Individuality across environmental context in *Drosophila melanogaster*

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Abstract

Over the past decade, several studies have demonstrated that idiosyncratic animal behaviors remain stable over long time periods. The stability of individually variable behaviors over time is often referred to as an animal's individuality, or personality. However, most experimental studies have focused on individuality in a single, well-defined environmental context, whereas it is well-established from population studies that animal behavior is highly context-dependent. The 'person-situation debate' in humans and decades of observations of animal individuality under intrinsically variable natural conditions raise the question of whether and to what extent animal behavior remains stable across different situations, such as changing environmental contexts. For instance, one individual might be generally more visually guided than another, or rely only on one particular visual cue, or even on this very cue only in a specific environmental context. Here, we use a combination of both well-established and novel behavioral assays to demonstrate the relationship between individual behavior and variable environmental context under tightly controlled laboratory conditions in the model system *Drosophila melanogaster*. The stability of three individual traits (termed exploration, attention, and anxiety) was investigated under changing environmental contexts (temperature, visual cues, arena shape), in both walking and flying flies. We find that individuality is highly context-dependent, but even under the most extreme environmental alterations tested, stability of behavior always persisted in at least one of the traits. Furthermore, our quantification reveals a hierarchical order of environmental features influencing individuality. In summary, our work demonstrates that, similar to humans, fly individuality persists across different contexts, and individual differences shape behavior across variable environments, thereby making the underlying developmental and functional mechanisms amenable to genetic dissection.

Introduction

Animal species display enormous interindividual differences in behavior. The temporal stability of these interindividual differences allows for defining these idiosyncrasies as animal individuality, also called animal temperament, behavioral syndrome, or even personality [1-4]. Animal individuality has been described for numerous species and behaviors, including: mouse exploration behavior [5], great tit feeding preferences [6], octopus threat and feeding behavior [7], pea aphid threat responses [8], as well as vinegar fly handedness [9,10], olfaction [11], phototaxis [12], and object orientation [13]. The temporal stability of individuality allows for discriminating it from short-term behavioral changes, like internal state variations. Furthermore, temporal stability of behavior allows modifications by learning [14] and behavior-based natural selection [15]. Historically, the 'nature vs. nurture' debate (i.e., genome vs. environment) has shaped our understanding of the origins of behavioral individuality [16]. Stochastic developmental processes are a more recently found factor [13,17], providing an additional layer to gene-environmental interactions [18]. Most experimental studies on animal individuality have focused on a single behavior set in a singular environment, yet commonly used definitions of animal individuality in predominantly field-based disciplines such as behavioral ecology [3], and the definition of human personality entails behavioral stability across situations [19]. Similar to humans, animals will never re-encounter the exact same situation under natural conditions, and individuality can be an important factor in determining the ecological success of an animal and, as a consequence, its entire population [20]. It is, therefore, crucial to understand how an animal's individuality interacts with variable environments. Despite decades of interest in animal individuality across situations, the experimental evidence remains insufficient, and the results are contradictory, with studies arguing for [21-24] and against [25,26] individuality across situations with many finding mixed results [27-29]. Here we sought to systematically investigate the stability of individual behavioral differences across situations using laboratory-controlled variable environmental conditions.

There is ample evidence for the stability of human personality and animal individuality over time [30,31]. Intuitively, humans attribute stable personalities to other humans [32]. The stability

of human behavior across situations, including environmental context, nevertheless remains highly debated, with several studies arguing for [30,33] and against it [19,34]. Summarized as the 'person-situation debate', psychologists have argued for over a century about the relationship between personality and the situational environment, placing this question at the core of personality psychology [19]. Animal models for behavioral stability across situations provide an opportunity to address this question. In our study, the situation is the sum of all environmental features where a specific behavioral task is performed. The work presented here establishes a quantitative animal model to investigate this and related questions about behavioral stability across environmental situations.

The vinegar fly *Drosophila melanogaster* provides an enormous toolset for studying individuality and its origins across multiple scales, from neurons to circuits, to behavior. This comprises a rapidly unfolding connectome [35,36], an extensive collection of cell type-specific, binary expression systems [37,38], and a similarly extensive collection of tools for manipulating neuronal function, including neuronal silencers [39,40], activators [41], and thermogenetic [42] and optogenetic [43,44] effectors. Hence, the availability of tools and reagents for manipulating small, defined sets of neurons is no longer a major limitation in answering questions about individual behavioral differences. Instead, what has been missing is an integrative approach combining different high-throughput behavioral assays, which would allow for behavioral screening and classification of large quantities of individual flies for several different behavioral parameters across a variety of environmental contexts.

