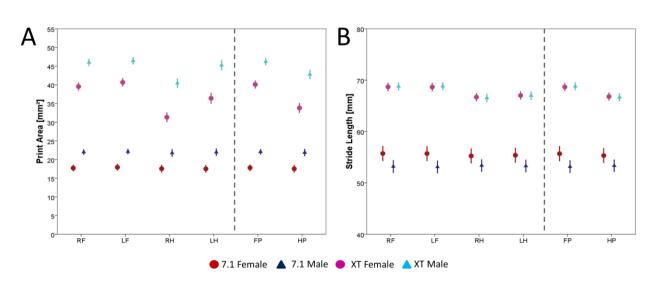




Fig. 2: Original data for "Print Area" (A) and "Stride Length" (B) of all paws, inlcuding front paws and
hind paws, to illustrate the differences between the two CW versions. Both "Print Area" and "Stride

9 Length" have higher values in the XT version. Shown are the means and the 95% confidence interval. RF-

10 right front paw, LF- left front paw, RH- right hind paw, LH- left hind paw, FP- front paws, HP- hind paws

1

2 Sex and body weight effects on print dimensions in both CW versions

3 In biomedical research sex effects are often ignored or not thoroughly described, although it is clear that 4 sex differences exist in several different behaviors and the prevalence, severity and etiology of (human) 5 diseases [13-18]. Also body weight can have an effect on gait parameters. To investigate sex and body 6 weight effects we conducted an ANOVA and calculated the Pearson correlation, respectively. As we saw 7 effects of both factors in some components we evaluated their influence with a multiple linear 8 regression model. The analyses occurred separately for both CW versions.

9 Sex differences appeared in the 7.1 version for C 01 Print Dimensions, C 02 Paw Contact, C 05 10 Coordination of diagonal and ipsilateral pairs, C 07 Turning Point and C 10 Variation FP girdle pairs and Print Position (C 01 F_(1,625)=70.479, p<0.001; C 02 F_(1,625)=54.920, p<0.001; C 05 F_(1,625)=8.010, p=0.005; C 11 12 07 $F_{(1.625)}$ =13.407, p<0.001; C 10 $F_{(1.625)}$ =4.864, p=0.028). For the XT version all components of the spatial 13 and temporal dimensions, as well as in the coordination of the diagonal and ipsilateral pairs show a 14 significant difference between the sexes (C 01 F_(1,870)=140.393, p<0.001; C 02 F_(1,870)=89.750, p<0.001; C 04 $F_{(1,870)}$ =21.799, p<0.001; C 05 $F_{(1,870)}$ =19.993, p<0.001; C 07 $F_{(1,870)}$ =13.599, p<0.001; C 08 15

- 16 F_(1,870)=47.880, p<0.001; C 10 F_(1,870)=17.327, p<0.001).
- 17 We went on to analyze the data for correlations between the components scores and the body weight 18 in the two CW versions separately. Only data from animals where body weight was measured within one 19 week of CW testing was included. As the number of animals was still very high (7.1 version: n= 364; XT 20 version: n=872) we got significant correlations for nearly all parameters. Thus we paid little attention to 21 the p-values and assessed the correlation only via the correlation coefficient. The only correlations that 22 are moderately strong in both CW versions are between body weight and C 01 Print Dimensions (7.1: 23 r₍₃₆₃₎= 0.293; XT: r₍₈₇₁₎= 0.417) and C 02 Paw Contact (7.1: r₍₃₆₃₎= 0.399; XT: r₍₈₇₁₎= 0.503) respectively, 24 suggesting that only those components are really influenced by the body weight. For the 7.1 version we 25 also see a mild correlation with C 07 Turning Point ($r_{(363)}$ = 0.312). All other correlation coefficients are 26 below [0.253] and therefore do not suggest a noteworthy influence of body weight on the component.
- 27 As noted above there is an influence of sex on C 01 Print Dimensions, C 02 Paw Contact and C 07 Turning
- 28 Point. To evaluate whether in fact the sex or rather the body weight as a confounder variable influences
- 29 the component we ran a multiple linear regression analysis on a subset of animals ($n_{(F,7.1)}=170$; $n_{(M)}$ 30
- 7.1)=194; n_(F XT)=408; n_(M XT)=464), where animals were measured for their body weight within one week of
- 31 CW testing and evaluated scatter plots. For the 7.1 version 11% of the variance in C 01 Print Dimensions 32 can be explained by the linear regression model and here both sex and body weight have a significant
- 33 influence (Fig. 3A; $F_{(2.361)}$ =21.616, p<0.001, R²=0.107, body weight: β =0.223, p<0.001, sex: β =-0.161,
- 34 p=0.004). For C 02 Paw Contact and C 07 Turning Point only 17% and 10 % of the variance, respectively,
- 35 can be explained by the model. For both components body weight has a significant influence, whereas
- 36 sex has no impact (C 02: $F_{(2,361)}$ =35.927, p<0.001, R²=0.166, body weight: β =0.359, p<0.001, sex: β =-
- 37 0.093, p=0.082; C 07: F_(2.361)=19.622, p<0.001, R²=0.098, body weight: β=0.300, p<0.001, sex: β=-0.029,
- 38 p=0.606). For the XT version we saw that C 01 Print Dimensions is influenced by both body weight and
- 39 sex (Fig. 3B; $F_{(2.869)}$ =111.077, p<0.001, R²=0.204, body weight: β =0.304, p<0.001, sex: β =-0.207, p<0.001)
- 40 explaining about 20% of the variation. In C 02 Paw Contact 25% of the variation can be explained by the
- 41 model and only body weight has a significant influence ($F_{(2,869)}$ =147.977, p<0.001, R²=0.254, body weight:
- 42 β =0.479, p<0.001, sex: β =-0.044, p=0.209). We correlated body weight to C 01 Print Dimensions

- separated by sex and CW version and saw positive correlations for the females in both versions (7.1: $r_{(170)}=0.331$; XT: $r_{(408)}=0.328$ respecitvely) and in the males of the XT version ($r_{(464)}=0.220$), but no correlation for the males of the 7.1 version ($r_{(194)}=0.104$) (see Fig. 3A and B). A similar picture appears when analyzing the original data (in Fig. 3C and D "FP Print Area", a parameter of C 01 is depicted).
- 5 In summary we detect sex effects in several components, especially in the CW XT versions. In C 01 Print
- 6 Dimensions we see an effect of sex as well as body weight. This is interesting to note, as it is intuitive
- 7 that the size of paw prints would correlate with body weight, but less intuitive that there is a sex effect.
- 8 For C 02 Paw Contact the analysis suggests an influence of body weight and not of sex. As body weight
- 9 increases with age, we explored the influence of age on the components in a next step.

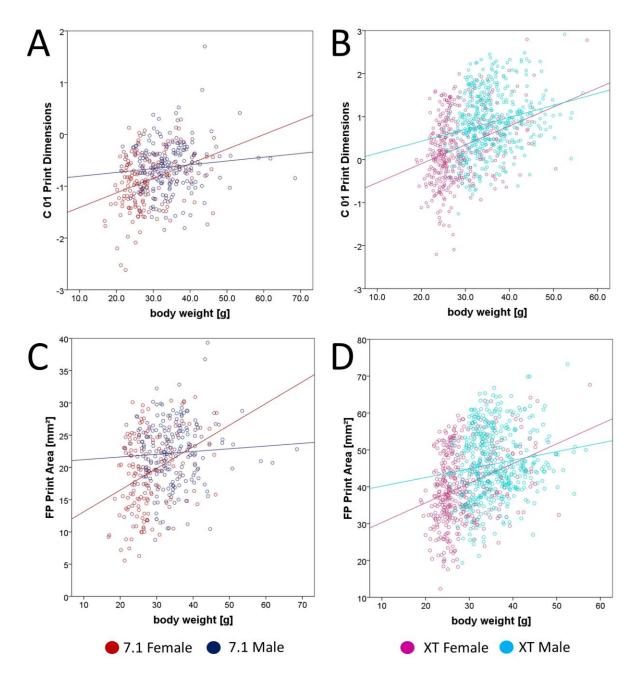


Fig. 3: Scatter plots for correlation between C 01 and Print Area with body weight for the 7.1 version and
 XT version. In the 7.1 version (A) there is an influence of sex and body weight on the component. When
 correlating body weight with C 01 per sex, a positive correlation is seen in the females but not in the

4 males. In the XT version (B) we see a positive correlation both for males and females in C 01. In Print Area

- 5 of the FP, a parameter of C 01, we can see the correlation with body weight also in the original data for
- 6 both the CW 7.1 (C) and XT (D). FP- front paws
- 7

8 Paw contact increases with age in both CW versions

9 Age can play an important role when analyzing gait parameters. Not only does the body weight change

10 with age (and thus some gait-related parameter change- see above), but also a decline in muscle mass

and neuromuscular dysfunction, amongst other changes, can occur, thus leading to an altered gait

12 performance. To investigate possible age effects we calculated the Pearson correlation coefficient and

specifically checked the influence of age and body weight on C 02 *Paw Contact* in both versions.

14 We included in the analysis only wild type animals that were subjected to the CW for the first time to 15 circumvent possible age-mutation interactions. Still we separately analyzed the CW versions and ran a 16 correlation analysis for age and the different components. Again we interpreted only the correlation 17 coefficient and paid less attention to the p-values, as they were nearly all significant due to the high 18 animal number (7.1 n= 209; XT: n= 198). C 02 Paw Contact showed a positive correlation to the age of 19 the animal in both versions (7.1: $r_{(208)}$ = 0.367, Fig. 4A; XT: $r_{(197)}$ = 0.500). The question again arises if the 20 age effects we see in C 02 Paw Contact are due to age and not to body weight differences. For this we 21 ran a multiple linear regression analysis for C 02 Paw Contact based on age and body weight. In the 7.1 22 version 16% of the variation can be explained by the model ($F_{(2,95)}$ =10.461, p<0.001; adjusted R²= 0.163), 23 but here only age was found to be significant (β =0.348, p=0.001; body weight: β =0.150, p=0.133). This is 24 illustrated by "HP Stand" (Fig. 4B); an increase in stand duration by age can be seen, but it is not a very 25 dramatic one. For the XT version both predictor variables explain 34% of the variation ($F_{(2.195)}$ =52.056, 26 p<0.001; adjusted R²= 0.341). Both body weight (β =0.362, p<0.001) and age (β =0.318, p<0.001) were 27 significant predictors of C 02.

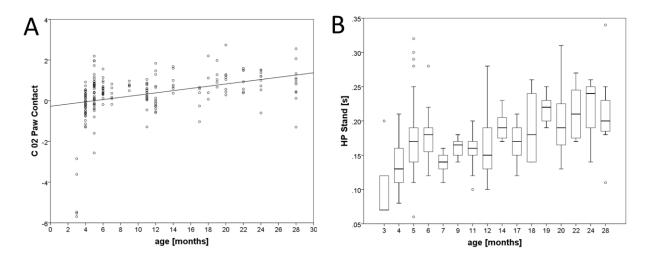


Fig. 4: Plots for depicting correlations between age and C 02 Paw Contact (A) and original data for "HP 1

2 Stand" by age (B) of animals of the 7.1 CW version. There is an increase in C 02 due to age, which can be observed also in the original data of, for example, "HP Stand". The stand duration is increasing with age.

- 3
- 4 HP- hind paw
- 5

6 Two other components showed a correlation coefficient above [0.250] with age. These are C 04 Swing 7 *Phase* and C 08 *Stride Length* for the XT version only $(r_{(197)} = 0.497 \text{ and } r_{(197)} = -0.255$, respectively). These 8 correlations are not very strong. As we saw sex effects in C 04 Swing Phase and C 08 Stride Length in the 9 analysis of all animals in the XT version we took a closer look at the influence of sex in these components 10 with this reduced animal number. We did a regression analysis for sex and age and found that 29% of the variation in C 04 can be explained by this model ($F_{(2.195)}$ =41.217, p<0.001, adjusted R²= 0.290). Here 11 12 both sex (β =0.220, p<0.001) and age (β =0.493, p<0.001) were significant predictors, but the beta values 13 suggest a stronger influence of age than sex. The regression analysis for C 08 Stride Length revealed an influence of both age (β =-0.258, p<0.001) and sex (β =0.172, p=0.013), but describing thereby only 9% of 14 15 the variation ($F_{(2.195)}$ =10.16, p<0.001; adjusted R²= 0.085). Again the beta values suggest a stronger

- 16 influence of age than of sex on C 08 Stride Length.
- 17

18 Taken all together we can see rather small effects of sex, body weight and age on the components. We 19 see sex and body weight effects in C 01 Print Dimensions in both CW versions. In C 08 Stride Length, the 20 second spatial component, we see an influence of age and sex only in the XT version. For C 02 Paw 21 Contact we find an age effect in both versions. In the XT version there is also an influence of body weight 22 on this component. For the other temporal dimensions there are different patterns in the two versions: 23 we see a sex effect in both C 04 Swing Phase and C 07 Turning Point in the XT version, together with an 24 age effect in C 04 Swing Phase. In the 7.1 version we only see a body weight effect for C 07 Turning 25 Point. In coordination both versions detect sex effects in C 05 Coordination of diagonal and ipsilateral 26 pairs. In C 10 Variation FP girdle pairs and Print Position we also see an influence of sex in both versions. 27 All other components related to coordination or variation do not show any effects of sex, age or body 28 weight.

29

30 Strong influence of genetic background on the gait profile

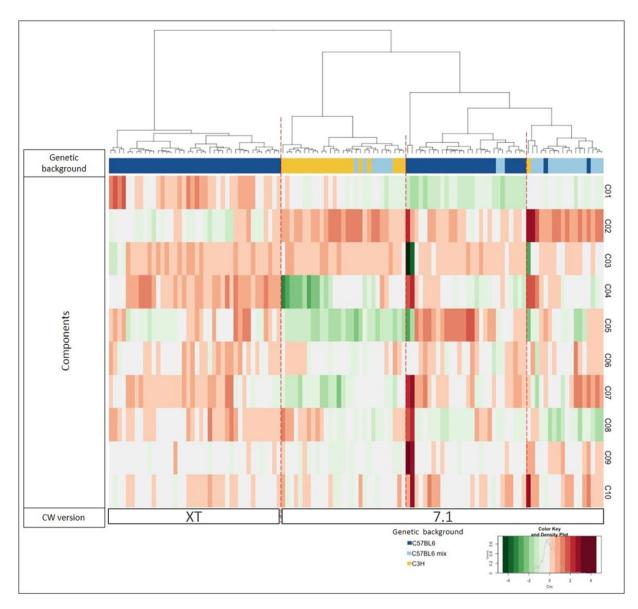
31 Using hierarchical clustering similarities in gait profile between groups of mice were investigated. We 32 conducted the hierarchical clustering by Wards method using the medians of the individual component 33 scores per mouse line, genotype and sex (one column in Fig. 5 refers to one group, for example: male 34 wild types of a specific mouse line). The dendrogram in Fig. 5 is used to represent the distance between 35 groups and clusters by its branch lengths. The shorter the branch length, the shorter the distance and 36 the greater the similarity between two pairs (e.g. groups or clusters). The heat map depicts the color-37 coded medians of the components for the single groups. The first separation in the dendrogram was 38 defined by the CW version (see dendrogram in Fig. 5). The heat map suggests that C 01 Print 39 Dimensions, C 02 Paw Contact and C 04 Swing Phase play a crucial role in separating the CW versions. 40 Consecutive separations occur on the basis of genetic background. First a group consisting essentially of

lines on a C3H background separates and then a partitioning between two C57BL6 backgrounds occurs 1 2 for animals tested on the 7.1 version. Here one group, the "C57BL6-mix", was less often backcrossed to a C57BL6 background compared to the "C57BL6" group. The components C 04 Swing Phase, C 05 3 4 Coordination of the diagonal and ipsilateral pairs and C07 Turning Point seem to influence the division 5 between background strains. In Fig. 6 graphs for selected parameters of the original data are presented. For component C 04 the parameters "FP Swing Speed" and "BOS HP", for C 05 "BOS FP" and "PhD RF-6 7 RH" (Phase Dispersion RF->RH) and for C 07 "HP Max Contact At" are presented. The C3H background 8 differs from the C57BL6 background in that they have smaller HP BOS but a larger FP BOS, a slightly higher swing speed and a different timing in PhD RF-RH. Also they have an earlier Maximum Contact. 9 10 The difference between the C57BL6 and the C57BL6 mix becomes obvious in the BOS HP and FP (higher values of the C57BL6 mix compared to the C57BL6 in both parameters) as well as the FP swing phase (a 11 reduced swing speed) ("Run duration": sex-background effect: $F_{(2,334)}=7.364$, p=0.001; Males: 12 13 F_(2,171)=9.655, p<0.001; post hoc Bonferroni, mean difference (MD): C57BL6 vs C57BL6 mix: -0.8695, 14 p<0.001; C57BL6 vs C3H: ns; C57BL6 mix vs C3H: 0.7187, p=0.011; Females: ns; "BOS FP": background effect: F_(2,334)=146.354, p<0.001; post hoc Bonferroni: MD: C57BL6 vs C57BL6 mix: -2.2612, p<0.001; 15 16 C57BL6 vs C3H: -3.7361, p<0.001; C57BL6 mix vs C3H: -1.4749, p<0.001; "BOS HP": sex-background 17 effect: F_(2,334)=14.860, p<0.001; Males: F_(2,171)=37.207, p<0.001; post hoc Bonferroni: MD: C57BL6 vs C57BL6 mix: -1.3186, p=0.01; C57BL6 vs C3H: 3.2298, p<0.001; C57BL6 mix vs C3H: 4.5484, p<0.001; 18 19 Females: F_(2.162)=64.546, p<0.001; post hoc Bonferroni: MD: C57BL6 vs C57BL6 mix: -5.0638, p<0.001, 20 C57BL6 vs C3H: 3.4776, p<0.001; C57BL6 mix vs C3H: 8.5414, p<0.001; "FP Swing Speed": background effect: F_(2.334)=26.425, p<0.001; post hoc Bonferroni: MD: C57BL6 vs C57BL6 mix: 0.1334, p<0.001; 21 22 C57BL6 vs C3H: -0.1010, p=0.002; C57BL6 mix vs C3H: -0.2343, p<0.001; "PhD RF-RH": background 23 effect: F(2 334)=81.338, p<0.001; post hoc Bonferroni: MD: C57BL6 vs C57BL6 mix: ns; C57BL6 vs C3H: -24 10.774, p<0.001; C57BL6 mix vs C3H: -9.843, p<0.001; "HP Max Contact At": sex-background interaction: F_(2,334)=9.087, p<0.001; Males : F_(2,171)=76.711, p<0.001; post hoc Bonferroni, MD: C57BL6 vs C57BL6 mix: 25 26 -8.1375, p<0.001, C57BL6 vs C3H: 6.3515, p<0.001; C57BL6 mix vs C3H: 14.4890, p<0.001; Females: 27 F_(2,162)=7.075, p=0.001; post hoc Bonferroni: MD: C57BL6 vs C57BL6 mix: ns; C57BL6 vs C3H: 3.9581, 28 p=0.017; C57BL6 mix vs C3H: 6.3927, p=0.001)

29

When taking a closer look at the distribution of the mouse lines it becomes clear that there is no strong separation between sexes or genotypes (mutant vs. wild type) within one mouse line. Generally the genotypes of one mouse line cluster together, thus emphasizing again the strong influence of the genetic background on gait parameters.

34

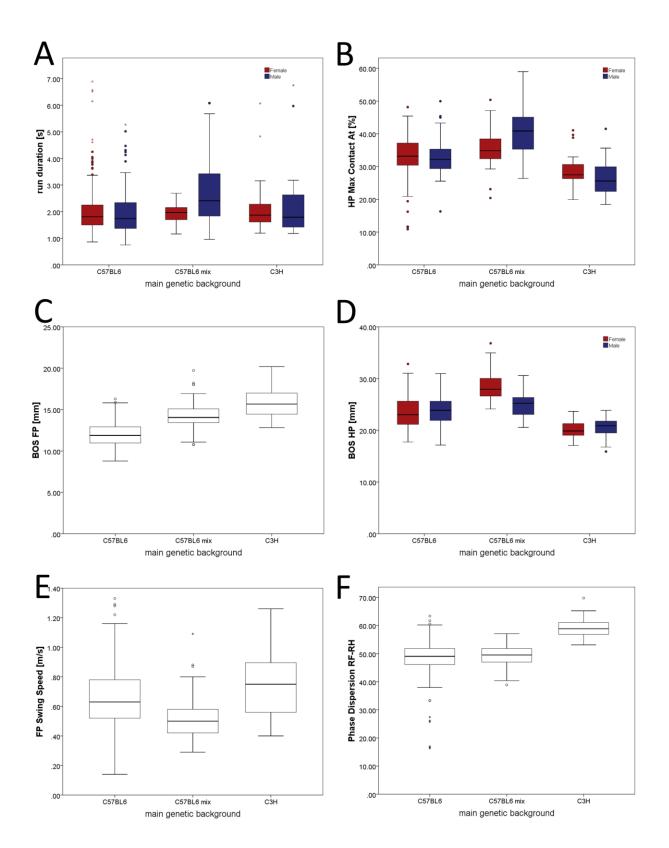




2 Fig. 5: Dendrogram and heat map from the hierarchical cluster analysis of groups

3 One vertical line represents one group, for example male wildtypes of a specific mouse line. The 4 horizontal lines represent the color-coded medians per group of the different components (see right hand 5 side; color-coding presented in the key underneath the plot). The first separation in the dendrogram 6 clusters the two different versions. The left branch includes all groups analyzed on the CW XT version and 7 the right branch includes exclusively those tested on the 7.1 version (see bottom of graph). The second 8 division is mainly defined by the background strain (compare bar on the top). With a few exceptions all 9 groups with a C3H background (yellow) cluster in the left branch, whereas most of the groups with a 10 C57BL6 (dark blue) or C57BL6 mix (light blue) background are on the right branch within the 7.1 cluster. 11 Groups of a C57BL6 and C57BL6 mix then diverge with the next branching.