In recent years, mainly low-throughput, single-fly visual assays like the classical two-stripe Buridan assay [45,46], arenas for tethered walking [47] or virtual flight [48] have been used to study behavioral individuality in vison-based tasks [13,49]. In Buridan's paradigm, flies walk between two inaccessible visual targets (two vertical high-contrast stripes). The stripes are presented in a homogenously illuminated surrounding with no other visual or non-visual cues. Apart from work on individuality involving Dorsal Cluster Neurons (DCNs) [50], also known as Lobula Columnar Neuron 14 (LC14) [51], Buridan's paradigm has been used to show the importance of the central complex for normal walking speed [52] and for studying photoreceptor

function or development [53,54], and visual memory [55]. Other circular arenas have been frequently

used to study locomotory activity without visual stimuli [56-58].

In contrast to these low throughput assays, high throughput assays for studying the

behavioral responses of large amounts of animals have been used primarily for genetic

screens. Examples include apparatuses to study phototaxis [59], geotaxis [60], olfactory learning

[61,62], and circadian rhythms [63]. The Ethoscope [64] and MARGO [65] are recent additions to this

repertoire that offer several additional high-throughput assays. Except for the *Drosophila*

activity monitors (DAM, Trikinetiks), Ethoscope, and MARGO, these group assays cannot

distinguish individual behavioral differences. Only the Ethoscope and MARGO provide some

visual modules. Hence, progress towards understanding individuality in visually guided

behaviors in a high-throughput fashion remains meager, with few exceptions [9,10,12].

To investigate the influence of the environment on the stability of visually guided

individuality and, to systematically quantify individuality across many behavioral traits and their

dependency on the environment in walking flies, we built a new multi-parameter assay called

Indytrax. Indytrax is a compact and affordable Mathworks MATLAB-based modular assay and

tracker for high-throughput individual ethological quantification. The assay includes swappable

modules for multiple high-throughput Buridan paradigms, a Y-maze choice assay for studying

the decision-making and locomotor handedness [10], and a simple setup to quantify locomotor

activity in large groups of flies [57]. We used this tracker to determine the relative contribution of

various environmental features (temperature, visual cues, arena shape) to the stability of

individual behavior. Our results show that identically defined visual stimuli (stripe color, width,

and length) do not necessarily evoke the same behavior in an individual when presented in

different environments. Conversely, we find that specific individual behavioral traits (attention,

exploration, and anxiety, categorized via multiple behavioral parameters) remain stable even

in different environments. Finally, using an LED-based virtual flight arena, we show that

attention towards a stimulus remains stable even across different modalities of locomotion

(walking versus flying) when using the same Buridan-type visual stimulus. Hence, we find

stable behavioral traits across situations in our fly model, in analogy to what has been

demonstrated for humans [30,33]. Altogether, our results provide the foundation for understanding

the complex interaction between individuality and the environment, including the question of

why individual behavior is more unpredictable in variable naturalistic conditions, much like the

stability across situations in human subjects [19].

Materials and Methods

Animals

All experiments were performed using one week old wild-type *Drosophila melanogaster* (CantonS (Source: Bassem Hassan) and TopBanana [66]) reared on a standard *Drosophila* yeast cornmeal medium (7.5g Agar, 64g cornmeal, 160g yeast, 85.5ml sugar cane syrup in 1L of water) at 25°C and 50% relative humidity under a 12/12 light/dark rhythm. Individual flies were placed into the behavioral arenas using a fine brush and were allowed to accommodate for about 10 minutes before any experiment. The behavioral arenas and flies were cooled on ice for Y-maze experiments before gently loading single flies into the chambers using a fine brush. For experiments with clipped wings, the wings were clipped close to the hinge under CO_2 anesthesia, and flies were tested after being allowed to recover within individual vials for at least two days.

Setups

Setup 1: Standard Buridan Arena

Assay

Two Buridan setups were used: type A uses fluorescent tubes for background illumination (as described in [13,67]). Type B utilizes a panoramical LED matrix (RGB, 384x128 pixels, pixel size 0.9°, 120Hz frame rate, 5kHz refresh rate, < 1ms Pixel update latency), which is easily addressable via HDMI for stimulus presentation (see Fig. 1a). For a detailed description of the assays and tracking procedures, see supplemental Fig. S1. Flies were walking on a circular platform (diameter 12cm) between two opposing black vertical stripes (angular extent 12° horizontal and 90° vertical measured from the center of the platform). A trench of water surrounded the platform to prevent the flies from escaping (type A) or a clear plastic hemispherical dome coated with Sigmacote (Sigma-Aldrich, SL2-25ML) from the inside (type B). Single-fly walking behavior was recorded at 25°C and 50% relative humidity for 15 minutes (450x450, 8-bit greyscale, 20 Hz, AVI, Mjpeg compression, 90% quality). Before changing an experimental condition, flies were placed back into single vials for recovery.