12



2 genetic background in the 7.1 version. In the run duration (A) we find a sex-background interaction; in 3 the males the C57BL6 mix is different to the two other groups; in females there is no difference between 4 groups. (B) In HP Max Con At (parameter of C 07) we detect a sex-background effect; all male groups 5 differ from each other. The female C3H are different from the other two groups. (C) BOS FP (a parameter 6 of C 05) a background effect is observed- all groups differ. (D) In the BOS HP (a parameter of C 04) there 7 is a sex-background interaction: in males and females all groups differ respectively. (E) FP Swing Speed (a 8 parameter of C 04) there is a background effect detectable. Here the C57BL6 mix group has a lower 9 swing speed compared to the other two groups. (F) In PhD RF-RH (a parameter of C 05) a background 10 effect is detectable. The C3H group differs from the other two groups. HP- hind paws, FP- front paws, 11 BOS- base of support, PhD- Phase Dispersion, RF- right front paw, RH- right hind paw

Fig. 6: Original data from parameters representative for the components that seem to differentiate the

12

1

13 No redundancy in CW, OF and VP measures

We wanted to know if different tests for locomotor phenotypes indeed assess different aspects of locomotion or if there are redundancies between them. To this end we analyzed two different data sets with PCA: One set included data from CW, OF and VP test and the other set consisted of data from the IMPC including data from OF, SHIRPA, RR, GS and HC.

The PCA on CW, OF and VP data included historical data from our lab. A total of 1057 cases were analyzed including 147 parameters of the three tests (113 from the CW, 31 from the OF and 3 from the VP test; see also Table S2). The sampling adequacy of this parameter set was confirmed by the KMO measure of 0.831 and the significant Bartlett's test (Chi²= 512264.116, df= 10731, p<0.001). Extraction of 13 components explaining 80% of the total variance was done using the oblique rotation. The loadings of parameters are depicted in the structure matrix table (see Table S5). The components were characterized according to the highest loadings of the individual parameters (see Table 2).

25

			Component name	Explain variance	
On an Field		C 1	OF_Activity	19.0 %	23%
Open Fie	iu	C 7	OF_Emotionality	4.0 %	25%
	Spatial Dimension	C 2	CW_Print Dimensions	17.0 %	17%
	Temporal Dimension	C 3	CW_Swing Phase	11.0 %	
		C 5	CW_Stand and Stride Length	5.3 %	19%
CatWalk		C 8	CW_Turning Point FPs	2.6 %	
CatWalk	Interlimb Coordination	C 4	CW_Coordination of diagonal and ipsilateral pairs	7.3 %	
		C 6	CW_Coordination of HP girdle pairs	4.2 %	13%
		C 11	CW_Coordination of FP girdle pairs	1.8 %	
	Variation	C 9	CW_Variation of diagonal and ipsilateral incl LH	2.4 %	
	variation	C 10	CW_Variation incl RH	2.3 %	6%

26 Table 2: PCA for CW, OF and VP: Component Characterization

	C 13	CW_Variation of FP girdle pairs and Turning Point HPs	1.5 %	
Vertical Pole	C 12	Pole	1.6 %	2%

1

OF- Open Field; CW- CatWalk; FPs- front paws; HPs- hind paws; LH- left hind paw

2 The parameters clearly cluster according to their respective tests. Component C 12 contains all 3 parameters measured by the VP test; two components contain OF parameters (C 1 and 7) and the rest 4 of the components contain the CW parameters. For the CW parameters it is noticeable that the 5 components have a similar make up of parameters as those in the first PCA and can be allocated to the 6 four dimensions, i.e. spatial and temporal dimensions as well as interlimb coordination and its variation. 7 In the component correlation matrix (see Table S6) it becomes clear that a few components correlate, 8 although not very strongly. The highest correlation is between the two OF components (r=-0.477). As 9 already suggested by the analysis there is no strong correlation between parameters of different tests. 10 From this analysis it becomes very clear that the three tests are not redundant but measure different 11 aspects of locomotion.

12

13 IMPC data set shows that OF and HC measure two dissociable aspects of locomotor activity

Also in the IMPC data set the parameters are organized according to their respective tests, suggesting no redundancies between tests. The PCA was done including 1327 animals and 20 parameters comprising of measures from the OF, GS, SHIRPA, RR and HC (Table S7). The sampling adequacy was confirmed by KMO of 0.706 and a significant Bartlett's test (Chi²= 64427.87, df= 190, p<0.001). We extracted 6 components explaining 84.1% of the total variance using the oblique rotation (Table 3; for the structure matrix see Table S8).

20

	Component name	Major parameters	Explained variance [%]
C 1	OF Locomotion	Distance and Speed	31.3 %
C 2	OF Emotionality	Time spent in center, Distance travelled in center, Resting time in center, Number of entries in center	18.2 %
C 3	OF Exploration	Resting time in arena and periphery, Number of rearings	11.0 %
C 4	Home cage activity	Distance, Rearings and Speed	9.1 %
C 5	Grip Strength	Grip strength of FP and all paws	8.0 %
C 6	Rotarod	Mean latency to fall, Number of passive rotations	6.5 %

21 Table 3: PCA for IMPC data: Component Characterization

22 23 FP- front paws, OF- Open Field

After characterizing the components according to their main parameters we again see an allocation of parameters related to their respective test. An exception is the activity parameter from SHIRPA, which has its major loading in C 5 *Grip Strength* (0.386), which is not very strong, and its second and third highest loadings into C 4 *Home cage activity* (0.369) and C 1 *OF Locomotion* (0.364), respectively,

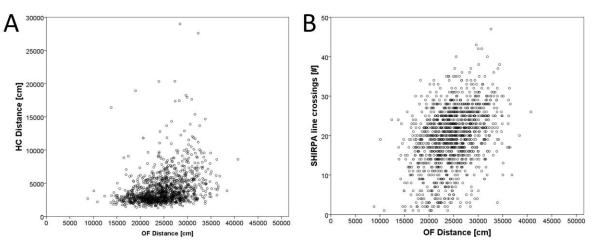
suggesting that all components do not explain a high amount of the parameter's variation. Interestingly

both C 1 and C 4 contain the activity measures from OF and the HC respectively and there is a mild

Page | 19

1 component correlation between them (r=0.327) (see Table S9). Taking a closer look at the original data 2 of all the horizontal activity parameters from OF, HC and SHIRPA show that there is only a mild 3 correlation between them (OF Distance- HC Distance: $r_{(1326)}$ = 0.326, OF Distance- SHIRPA: $r_{(1326)}$ =0.370, 4 HC Distance- SHIRPA: $r_{(1326)}$ =0.240, see also Fig. 7). This suggests that dissociable aspects of activity are 5 measured by the different tests, again strengthening the interpretation of no redundancy between 6 tests.

- 7
- 8



9 OF Distance [cm]
 10 Fig. 7: Scatter plots for correlations between measures of activity. Graph A depicts "OF Distance" and
 11 "HC Distance" and graph B "OF Distance" and "SHIRPA Line crossings". There are only weak correlations
 12 between the parameters. OF- Open Field, HC- Home Cage

13

14 **Discussion**

15 In our study we performed PCAs on different data sets related to locomotor phenotypes to refine future data analysis and to investigate possible redundancies between different tests. Knowledge of the latter 16 17 better informs future decisions on phenotyping strategies: if tests were redundant, one could focus on 18 the simplest one. Our results suggest that for future CW analysis a primary focus can be on FP and HP 19 measures, and not all individual paws, as well as one of the interlimb coordination measures (e.g. only 20 Phase Dispersions). Also we detected minor effects of sex, body weight and age on the components, but 21 a large effect of genetic background and CW version. The analysis of different tests revealed no 22 significant redundancies between them, thereby emphasizing their originality.

A PCA is a useful tool to reduce dimensions of large data sets and consequently analyze the components. This approach has already been successfully used in several papers to identify linear combinations of variables in a high-dimension space best representing the variance that is present in the data [7,19,20]. Nonetheless it is clear that in some publications important specifications for corroborative purposes are missing. This is mainly missing information on sampling adequacy, such as the KMO and Bartlett's test, and/or details on data processing and component extraction methods. All these are important for the reader to judge the validity of the results [5].

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Locomotor phenotypes are composed of several different dimensions including gait performance. Gait 1 2 performance can be analyzed by the CW system measuring over 200 parameters. We used the PCA to 3 examine the relative contribution of 112 parameters to the variance in the data obtained from CW 4 measurements. Our results show that the large amount of parameters measured by the CW system can 5 be compacted via PCA into 10 components explaining 80% of the total variance. These 10 components 6 nicely describe several aspects of gait performance, including spatial (C 01 Print Dimensions and 08 7 Stride Length) and temporal (CO2 Paw Contact, 04 Swing Phase and CO7 Turning Point) dimensions as 8 well as interlimb coordination (C 05 and 06) and its variation (C 03, 09 and 10). Interestingly the spatial 9 and temporal dimensions as well as coordination including its variation, all contribute equally to the 10 explained variance in the data. With respect to future studies, the data suggests that focusing on the 11 parameters of FP and HP is sufficient (instead of considering parameters of each paw individually), as these always cluster together with their respective single paws in one component. Obviously this is only 12 13 applicable if the experimental design does not dictate a different approach, for example for lateral 14 lesion models, where the sides/single paws have to be compared. Also for the coordination parameters, 15 i.e. couplings and phase dispersions, it is adequate to decide on one. When analyzing individual mouse 16 lines for genotype effects separately it becomes apparent that the FP and HP reliably reflect effects in 17 single paws, if these are not too small. This applies also for phase dispersion representing in essence the 18 couplings. BOS FP and HP should be analyzed individually, as they have quite opposing patterns of 19 loadings onto the different components. This is reflected in their correlation with other parameters. The 20 heavier the mouse the larger the BOS HP, but this effect is far less strong in the BOS FP. When looking at 21 other correlations within the original data and BOS, we have to check the two CW versions separately 22 (see below). Generally, when the BOS gets wider the run duration increases. Furthermore, the stance 23 phase gets longer and the stride length shorter all culminating in a slowing in movement. In the 7.1 24 version the widening of the BOS FP is correlated with changes in the interlimb coordination, the step 25 pattern used and a reduction in support of diagonal paw pairs. In the XT version we also see a reduction 26 in diagonal support but with a wider BOS HP. Here a positive correlation with "Print position" indicates 27 that the hind paw is placed behind the front paw, again suggesting a slowing of movement. For the BOS 28 FP in the XT version we see a different picture; there is no correlation with run duration or stride length. 29 We do see an increased stance phase, indicated by an increased duty cycle, and a change in lateral 30 support. Also the turning point in the FP is delayed.

31 Further analysis of the extracted components indicated that the two CW versions show clear 32 differences, as we see in 9 out of 10 components significant version effects. When looking at the 33 parameters we see strong differences between the two versions appearing in spatial and temporal 34 dimensions. As there are less strong or no differences at all in the coordination and variation 35 components between the two versions this again suggests a discrepancy in the accuracy of measuring 36 dimensions between the versions. Possibly the precision of the system has changed and as such 37 dimensions differ. In the XT version a high-speed camera is recording the images at a much higher rate 38 and a red background illumination has been integrated. Thus it is possible that the paw print can be 39 delineated in more accuracy. This is also in accordance with Chen at al [21], who found that the XT 40 version can detect more subtle alterations, which cannot be detected by the earlier version in a model 41 of sciatic nerve injury. Theoretically another issue in our study could be the heterogeneity of animals 42 measured by the two versions. The age range, for example, of the 7.1 version is between 3 to 28 months and for the XT version it is between 3 and 15 months of age (see Fig. S2). Another factor differing 43 44 between the two versions is the variability in genetic background. The genetic background for the

1 animals tested on the 7.1 version is broader (including C57BL6, C3H, and mixed backgrounds), for the XT 2 version the animals were of a more similar genetic background. Both factors - age and genetic 3 background- do have an influence on CW performance and this has to be taken into account. It stresses 4 the importance of comparing animals with their corresponding littermates tested at the same time and 5 same test version. To take a closer look at these possible confounding factors in our study we also 6 compared the two versions by including only animals on a C57BL6 (mix) background and only up to an 7 age of 15 months, so that a vast amount of heterogeneity is circumvented. Nonetheless we still observe 8 significant differences between the versions thus confirming a profound difference between the CW 9 versions. Therefore we decided to stay with the approach of including all animals. Also as this has the 10 benefit to analyze effects of age, body weight and genetic background strains with decent animal 11 numbers.

12 We continued our analysis by checking the two versions separately and looking for the influence of sex, 13 body weight and age on the components. We could see that body weight has an influence on C 01 Print 14 Dimensions in both CW versions. This is not too surprising as parameters of print dimension, i.e. print 15 length, width and area, all load highly onto this component and are expected to change with body 16 weight [22]. Interestingly, we see that in females the changes due to body weight seem to be stronger 17 than in males, where especially in the 7.1 version males show no correlation between C 01 and body weight. Body weight also has an influence on C 07 Turning Point in the 7.1 version. Here we see an 18 19 increase in all parameters of "Maximum contact at %", indicating a delayed turning point within the 20 stand phase with increasing body weight.

21 Age has an influence on C 02 Paw Contact in both versions, as well as C 04 Swing Phase and C 08 Stride Length in the XT version. For C 02 Paw Contact we find also a body weight influence in the XT. Generally, 22 23 in C 02, parameters such as the duty cycle and stand duration increase with age suggesting a slowing of 24 movement, which is also supported by the increase of run duration and a reduction in diagonal support. 25 An age-related increase in swing duration and a reduction in swing speed can be observed, evident 26 mainly in the XT version, where also the corresponding component, C 04 Swing Phase, shows a 27 significant age effect. In the XT version C 08 Stride Length is influenced by age as well. We see a negative 28 correlation between the parameters of "Stride length" and age, again evidence of a slowing of 29 movement with age. This slowing of movement we see in mice is also true in humans [23,24]. Summing 30 up, both body weight and age do have an effect on the components, but it is rather small.

31 Two components showing sex dependent effects are C 10 Variation excl RH for girdle pairs and C 05 32 Coordination of the diagonal and ipsilateral pairs for both versions. In C 10 Variation FP girdle pairs and 33 Print Position the parameter of "Print position", which measures the placement of the HP with respect 34 to the FP, is larger in males than in females. This is partly due to body weight differences. There is no 35 obvious sex difference in the parameters of variation in this component. For C 05 Coordination of the 36 diagonal and ipsilateral pairs we see opposing effects in the two CW versions: in the 7.1 version females 37 have a higher score than males and in the XT version it is the opposite. The parameter "Step sequence 38 AB"- the most frequently used step pattern by all animals- has a higher percentage in the males of the 39 7.1 version compared to the respective females and a higher value for the females of the XT version 40 compared to the respective males. The couplings and phase dispersions of C 05 show the opposite in the 41 two versions between the sexes.

Clustering the medians of the individual component scores of the different mouse lines according to 1 2 their line, genotype and sex showed that the CW version defines the first separation in the dendrogram, 3 again depicting discrepancies between the two. As mouse lines, independent of their mutation and 4 according to their genetic background, cluster together, a strong influence of the genetic background 5 (and not the mutation) is depicted. Seemingly the genetic background has a profound influence on gait 6 parameters, which can be detected by the CW system. The heat map suggests that C 04 Swing Phase, C 7 05 Coordination of the diagonal and ipsilateral pairs and C 07 Turning Point, contribute strongly to the 8 separation between a C3H and a C57BL6 background. This can be illustrated by several parameters: 9 although all animals need approximately the same time to cross the runway, we see reduced swing 10 speed in the C57BL6 mix animals and a slightly higher swing speed in the C3H animals compared to 11 C57BL6. Also the inter-paw relations differ between the background strains. The same holds true for 12 BOS; in both the FP and HP C57BL6 mix animals have a broader BOS compared to C57BL6, whereas the 13 C3H have a broader FP BOS but narrower HP BOS compared to C57BL6. For the parameter HP Max Con 14 At [%] C3H animals show an earlier Max contact. It is not too surprising that also here different strains 15 show different patterns. Strain differences have been shown for several tests and this stresses the 16 importance of comparing littermates and not similar genetic backgrounds with each other, when looking 17 for effects of a mutation [25,26].

Still the CW is a valuable tool for investigating genotype specific (in terms of a mutation) gait phenotypes. In the 25 lines we included in our study 21 of them showed a genotype specific phenotype. These numbers seem very high and have to be considered in respect to the selection of lines for CW testing; we subjected lines to the CW for which we hypothesized a gait phenotype, due to the mutation or results from other locomotor tests.

23 Next we addressed the question of redundancies between tests. We asked how parameters of several 24 different tests would group into principal components. We did this with a subset of the animals used 25 above for the CW analysis, which all performed the CW, OF and VP test. We extracted 13 components 26 explaining 80% of the total variance. The parameters from different tests group into different 27 components, so that each component can be assigned to one test. This indicates that the tests do not 28 predict the outcome of a different one and are therefore useful expansions and complementations of 29 each other and not redundant. This is also suggested by a totally different data set. Here data from IMPC 30 animals of the German Mouse Clinic was taken from different tests measuring different aspects of 31 locomotor performance. This included parameters of the OF, GS, RR, HC activity as well as activity 32 measured by SHIRPA. Again we see a separation of the parameters to their tests. An exception is the 33 activity measure of the SHIRPA protocol which is attributed to C 5 Grip Strength. Nonetheless it should 34 be mentioned that the loading onto this component is not very high (0.386) and it loads onto C 4 Home 35 cage activity and C 1 OF Locomotion with a similar loading (0.369 and 0.364, respectively), suggesting 36 that none of the components describe a high variation of this parameter. The allocation of parameters 37 of a single test to one component stresses the validity of each single test. This is in accordance with the 38 specificity the tests were designed for, for example GS for measuring muscle force, RR for motor 39 coordination and balance and OF for spontaneous locomotion as well as emotional phenotypes. 40 Generally one would not expect that these highly specialized tests can replace another, but it is possible 41 that some of the measured parameters can correlate with parameters of the other test. For example 42 one would not expect that HC observations can replace OF, but it would be possible that the measured 43 activity in HC would hint towards effects on activity parameters of the OF, or vice versa. Two

components, C 1 and C 4, contain the activity measures from OF and the HC, respectively, which 1 2 moderately correlate. When going back to the original data we can see that the correlation between the 3 activity measures of the different tests are not that impressive ("OF Distance" and "SHIRPA Activity" 4 r=0.370; "OF Distance" with "HC Distance" r=0.326; "SHIRPA Activity" with "HC Distance" r=0.240; "OF 5 ArenaAverageSpeed" with "HC Speed_mean" r=0.336; "OF Rearings" with "HC Rearings" r=0.123), 6 suggesting that, to some extent, there is a small relationship between activity measures between tests 7 but they do not really have predictive value. This can be explained by the different designs for 8 measuring activity. In the SHIRPA protocol used here the activity is measured in the first 30 seconds 9 after placing the animal into a new environment. It is measured by counting line crossings. In the OF and 10 HC, activity is measured via infra-red beam breaks. In OF the activity is recorded for 20 minutes whereas 11 in the HC activity is measured for at least 21 h. In both SHIRPA and OF, activity is measured in a new unfamiliar arena, whereas in the HC the activity is measured in a familiar environment. Thus different 12 13 aspects come into play, such as stress-reactions and anxiety-related behaviors in a new environment. It 14 is known that animals can differ in these distinct tests and there are underlying genetic factors 15 contributing to these discrepancies as shown by studies with collaborative crosses [27-30]. Again this 16 strengthens the value of each single test applied, as they do measure different aspects of locomotor 17 performance and thus are not redundant.

18 In conclusion, we have shown that a large amount of data points generated by the CW system can be 19 meaningfully compacted via PCA to a few components. Here we extracted 10 components explaining 20 80% of the total variance in the data. The loadings of parameters onto components suggest the 21 possibility to focus on FP and HP parameters neglecting the parameters for every single paw, as well as 22 analyzing only one measure of coordination, i.e. couplings or phase dispersion, as long as the 23 experimental question does not require specific analysis. It is necessary to look at the BOS FP and HP 24 separately, as they seem to be independent of each other. The version of the CW system seems to have 25 a great influence on the data, especially in the paw dimensions. Sex, body weight, age and genetic 26 background effects influence the different components, suggesting to carefully consider these in 27 experimental planning- especially when comparing data over time. We could show that the use of a PCA 28 can reduce a big data set to a few meaningful components, which can easily be used to unsheathe 29 underlying patterns. The PCAs, which included different tests, reveal the validity of each single test, as 30 they measure different dimensions of locomotor performance [31].

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Supporting Information

2

- 3 Table S1: Number of cases used for PCA
- 4 Note that the cases from the CW OF VP Analysis are part of the CW Analysis cases. CW- CatWalk; OF-
- 5 Open Field; VP- Vertical Pole

Number of cases	CV	V Analysi	S	cw o	lysis	IMPC Analysis	
	Total	7.1	ХТ	Total	7.1	ХТ	Total
Total cases	1499	627	872	1057	358	399	1327
female cases	691	283	408	482	153	329	618
male cases	808	344	464	575	205	370	709
Wild type cases	660	276	384	436	157	279	665
female cases	316	136	180	197	71	126	312
male cases	344	140	204	239	86	153	353
Mutant cases	838	350	488	620	200	420	662
female cases	375	147	228	285	82	203	306
male cases	463	203	260	335	118	217	356
			1		1		
Age range at CW (for IMPC at OF) testing (months)	3-26	3-26	3-15	3-18	4-18	3-12	2