Automated tracking procedure

Data was analyzed either in real-time using previously described code (type A, see [67])

or offline using a custom-written MATLAB code (type B, available for download from

github.com/LinneweberLab/individuality-assays, see supplemental Fig. S1c), respectively.

Thresholded video frames and overlaying trajectories are displayed continuously in a separate

window to allow the experimenter to assess tracking quality. The MATLAB code works as

follows: the circular region of interest (ROI) is automatically detected based on adaptive

thresholding and subsequent filtering of the thresholded ROI based on area. Once the ROI is

detected, a background image is generated by calculating the maximum intensity between a

recorded video's first and last frames. All flies that moved between those frames will not appear

in this background-only image. During tracking, the background image is subtracted from each

consecutive frame to simplify the thresholding of the fly. For each frame, the centroid of these

thresholded pixels is stored. Should more than one object be detected within a ROI (e.g.,

through pixel noise), the x-/y-coordinates of the object closest to the fly's position in the

previous frame are stored. If a fly is lost, the tracker keeps the last known position of the fly

until it is detected again.

Data output

Once tracking is completed, the raw tracking results are converted to real-world units

(mm) and saved into a .mat-file. Before further analysis, minimal movements are filtered out

by only updating the fly's position if a certain walking threshold (0.8mm/frame) is detected. For

each fly, a figure can be saved, allowing for a quick overview of the walking trajectory, and a

variety of behavioral parameters are also saved into a separate .txt-file to simplify further data

analysis. For a detailed explanation of calculated parameters, see supplemental table 1. For

an in-depth analysis of Buridan data, sophisticated data analysis routines describing many

behavioral parameters were previously developed [67]. To simplify data integration and

compatibility, the raw trajectories can also be saved to a pair of .TXT and .DAT files which can

later be read and analyzed by previously available software [67].

Setup 2: IndyTrax multi-arena platform

Assay

To allow for high-throughput behavioral quantification under various tightly controlled

environmental contexts, we designed the novel assay system IndyTrax. In short: single flies

are transferred to an array of small behavioral arenas. They are filmed from below against

near-infrared (850nm + Lee filters 87C IR-pass filter on camera objective) LED illumination

(Fig. 3). Compared to published Buridan assays [67], this allows tracking flies in the absence of

visible light, but also has the advantage that flies cannot see the camera lens which might

otherwise appear as a high-contrast object from within the arena. The assay was designed to

allow for the tracking of flies in various possible arena shapes and numbers. We created three

different 3D-printable arena types (supplemental Fig. S4e-g), which can be downloaded from

github.com/LinneweberLab/individuality-assays. For individuality analysis, we used the multi-

Buridan and Y-maze arenas. The setup was built from commercially available and 3D-printed

parts. For a part list, see the supplemental table 2. All 3D-printed parts and behavioral assays

were printed on a Prusa i3 mk3 printer using black or white PLA filament (Verbatim 55318) and

can be downloaded from github.com/LinneweberLab/individuality-assays. Similarly, the

MATLAB code tracking and data analysis can be downloaded from there. For detailed

instructions on how to install and run the code, see supplemental information. With a computer

running MATLAB already available, we estimate the building cost for a complete, temperature-

and humidity-controlled enclosure, including all behavioral assays, at about 3500€.

Temperature- and humidity-controlled enclosure

Behavioral arenas are located within a black climate-controlled metal enclosure.

Temperature and humidity within the enclosure can be regulated using two 3D-printed

humidifiers and a heating plate (see supplemental Fig. S4a). Ambient temperature and

humidity can be set at any desired value using this configuration and kept stable over a long

time (see supplemental Fig. S4c). Since small enclosed arenas are prone to heat build-up, we

placed all temperature- and humidity-sensors within the center arena (arena 13) for Buridan

experiments and close to the flies in Y-maze experiments. For Buridan experiments, this limits

the number of simultaneously trackable arenas to 24 but allows more accurate climate control

within the chambers.

<u>Small-scale Buridan-stimuli arrays</u>

We modified the basic design of previously published behavioral arenas optimized for

Buridan object orientation behavior [67] by implementing changes in four major areas to allow

for high throughput tracking: Firstly, we reduced the arena diameter to allow for filming an array

(5x5) of optically isolated Buridan arenas with only a single camera (Fig. 3). Secondly, flies are

walking on circular translucent acrylic plates (49mm diameter) which allows for filming flies

from below against infrared back-illumination, hence avoiding problems with dead angles that

would arise when filming cylindrical arrays from above. Thirdly, during experiments, flies are

contained on their respective acrylic platform by clear hemispherical plastic domes (Fig. 3b).