1 Table S2: Parameters from CatWalk, Open Field and Vertical Pole

	CatWalk P	arameters	Open Field Parameters	Vertical Pole Parameter
Print Length		Run_Duration	OF_Dist05	Pole_TimeTurn
Print Width		StepSequence_CA	OF_dist10	Pole_TimeComplete
Print Area		StepSequence_CB	OF_dist15	Pole_TimeDown
lax Contact Area		StepSequence_AA	OF_dist20	
tand duration	for RF, RH, LF,	StepSequence_AB	OF_distTot	
wingduration	LH, FP and HP	StepSequence_RA	OF_Rear05]
win Speed		StepSequence_RB	OF_Rear10	
tride Length		StepSequence_RegularityIndex	OF_Rear15	
uty Cycle		Support_Zero	OF_Rear20]
lax Conatct At		Support_Single	OF_RearTot	
OS_FP		Support_Diagonal	OF %DistCen05	
OS_HP		Support_Girdle	OF_%DistCen10	1
rintPosition_RP		Support_Lateral	OF_%DistCen15]
rintPosition_LP			OF_%DistCen20]
hDig_RFLH_CStat		PhDig_RFLH_CStat_R	OF_%DistCenTot]
hDig_LFRH_CStat		PhDig_LFRH_CStat_R	OF_%TiCen05	1
hGird_LHRH_CStat		PhGird_LHRH_CStat_R	OF %TiCen10	1
hGird LFRF CStat		PhGird LFRF CStat R	OF %TiCen15	1
hlps_RFRH_CStat		Phips_RFRH_CStat_R	OF %TiCen20	1
hlps LFLH CStat		Phips LFLH CStat R	OF %TiCenTot	1
oupDig RFLH CStat		CoupDig RFLH CStat R	OF WholeRest	1
oupDig_LFRH_CStat		CoupDig_LFRH_CStat_R	OF WholeAvSpeed	1
oupDig_LHRF_CStat		CoupDig LHRF_CStat_R	OF PerDist	1
oupDig RHLF CStat		CoupDig RHLF CStat R	OF PerRest	1
oupGir LHRH CStat		CoupGir LHRH CStat R	OF PerPerm	1
oupGir LFRF CStat		CoupGir_LFRF_CStat_R	OF PerAvSpeed	1
oupGir RHLH CStat		CoupGir RHLH CStat R	OF CenDist	1
oupGir RFLF CStat		CoupGir RFLF CStat R	OF CenRest	1
ouplpsi RFRH CStat		Couplpsi RFRH CStat R	OF CenPerm	1
ouplpsi LFLH CStat		Couplpsi LFLH CStat R	OF CenAvSpeed	1
ouplpsi_RHRF_CStat		Couplpsi RHRF CStat R	OF NbEntriesCen	1
oupIpsi_LHLF_CStat		Couplpsi_LHLF_CStat_R		-
F - rightfront	FP - front paws	CA - Cruciate (RF-LF-RH-LH)	dist-distance	TimeTurn-time to turn
H - right hind	HP - hind paws	CB - Cruciate (LF-RF-LH-RH)	rear-rearings	TimeComplete- total time
- left front		AA -Alternate (RF-RH-LF-LH)	%DistCen- Distance travelled in the centre (%)	TimeDown-time needed from tur
H - left hind	RP - right paws	AB - Alternate (LF-RH-RF-RH)	%TiCen-Time in Centre (%)	to fully descend
	LP - left paws	RA - Rotate (RF-LF-LH-RH)	WholeRest- Resting time in whole arena	
OS - base of support		RB - Rotate (LF-RF-RH-LH)	WholeAvSpeed- average speed in whole arena	
			PerDist- Distance Travelled in Periphery	
h - Phase dispersion		Cstat - Circular Statistics	PerRest- Resting time in Periphery	
oup - Couplings		CStat R-Strength of directedness and	PerPerm-Permanence time in Periphery	
ig - diagonal pairs		therefore a measure of Variation	PerAvSpeed- Average Speed in periphery	
osi - ipsilateral pairs			CenDist-Distance Travelled in Center	
ir - girdle pairs			CenRest- Resting time in Center	
			CenPerm-Permanence time in Center	
			CenAvSpeed- Average Speed in Center	
			NbEntriesCen-Number of Entries into Center	

1 Table S3: Structure matrix for CatWalk PCA

2 For abbreviations used see Table S2

Structure matrix

	Component										
	1	2	3	4	5	6	7	8	9	10	
HP_PrintArea	.965	.005	006	.008	.170	.031	244	.083	011	306	
FP_MaxContactArea	.963	164	017	.094	.106	.053	048	.253	015	290	
FP_PrintArea	.960	179	019	.141	.086	.056	037	.296	020	283	
HP_MaxContactArea	.958	061	004	.008	.189	.039	233	.104	017	305	
RF_PrintArea	.958	167	013	.135	.089	.036	044	.294	017	284	
RF_MaxContactArea	.957	143	016	.083	.104	.024	057	.246	010	293	
LF_MaxContactArea	.954	182	018	.105	.108	.080	038	.256	018	284	
LF_PrintArea	.952	189	025	.147	.083	.076	028	.294	022	280	
LH_PrintArea	.950	040	050	.023	.162	.095	214	.091	049	290	
LH_MaxContactArea	.941	095	047	.016	.181	.084	207	.108	053	288	
RH_MaxContactArea	.931	019	.036	.000	.190	014	251	.094	.025	312	
RH_PrintArea	.931	.056	.033	010	.171	043	266	.070	.033	312	
FP_PrintLength	.899	115	.012	.149	.027	.065	001	.438	080	270	
RF_PrintLength	.879	110	.020	.145	.036	.041	012	.429	071	259	
HP_PrintLength	.879	.247	.038	049	.066	.001	155	.040	046	303	
LF_PrintLength	.878	116	.003	.148	.018	.088	.013	.427	085	270	
FP_PrintWidth	.874	.102	.048	021	.015	.044	082	.365	105	324	
LH_PrintLength	.858	.197	014	014	.051	.085	135	.057	085	277	
LF_PrintWidth	.845	.056	.047	.002	002	.104	056	.367	127	307	
RF_PrintWidth	.841	.139	.044	041	.032	018	099	.337	074	321	
RH_PrintLength	.814	.270	.076	079	.075	080	160	.022	.000	303	
LH_PrintWidth	.787	.329	003	116	.084	.053	290	072	107	267	
HP_PrintWidth	.765	.432	.058	169	.078	.002	335	098	048	280	
RH_PrintWidth	.653	.489	.103	206	.064	053	345	112	.023	266	
HP_Stand	049	.894	188	.287	304	061	.038	308	.229	.127	
RH_Stand	062	.891	160	.261	294	112	.027	299	.256	.112	
HP_DutyCycle	124	.888	.009	312	141	063	050	476	.007	114	
LH_Stand	034	.874	217	.306	305	008	.047	302	.198	.137	
RH_DutyCycle	142	.854	.014	320	121	193	069	444	.044	108	
LH_DutyCycle	090	.851	004	277	151	.088	024	471	034	116	
FP_DutyCycle	.012	.815	.021	113	595	009	.103	192	042	137	
RF_DutyCycle	.024	.789	.033	105	575	098	.097	185	020	141	
FP_Stand	.008	.789	242	.428	418	033	.090	148	.263	.102	
LF_DutyCycle	002	.784	.008	112	573	.081	.103	187	060	126	
LF_Stand	.009	.783	244	.428	417	009	.095	153	.250	.098	
RF_Stand	.012	.781	236	.428	409	056	.085	141	.271	.099	
Support_Diagonal	.309	751	.208	.110	.438	.095	.027	.448	245	219	
Run_Duration	195	.613	254	.322	265	037	.089	363	.259	.206	
Support_Single	064	527	008	.303	.060	.072	.214	.416	040	.071	
Support_Zero	160	358	022	.186	035	.045	.064	.250	045	.073	
Support_Girdle	112	340	037	.208	057	.025	.212	.201	.066	.085	

Couplpsi_RFRH_CStat_R	.000	052	.945	079	048	.050	.092	.145	502	188
Phlps_RFRH_CStat_R	.045	058	.932	057	047	.056	.098	.172	501	196
PhDig_LFRH_CStat_R	040	.034	.918	053	165	.034	.071	.105	432	096
CoupDig_LFRH_CStat_R	035	.059	.918	049	135	.036	.079	.109	444	130
Couplpsi_RHRF_CStat_R	003	070	.915	017	037	.045	.049	.176	437	173
PhGird_LHRH_CStat_R	053	021	.863	035	095	.057	.054	.030	712	001
CoupGir_RHLH_CStat_R	107	021	.862	030	106	.033	.034	.021	685	.047
CoupGir_LHRH_CStat_R	086	031	.844	046	101	.039	.037	.019	734	.027
StepSequence_RegularityIndex	180	053	.615	075	118	.041	050	107	482	.054
CoupDig_RHLF_CStat_R	.191	122	.600	097	.016	.106	.009	.304	556	472
StepSequence_RA	.006	.100	451	.071	008	012	.090	.141	.174	.073
HP_SwingSpeed	.077	.073	.195	908	.151	028	237	.224	230	089
RH_SwingSpeed	.042	.103	.198	880	.142	150	241	.205	183	085
LH_SwingSpeed	.111	.034	.168	875	.151	.103	219	.230	262	092
FP_Swing	.166	.197	030	.839	.180	.058	.076	.108	.050	199
RF_Swing	.145	.196	051	.827	.172	.125	.085	.105	.044	177
HP_Swing	.199	279	.100	.817	100	.115	.185	.440	100	063
LF_Swing	.177	.184	012	.813	.167	019	.071	.097	.056	217
LH_Swing	.195	242	.056	.801	070	026	.182	.444	101	078
RH_Swing	.189	276	.114	.749	108	.213	.174	.375	077	049
FP_SwingSpeed	.242	318	.194	714	.023	.022	182	.593	276	097
LF_SwingSpeed	.231	315	.181	706	.026	.068	173	.584	280	093
RF_SwingSpeed	.246	313	.204	704	.023	023	184	.584	263	103
BOS_HP	.299	.077	102	.339	.209	.031	.259	108	.156	242
StepSequence_AB	136	.290	.074	.056	864	088	092	164	026	.083
Phlps_RFRH_CStat	038	.404	.158	.016	827	.102	.113	.121	142	066
Couplpsi RFRH CStat	038	.411	.152	.021	823	.093	.117	.126	137	046
CoupIpsi_RFRH_CStat StepSequence AA	038 .024		.152 088			.093 .025	.117 051	.126 .113	137 .028	
StepSequence_AA	.024	132	088	082	.816	.025	051	.113	.028	069
StepSequence_AA CoupIpsi_LFLH_CStat	.024 059	132 .420	088 .121	082 013	.816 798	.025 083	051 .052	.113 .100	.028 252	069 037
StepSequence_AA CoupIpsi_LFLH_CStat PhIps_LFLH_CStat	.024 059 058	132 .420 .414	088 .121 .101	082 013 013	.816 798 798	.025 083 092	051 .052 .071	.113 .100 .092	.028 252 247	069 037 049
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat	.024 059 058 .171	132 .420 .414 185	088 .121 .101 .101	082 013 013 .011	.816 798 798 .794	.025 083 092 .001	051 .052 .071 .112	.113 .100 .092 .137	.028 252 247 218	069 037 049 369
StepSequence_AA CoupIpsi_LFLH_CStat PhIps_LFLH_CStat CoupIpsi_RHRF_CStat CoupIpsi_LHLF_CStat	.024 059 058 .171 .214	132 .420 .414 185 229	088 .121 .101 .101 .173	082 013 013 .011 .007	.816 798 798 .794 .784	.025 083 092 .001 .203	051 .052 .071 .112 .148	.113 .100 .092 .137 .150	.028 252 247 218 152	069 037 049 369 418
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat	.024 059 058 .171 .214 .141	132 .420 .414 185 229 058	088 .121 .101 .101 .173 117	082 013 013 .011 .007 .013	.816 798 798 .794 .784 .708	.025 083 092 .001 .203 417	051 .052 .071 .112 .148 .033	.113 .100 .092 .137 .150 .137	.028 252 247 218 152 .015	069 037 049 369 418 175
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat	.024 059 058 .171 .214 .141 075	132 .420 .414 185 229 058 .091	088 .121 .101 .101 .173 117 .105	082 013 013 .011 .007 .013 .047	.816 798 798 .794 .784 .708 704	.025 083 092 .001 .203 417 .425	051 .052 .071 .112 .148 .033 .008	.113 .100 .092 .137 .150 .137 .137 115	.028 252 247 218 152 .015 061	069 037 049 369 418 175 .084
StepSequence_AA CoupIpsi_LFLH_CStat PhIps_LFLH_CStat CoupIpsi_RHRF_CStat CoupIpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat	.024 059 058 .171 .214 .141 075 .111	132 .420 .414 185 229 058 .091 030	088 .121 .101 .101 .173 117 .105 113	082 013 013 .011 .007 .013 .047 .011	.816 798 798 .794 .784 .708 704 .703	.025 083 092 .001 .203 417 .425 404	051 .052 .071 .112 .148 .033 .008 .027	.113 .100 .092 .137 .150 .137 115 .101	.028 252 247 218 152 .015 061 .021	069 037 049 369 418 175 .084 170
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat	.024 059 058 .171 .214 .141 075 .111 093	132 .420 .414 185 229 058 .091 030 .155	088 .121 .101 .101 .173 117 .105 113 .121	082 013 .011 .007 .013 .047 .011 .003	.816 798 798 .794 .784 .708 704 .703 671	.025 083 092 .001 .203 417 .425 404 512	051 .052 .071 .112 .148 .033 .008 .027 014	.113 .100 .092 .137 .150 .137 115 .101 050	.028 252 247 218 152 .015 061 .021 019	069 037 049 369 418 175 .084 170 047
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat CoupDig_RFLH_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138	132 .420 .414 185 229 058 .091 030 .155 121	088 .121 .101 .101 .173 117 .105 113 .121 121	082 013 .011 .007 .013 .047 .011 .003 .054	.816 798 798 .794 .784 .708 704 .703 671 .663	.025 083 092 .001 .203 417 .425 404 512 .544	051 .052 .071 .112 .148 .033 .008 .027 014 .049	.113 .100 .092 .137 .150 .137 115 .101 050 .085	.028 252 247 218 152 .015 061 .021 019 .035	069 037 049 369 418 175 .084 170 047 044
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplg_LFRH_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138 .101	132 .420 .414 185 229 058 .091 030 .155 121 100	088 .121 .101 .101 .173 117 .105 113 .121 121 104	082 013 .011 .007 .013 .047 .011 .003 .054 .054	.816 798 798 .794 .784 .708 704 .703 671 .663 .657	.025 083 092 .001 .203 417 .425 404 512 .544 .538	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062	.028 252 247 218 152 .015 061 .021 019 .035 .018	069 037 049 369 418 175 .084 170 047 044 050
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplog_LFRH_CStat CoupDig_LFRH_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat CoupDig_LHRF_CStat PhDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185	132 .420 .414 185 229 058 .091 030 .155 121 100 .420	088 .121 .101 .103 117 .105 113 .121 121 104 108	082 013 .011 .007 .013 .047 .011 .003 .054 .054 125	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469	.025 083 092 .001 .203 417 .425 404 512 .544 .538 044	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125	069 037 049 369 418 175 .084 170 047 044 050 .067
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264	082 013 .011 .007 .013 .047 .011 .003 .054 .054 125 .034	.816 798 .798 .794 .784 .708 704 .703 671 .663 .657 469 038	.025 083 092 .001 .203 417 .425 404 512 .544 .538 044 .785	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183	.113 .100 .092 .137 .150 .137 .115 .101 .085 .062 .374 .145	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200	069 037 049 369 418 175 .084 170 047 044 050 .067 329
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129 .128	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068 .048	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264 .270	082 013 .011 .007 .013 .047 .011 .003 .054 .054 125 .034 .028	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469 038 030	.025 083 092 .001 .203 417 .425 404 512 .544 .538 044 .785 .783	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183 .162	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374 .145 .143	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200 231	069 037 049 369 418 175 .084 170 047 047 044 050 .067 329 339
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat PhGird_LHRH_CStat StepSequence_CA	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129 .128 .140	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068 .048 240	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264 .270 .008	082 013 .011 .007 .013 .047 .011 .003 .054 .054 125 .034 .028 .038	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469 038 030 .191	.025 083 092 .001 .203 417 .425 404 512 .538 044 .785 .783 .750	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183 .162 .113	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374 .143 .143 .043	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200 231 037	069 037 049 369 418 175 .084 170 047 047 044 050 .067 329 339 .024
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat PhGird_LHRH_CStat StepSequence_CA StepSequence_CB	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129 .128 .140 .056	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068 .048 240 085	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264 .270 .008 .078	082 013 .011 .007 .013 .047 .011 .003 .054 .054 125 .034 .028 .038 054	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469 038 030 .191 .209	.025 083 092 .001 .203 417 .425 404 512 .544 .538 044 .785 .783 .750 709	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183 .162 .113 .091	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374 .145 .143 .043 .064	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200 231 037 008	069 037 049 418 175 .084 170 047 047 044 050 .067 329 339 .024 114
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat CoupDig_LHRF_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat StepSequence_CA StepSequence_CB CoupGir_LFRF_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129 .128 .140 .056 .083	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068 .048 240 085 .145	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264 .270 .008 .078 .156	082 013 .011 .007 .013 .047 .011 .003 .054 .054 .054 .028 .038 054 .066	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469 038 030 .191 .209 003	.025 083 092 .001 .203 417 .425 404 512 .544 .538 044 .785 .783 .783 .750 .709 .616	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183 .162 .113 .091 .212	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374 .145 .143 .043 .064 .200	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200 231 037 008 389	069 037 049 369 418 175 .084 170 047 044 050 .067 329 339 .024 114 367
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_RFLH_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat StepSequence_CA StepSequence_CB CoupGir_LFRF_CStat PhGird_LFRF_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129 .128 .140 .056 .083 .089	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068 .048 240 085 .145 .150	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264 .270 .008 .078 .156 .160	082 013 .011 .007 .013 .047 .011 .003 .054 .054 .054 .028 .038 054 .066 .062	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469 038 030 .191 .209 003 .005	.025 083 092 .001 .203 417 .425 404 512 .538 044 .785 .783 .750 .709 .616 .615	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183 .162 .113 .091 .212 .199	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374 .143 .043 .043 .064 .200 .197	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200 231 231 037 008 389 389	069 037 049 369 418 175 .084 170 047 047 044 050 .067 329 339 .024 114 367 369
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_LHRF_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat PhGird_LHRH_CStat StepSequence_CA StepSequence_CB CoupGir_LFRF_CStat PhGird_LFRF_CStat CoupGir_RHLH_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129 .128 .140 .056 .083 .089 .104	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068 .048 240 085 .145 .150 .250	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264 .270 .008 .078 .156 .160 .038	082 013 013 .011 .007 .013 .047 .011 .003 .054 .054 125 .034 .028 .038 054 .066 .062 .008	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469 038 030 .191 .209 003 .005 .041	.025 083 092 .001 .203 417 .425 404 512 .538 044 .538 044 .785 .783 .785 .783 .750 709 .616 .615 .605	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183 .162 .113 .091 .212 .199 .171	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374 .143 .043 .043 .064 .200 .197 .233	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200 231 037 008 389 391 251	069 037 049 369 418 175 .084 170 047 047 044 050 .067 329 339 .024 114 367 369 313
StepSequence_AA Couplpsi_LFLH_CStat Phlps_LFLH_CStat Couplpsi_RHRF_CStat Couplpsi_LHLF_CStat CoupDig_LFRH_CStat CoupDig_RHLF_CStat PhDig_LFRH_CStat CoupDig_RFLH_CStat CoupDig_RFLH_CStat PhDig_RFLH_CStat BOS_FP CoupGir_LHRH_CStat StepSequence_CA StepSequence_CB CoupGir_LFRF_CStat PhGird_LFRF_CStat	.024 059 058 .171 .214 .141 075 .111 093 .138 .101 185 .129 .128 .140 .056 .083 .089	132 .420 .414 185 229 058 .091 030 .155 121 100 .420 .068 .048 240 085 .145 .150	088 .121 .101 .101 .173 117 .105 113 .121 121 104 108 .264 .270 .008 .078 .156 .160	082 013 .011 .007 .013 .047 .011 .003 .054 .054 .054 .028 .038 054 .066 .062	.816 798 798 .794 .784 .708 704 .703 671 .663 .657 469 038 030 .191 .209 003 .005	.025 083 092 .001 .203 417 .425 404 512 .538 044 .785 .783 .750 .709 .616 .615	051 .052 .071 .112 .148 .033 .008 .027 014 .049 .044 .047 .183 .162 .113 .091 .212 .199	.113 .100 .092 .137 .150 .137 115 .101 050 .085 .062 374 .143 .043 .043 .064 .200 .197	.028 252 247 218 152 .015 061 .021 019 .035 .018 .125 200 231 231 037 008 389 389	069 037 049 369 418 175 .084 170 047 047 044 050 .067 329 339 .024 114 367 369

LH_MaxContactAt	309	.021	.095	.047	.066	.095	.723	.030	109	.133
FP_MaxContactAt	.073	.021	039	.129	079	.042	.719	065	.054	394
RH_MaxContactAt	226	005	.122	.057	.097	.010	.718	.012	145	.090
RF_MaxContactAt	.201	083	.014	.191	022	007	.620	.036	005	330
LF_MaxContactAt	080	.120	082	.022	112	.078	.583	144	.097	330
RF_StrideLength	.552	466	.179	072	.284	.078	159	.788	263	276
FP_StrideLength	.556	469	.181	076	.282	.077	159	.787	266	279
HP_StrideLength	.516	462	.323	115	.240	.075	152	.783	373	188
LF_StrideLength	.555	471	.182	080	.282	.077	155	.781	267	283
RH_StrideLength	.504	453	.380	110	.231	.068	146	.780	313	198
LH_StrideLength	.519	460	.243	117	.246	.081	155	.768	414	182
Support_Lateral	.106	306	100	.324	.173	.116	.215	.379	.038	.067
Couplpsi_LFLH_CStat_R	.044	047	.540	100	020	.073	.062	.146	926	117
Phlps_LFLH_CStat_R	.088	089	.533	091	017	.080	.069	.174	921	145
CoupDig_RFLH_CStat_R	.015	022	.501	079	070	.071	.019	.104	919	114
PhDig_RFLH_CStat_R	.020	016	.483	074	115	.040	.013	.105	914	077
CoupIpsi_LHLF_CStat_R	.074	136	.545	087	.017	.080	.034	.166	889	177
CoupDig_LHRF_CStat_R	.152	164	.497	087	.056	.073	039	.266	718	481
CoupGir_LFRF_CStat_R	.273	219	.429	072	.150	.136	.054	.375	564	677
PhGird_LFRF_CStat_R	.276	224	.406	048	.120	.159	.095	.385	549	659
CoupGir_RFLF_CStat_R	.260	229	.449	114	.146	.138	.075	.344	596	638
PrintPos_LP	.325	029	002	.250	.122	.084	.130	083	.219	613
PrintPos_RP	.341	033	102	.219	.143	.034	.106	108	.073	586

2

3 Table S4: Component correlation matrix for CatWalk PCA

Component 7 3 5 6 8 9 2 4 10 1 1.000 -.009 .012 .033 .111 .040 -.133 .198 -.034 -.306 1 2 -.009 1.000 -.060 -.016 -.206 -.053 .000 -.252 .047 -.050 3 .012 -.060 1.000 -.092 -.046 .036 .001 .088 -.490 -.138 4 .033 -.016 -.092 1.000 -.027 .043 .183 .029 .118 .000 5 .111 -.206 -.046 -.027 1.000 .010 -.023 .050 .012 -.094 .040 6 -.053 .036 .043 .010 .053 .028 -.070 1.000 -.029 7 -.133 .000 .001 .183 -.023 .053 1.000 .040 -.026 -.120 8 .198 -.252 .088 .029 .050 .028 .040 1.000 -.210 -.090 9 -.034 .047 -.490 .118 .012 -.070 -.026 -.210 1.000 .153 10 -.306 -.050 -.138 .000 -.094 -.029 -.120 -.090 .153 1.000

Component correlation matrix

1 Table S5: Structure matrix for CatWalk- Open Field- Vertical Pole PCA

2 For abbreviations see Table S2

Structure matrix

	Compo	nent											
	1	2	3	4	5	6	7	8	9	10	11	12	13
OF_WholeAvSpeed	.950	162	.163	037	197	.046	631	.105	236	.263	.085	076	097
OF_PerAvSpeed	.944	154	.177	029	187	.042	627	.108	230	.265	.082	069	094
OF_PerDist	.934	173	.160	054	172	.044	444	.114	228	.253	.068	044	129
OF_distTot	.932	169	.161	044	181	.045	627	.112	232	.262	.074	058	117
OF_RearTot	.926	094	.169	038	198	.053	597	.064	240	.245	.089	120	.040
OF_dist10	.903	143	.172	046	169	.048	604	.101	206	.256	.076	055	130
OF_Rear10	.884	087	.169	033	187	.057	575	.052	220	.235	.091	114	.041
OF_dist15	.883	128	.171	062	164	.032	585	.127	193	.221	.078	033	111
OF_dist20	.870	110	.149	083	150	.035	557	.104	183	.202	.051	043	083
OF_Rear15	.864	073	.145	041	162	.054	571	.060	214	.214	.090	110	.057
OF_Dist05	.858	232	.122	.008	193	.045	608	.097	270	.291	.063	069	106
OF_Rear20	.852	035	.113	063	165	.025	551	.043	198	.211	.071	113	.047
OF_Rear05	.842	148	.197	007	217	.063	523	.084	253	.245	.077	100	.012
OF_WholeRest	.752	119	.060	.009	230	.044	480	029	214	.176	.143	200	.150
OF_PerRest	.743	145	.074	.003	206	.053	360	029	218	.161	.144	184	.150
OF_CenAvSpeed	.452	018	.007	102	237	.074	.131	.069	071	.112	001	097	211
HP_PrintArea	153	.962	.024	180	012	010	.118	061	.102	127	071	046	376
FP_MaxContactArea	074	.960	.016	105	201	.029	.066	.050	.103	153	061	089	360
FP_PrintArea	040	.957	.052	079	252	.039	.048	.050	.099	153	055	099	365
RF_PrintArea	042	.955	.053	079	244	.033	.040	.039	.104	149	087	090	357
HP_MaxContactArea	150	.954	015	188	056	010	.118	050	.086	123	060	069	394
RF_MaxContactArea	085	.953	.014	101	186	.011	.063	.039	.108	150	093	077	352
LF_MaxContactArea	064	.951	.019	107	211	.047	.069	.060	.097	154	029	098	361
LF_PrintArea	038	.949	.051	078	256	.046	.056	.061	.094	157	023	107	368
LH_PrintArea	145	.944	.021	173	045	.045	.124	037	.047	189	026	064	366
LH_MaxContactArea	150	.934	020	182	080	.028	.131	031	.035	183	026	083	381
RH_PrintArea	157	.930	.025	180	.026	072	.107	084	.161	063	118	024	368
RH_MaxContactArea	144	.928	009	187	026	053	.099	070	.141	061	095	050	390
FP_PrintLength	.002	.910	.083	036	321	.054	.032	008	.091	131	044	094	342
HP_PrintLength	183	.905	.074	110	.136	009	.127	054	.135	095	092	.044	229
LF_PrintLength	.008	.896	.065	032	310	.062	.038	.006	.078	127	010	099	352
FP_PrintWidth	208	.891	035	014	090	.048	.116	143	.117	117	093	.000	287
RF_PrintLength	003	.888	.098	038	318	.045	.026	022	.100	130	075	086	319
LH_PrintLength	167	.881	.090	095	.073	.073	.118	043	.070	165	032	.026	222
LF_PrintWidth	172	.862	036	.003	125	.074	.106	118	.075	102	.010	038	297
RF_PrintWidth	229	.849	030	029	047	.020	.118	156	.151	124	187	.038	254
RH_PrintLength	183	.842	.051	116	.183	089	.124	059	.191	031	142	.055	217
LH_PrintWidth	276	.802	.044	107	.284	.014	.197	199	.049	169	064	.093	219
HP_PrintWidth	327	.796	.032	107	.376	028	.206	260	.143	077	102	.136	212
RH_PrintWidth	342	.682	.013	093	.422	071	.190	291	.230	.014	129	.162	182
FP_Swing	.183	.150	.883	016	077	.042	078	047	.218	131	026	.009	018
RF_Swing	.182	.138	.864	008	074	.027	078	049	.220	130	.135	.012	022
FP_SwingSpeed	179	.190	863	114	360	003	.038	343	162	.093	071	042	227

LF_SwingSpeed	178	.184	858	114	351	017	.034	339	161	.090	.032	046	2
LF_Swing	.186	.158	.857	022	074	.054	070	031	.204	124	178	.002	02
HP_SwingSpeed	326	.046	856	253	.123	062	.140	383	118	.116	100	.098	0
RF_SwingSpeed	177	.192	846	112	359	.012	.039	336	159	.094	173	038	2
LH_SwingSpeed	295	.078	840	260	.093	.090	.127	365	168	.057	049	.072	0
RH_SwingSpeed	336	.017	816	229	.141	200	.145	372	060	.151	140	.113	0
LF_Stand	191	.057	.740	.441	.478	.087	.126	120	.359	142	014	.328	.30
FP_Stand	195	.051	.733	.442	.469	.098	.129	134	.370	153	083	.322	.3′
RF_Stand	196	.047	.721	.436	.453	.103	.127	140	.373	161	139	.317	.3
LH_Stand	230	006	.660	.325	.638	.108	.148	166	.272	151	072	.345	.3
Run_Duration	111	137	.655	.338	.553	.025	.103	002	.275	133	036	.280	.3
LH_Swing	.363	.139	.652	.210	534	040	218	.050	072	.045	.003	121	1
HP_Swing	.405	.149	.651	.219	555	.138	260	.077	071	.103	.055	128	1
RH_Swing	.397	.151	.594	.200	509	.260	259	.108	056	.129	.080	127	1
BOS_HP	.038	.274	.365	128	.092	.067	003	.338	.218	176	113	.074	*
Couplpsi_LFLH_CStat	142	105	.209	.901	.096	134	.037	089	018	.096	.118	.261	.1
Couplpsi_LHLF_CStat	.156	.145	180	883	099	.162	046	.144	017	059	098	267	*
Phlps_LFLH_CStat	127	093	.168	.880	.082	159	.034	043	083	.136	.132	.235	.0
StepSequence_AB	.003	126	.291	.864	.142	.067	020	.032	036	.119	021	.174	.0
Couplpsi_RFRH_CStat	097	080	.225	.855	.092	.401	.039	043	.006	.108	133	.214	.0
Phlps_RFRH_CStat	098	075	.222	.851	.092	.404	.043	036	008	.114	126	.210	.1
Couplpsi RHRF CStat	.101	.131	183	824	060	379	066	.094	020	032	.136	201	
StepSequence_AA	061	.019	195	816	066	080	.039	102	.031	090	038	103	(
CoupDig_RFLH_CStat	.029	.013	016	706	066	.348	003	022	.060	175	.359	133	(
	.029	.093			060	.340	003	022	.000		.359		(
PhDig_RFLH_CStat	029		011 .086	705			.002	.051	.032	150		139	 .0
CoupDig_LHRF_CStat		057		.705	.066	324				.135	355	.127	
CoupDig_LFRH_CStat	037	.166	050	668	070	459	.028	.017	.082	137	338	049	'
CoupDig_RHLF_CStat	.050	101	.117	.668	.072	.429	028	.034	073	.086	.355	.048	.0
PhDig_LFRH_CStat	050	.131	042	665	046	447	.040	.033	.078	128	334	045	(
HP_DutyCycle	391	109	.077	.071	.931	044	.234	129	.166	015	087	.358	.3
LH_DutyCycle	335	080	.103	.066	.893	.137	.200	100	.100	058	021	.334	.3
RH_DutyCycle	408	122	.048	.069	.880	200	.243	143	.211	.015	137	.346	.2
Support_Diagonal	.320	.236	290	354	774	.002	228	.081	299	.117	.063	388	4
FP_DutyCycle	323	.019	.257	.585	.719	.092	.178	009	.197	.001	069	.388	.2
RF_StrideLength	018	.508	339	267	699	.028	041	334	096	.006	087	175	4
FP_StrideLength	014	.511	350	270	699	.025	042	329	101	.010	081	176	
LF_StrideLength	011	.511	358	271	695	.022	044	322	105	.014	074	177	4
HP_StrideLength	.018	.449	395	218	685	.022	075	384	270	.211	071	150	:
RF_DutyCycle	306	.029	.255	.552	.684	.108	.171	007	.191	004	292	.367	.2
LF_DutyCycle	308	.009	.234	.559	.682	.067	.167	010	.185	.005	.165	.371	.2
LH_StrideLength	.007	.450	390	224	675	.016	069	385	337	.087	051	153	:
RH_StrideLength	.022	.437	385	208	671	.028	073	364	178	.303	090	142	:
HP_Stand	251	018	.647	.329	.652	.046	.162	174	.321	112	091	.355	.3
RH_Stand	265	027	.618	.321	.648	013	.172	182	.362	078	110	.357	.3
Support_Single	.340	135	.093	038	598	.023	133	.151	012	067	.082	242	(
Support_Lateral	.269	.141	.146	238	532	.067	153	.079	.003	083	.102	091	(
BOS_FP	225	116	.124	.390	.434	.026	.173	.186	.122	046	016	.264	.1
Support_Zero	.166	187	.055	.049	382	.038	075	.051	060	048	.069	155	(
Support_Girdle	.222	161	.062	.077	330	028	136	.187	.016	006	.111	186	(
	1		1	1						1	1	1	1

	004	0.40	0.44	047	005	0.07	004	000	044	000	000	0.47	000
PhGird_LHRH_CStat	.084	.040	.041	017	025	.937	024	.083	011	.038	.323	047	033
CoupGir_RHLH_CStat	079	.080	019	032	.023	879	.010	.008	024	.001	284	.023	058
StepSequence_CB	042	.075	127	089	013	631	.026	.070	.038	.081	571	023	.011
OF_CenPerm	.525	106	.075	003	105	.000	988	.064	170	.199	.067	080	011
OF_PerPerm	525	.106	075	.003	.105	.000	.988	064	.170	199	067	.080	.011
OF_PerTiCenTot	.525	107	.075	003	106	.000	988	.064	170	.199	.067	080	011
OF_PerDistCenTot	.512	119	.005	070	146	.017	966	.063	183	.176	.068	118	042
OF_PerDistCen15	.480	106	.038	090	117	.012	885	.045	178	.129	.075	102	035
OF_PerTiCen15	.472	092	.085	030	090	.003	883	.047	160	.144	.072	076	008
OF_PerTiCen10	.399	118	.025	.003	091	.006	877	.031	118	.174	.066	103	022
OF_PerDistCen10	.425	151	037	066	117	.013	876	.041	149	.168	.067	120	043
OF_CenDist	.770	133	.136	019	167	.039	869	.089	200	.236	.074	075	075
OF_PerDistCen20	.510	065	.026	099	095	015	846	.093	128	.160	.034	043	.009
OF_PerTiCen20	.493	051	.081	039	065	036	837	.089	112	.160	.043	013	.029
OF_NbEntriesCen	.802	137	.119	010	178	.037	824	.094	214	.244	.078	076	082
OF_PerDistCen05	.429	142	002	.038	172	.031	815	.045	219	.197	.053	130	047
OF_PerTiCen05	.439	121	.064	.074	131	.039	811	.051	204	.216	.047	086	041
OF_CenRest	.542	008	002	.023	230	.004	683	021	135	.164	.092	185	.104
FP_MaxContactAt	074	.057	.088	.059	.040	.018	.018	.715	.146	070	022	.140	.095
RF_MaxContactAt	.011	.163	.134	.019	076	077	022	.612	.086	026	015	.074	.017
LF_MaxContactAt	134	072	.009	.079	.145	.110	.052	.561	.158	092	024	.156	.141
PrintPositions_LP	.189	.318	.276	108	010	.124	133	.448	.306	050	048	065	438
Phlps_LFLH_CStat_R	.216	070	163	.019	077	.001	194	057	940	.385	.075	059	044
Couplpsi_LFLH_CStat_R	.197	101	157	.032	030	001	184	082	936	.388	.072	014	011
CoupDig_RFLH_CStat_R	.185	130	110	.094	024	007	165	103	928	.340	.074	017	.038
PhDig_RFLH_CStat_R	.185	126	099	.128	032	017	165	103	925	.337	.051	031	.029
Couplpsi_LHLF_CStat_R	.197	075	145	008	100	009	194	073	920	.377	.086	088	091
CoupDig_LHRF_CStat_R	.121	.024	209	051	266	058	096	029	632	.285	.051	222	366
StepSequence_RB	141	.031	.142	.090	.025	.016	.022	127	.367	288	030	.033	.101
Couplpsi_RFRH_CStat_R	.246	155	100	.091	018	.027	206	006	368	.960	007	.002	.002
CoupDig_LFRH_CStat_R	.212	164	048	.144	.020	.034	173	021	325	.949	.000	.033	.036
Phips RFRH CStat R	.254	115	087	.090	040	.041	198	003	369	.947	008	002	015
PhDig_LFRH_CStat_R	.217	172	059	.166	.019	.039	175	017	338	.945	.013	.034	.025
Couplpsi_RHRF_CStat_R	.228	132	036	.087	065	.037	194	052	306	.916	013	031	026
CoupGir_RHLH_CStat_R	.277	220	024	.129	.063	.013	225	084	674	.837	.037	.019	.114
PhGird LHRH CStat R	.289	194	021	.123	.059	.035	225	057	702	.826	.048	.008	.075
CoupGir_LHRH_CStat_R	.284	223	036	.131	.053	.011	232	080	735	.799	.052	.001	.097
StepSequence RegularityIndex	.150	213	040	.103	.045	.027	157	035	550	.619	.075	.022	.041
StepSequence_RA	108	.079	.028	056	061	.050	.104	040	.131	532	.059	009	.034
CoupDig_RHLF_CStat_R	.081	.084	226	056	270	.030	117	.025	398	.480	.070	166	341
CoupGir LFRF CStat						.287			098	.011			
	.070	002	.030	004	029		078	.030			.936	003	051
PhGird_LFRF_CStat	.056	.005	.041	002	018	.295	057	.010	085	.000	.924	.015	032
CoupGir_RFLF_CStat	008	.090	.039	024	008	243	.010	.066	.023	.044	913	057	055
StepSequence_CA	.137	.088	078	154	109	.582	061	.035	074	050	.604	107	003
Pole_TimeComplete	.063	105	.010	.113	.041	001	005	.130	.000	.038	.040	.909	.061
Pole_TimeTurn	.126	066	.053	.105	.070	.011	065	.171	.005	.055	.037	.829	.037
Pole_TimeDown	055	126	052	.112	016	019	.087	.033	006	.002	.040	.798	.080
HP_MaxContactAt	.153	461	014	059	027	.039	101	.421	101	.097	.040	.022	.678
CoupGir_LFRF_CStat_R	.140	.224	307	219	415	.046	133	.136	350	.208	.044	218	648
PhGird_LFRF_CStat_R	.164	.220	265	190	421	.061	156	.171	341	.201	.060	227	634

RH_MaxContactAt	.119	369	008	050	026	004	052	.386	086	.078	.000	.022	.616
CoupGir_RFLF_CStat_R	.138	.178	353	241	367	.008	161	.178	410	.234	.050	212	589
LH_MaxContactAt	.150	445	017	056	023	.070	122	.362	089	.093	.065	.017	.587
PrintPositions_RP	.209	.349	.249	132	011	.000	132	.431	.126	190	.019	084	437

2

3 Table S6: Component correlation matrix for CatWalk- Open Field- Vertical Pole PCA

Component correlation matrix

Component	1	2	3	4	5	6	7	8	9	10	11	12	13
1	1.000	129	.155	065	235	.048	477	.124	186	.190	.084	128	057
2	129	1.000	.023	106	076	.021	.075	076	.094	101	076	019	323
3	.155	.023	1.000	.183	.115	.049	050	.105	.190	102	.010	.059	.104
4	065	106	.183	1.000	.102	.053	.016	013	.001	.087	010	.157	.129
5	235	076	.115	.102	1.000	017	.117	030	.126	006	041	.232	.223
6	.048	.021	.049	.053	017	1.000	004	.007	001	.000	.256	006	.008
7	477	.075	050	.016	.117	004	1.000	043	.152	166	059	.080	.013
8	.124	076	.105	013	030	.007	043	1.000	.048	034	.023	010	.004
9	186	.094	.190	.001	.126	001	.152	.048	1.000	378	086	.085	.071
10	.190	101	102	.087	006	.000	166	034	378	1.000	027	.011	037
11	.084	076	.010	010	041	.256	059	.023	086	027	1.000	043	.007
12	128	019	.059	.157	.232	006	.080	010	.085	.011	043	1.000	.135
13	057	323	.104	.129	.223	.008	.013	.004	.071	037	.007	.135	1.000

4

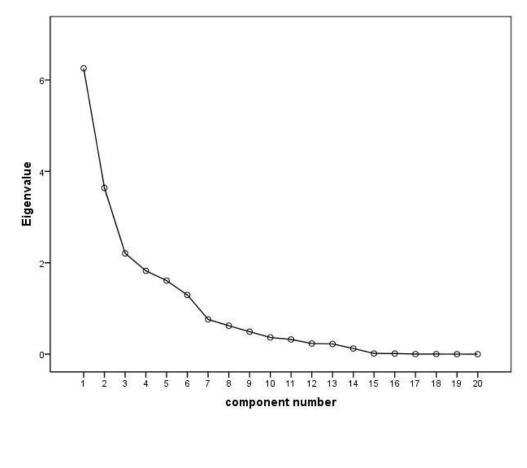
5 Table S7: Parameters of the IMPC data set

IMPC parameters
OF_Distance
OF_Rearings
OF_ArenaAverageSpeed
OF_ArenaRestingTime
OF_PeripheryDistance
OF_PeripheryAverageSpeed
OF_PeripheryRestingTime
OF_CenterAverageSpeed
OF_%CenterTime
OF_%CenterDistance
OF_CenterRestingTime
OF_CenterEntries
HC_Speed_mean

HC_Distance_mean
HC_Rearings_mean
GS_FP_bw_ratio
GS_4paws_bw_ratio
SHIRPA_activity
RR_PassiveRotations
RR_mean_Latency
OF-Open Field HC- home cage GS- Grip Strength FP- front paws bw- body weight RR- Rotarod



2 Fig. S1: Scree plot for the IMPC data set



1 Table S8: Structure matrix for the IMPC data set

2 For abbreviations see Table S7

Structure matrix

	Component								
	1	2	3	4	5	6			
OF_Distance	.982	.150	.181	.393	.214	.125			
OF_ArenaAverageSpeed	.945	.247	.446	.418	.282	.158			
OF_PeripheryAverageSpeed	.895	.328	.449	.416	.316	.162			
OF_CenterAverageSpeed	.830	303	.293	.271	.099	.068			
OF_PeripheryDistance	.776	574	.073	.212	.067	.050			
OF_%CenterTime	.004	.986	.126	.128	.189	.081			
OF_%CenterDistance	.058	.977	.117	.151	.166	.080			
OF_CenterRestingTime	.029	.757	.612	.095	.258	.108			
OF_CenterEntries	.586	.702	.133	.311	.203	.112			
OF_ArenaRestingTime	.179	.368	.950	.191	.300	.156			
OF_PeripheryRestingTime	.245	116	.894	.205	.224	.141			
OF_Rearings	.380	.082	.832	.095	.135	.086			
HC_Speed_mean	.257	.059	.087	.963	.154	.024			
HC_Distance_mean	.257	.060	.088	.963	.154	.025			
HC_Rearings_mean	.291	.131	.162	.671	.238	.123			
GS_FP_bw_ratio	.116	.155	.181	.170	.933	.182			
GS_4paws_bw_ratio	.132	.050	.157	.182	.924	.145			
SHIRPA_activity	.364	.277	.271	.369	.386	.188			
RR_PassiveRotations	009	012	004	064	.068	.858			
RR_mean_Latency	.147	.096	.189	.172	.232	.847			

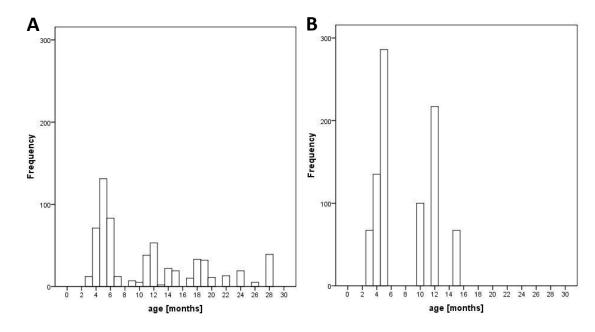
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4

5 Table S9: Component correlation matrix for the IMPC data set

Component correlation matrix

Component	1	2	3	4	5	6
1	1.000	.031	.238	.327	.171	.100
2	.031	1.000	.142	.116	.159	.068
3	.238	.142	1.000	.148	.221	.119
4	.327	.116	.148	1.000	.232	.080
5	.171	.159	.221	.232	1.000	.192
6	.100	.068	.119	.080	.192	1.000



2 Fig. S2: Histogram for age in the 7.1 (A) and XT CW version (B)