This removes the necessity of cutting the flies' wings before experiments and saves space by

making the water-filled moats of the original design superfluous. Plastic domes were coated

with Sigmacote (Sigma-Aldrich, SL2-25ML) from the inside to prevent the flies from sitting on

the dome surface. Fourthly, to provide a scalable, homogeneously illuminated white

background for optical stimulus presentation, the fluorescent tubes of the original design were

replaced with electroluminescent (el-)foil wrapped around translucent acrylic cylinders (Fig.

3b). Such el-foils can be cut to size and have low heat emission. They emit light evenly over

the whole surface, minimizing unwanted intensity cues within the arenas. El foils are driven at

2 kHz, thus providing virtually flicker-free white illumination to the flies. For creating the Buridan

object orientation stimulus, thin black stripes (12° wide as measured from the arena center)

were 3D printed and glued vertically to the inside of the arena walls 180° apart, thereby creating

high-contrast objects against a uniform white background (Fig. 3b). Individual arenas were

fitted into a 5x5 array. Every other arena within the array was rotated by 90° counter-clockwise

to increase the robustness of behavioral readouts by canceling out possible directional artifacts

created within the enclosure.

Walking behavior was recorded at 50% relative humidity for 15 minutes (1200x1200

pixel, 8-bit greyscale, 15 Hz, AVI, Mjpeg compression, 90% quality) with either the el-foil-

illumination on, in darkness, at 23°C or 32°C, respectively. After being tested in the Buridan

assay, the flies were put back into single vials overnight to allow for recovery before testing

them in the Y-maze assay the next day around the same time.

<u>Y-maze array</u>

We constructed a modular 3D-printable Y-maze array (144 arenas, 300x300x2mm)

based on a previously published design [65]. The Y-maze plates were placed between two

320x320x3mm borosilicate glass plates. The bottom side of the upper glass plate was coated

three times using Sigmacote (Sigma-Aldrich, SL2-25ML) to prevent the flies from walking on

the upper plate (Fig. 3c). Directly below the arenas and above the upper glass plate, we placed

a layer of white paper (Evercopy premium 80g, Ref. 1902C) for light diffusion and to restrict

the view of the flies outside of the arenas. For individuality experiments, we covered the outer

arenas with black tape and tracked flies only within 100 arenas located in the center. This

made keeping track of individual fly identities easier when switching between the different

behavioral assays. Videos of walking flies were recorded at 50% relative humidity for 30

minutes (1200x1200 pixel, 8-bit greyscale, 15 Hz, AVI, Mjpeg compression, 90% quality) under

white light illumination, in darkness, at 23°C or 32°C, respectively. For white light illumination,

we placed a 30x30cm LED panel (Tween Light 30 x 30 cm LED panel, 4000k, 16W) on top of

the arenas and temporarily removed the infrared-pass filter from the camera objective. Y-maze

experiments were all done consecutively on the same day (light on 23°C, light off 23°C, light

on 32°C, light off 32°C).

Automated tracking procedure

We developed a fully-automated MATLAB-based tracking code that allows users to adjust parameters like arena number and size and generates output data for all videos within a user-selectable folder. The code assumes only one fly per region of interest. This saves processing time and increases tracking robustness by not having to match multiple fly identities between frames. Before running the code, the experimenter must create a camera calibration file using the MATLAB camera calibration toolbox (see supplemental information). This allows for more accurate tracking results by correcting lens distortions (supplemental Fig. S4d). If the camera stays immobile, this procedure must be done only once, usually only a few minutes. As long as the arenas filmed are arranged in rows and columns, the code can handle virtually any shape or number of arenas, making it ideal for tracking flies across different behavioral assays (supplemental Fig. S4e-g). Regions of interest (ROIs) are detected based on adaptive thresholding and subsequent filtering of thresholded ROI-objects based on the area (supplemental Fig. S4e-g, middle). Once the ROIs are detected and sorted, a background-subtracted image is generated, allowing the user to determine a detection threshold, above which all pixels will be interpreted as foreground objects (supplemental Fig. S4e, third from left). During tracking, background subtraction and object thresholding is performed for each frame, and for each ROI the centroid of the thresholded fly is extracted and saved. Thresholded pixels and trajectories are displayed continuously in a separate window to allow the experimenter to assess tracking quality. If more than one object is detected within a ROI (e.g. through pixel noise), the x-/y-coordinates of the object closest to the fly's position in the previous frame is stored. If a fly is lost, the tracker keeps the last known position of the fly until it is detected again. The average tracking speed for 24 ROIs/flies was about 20hz with MATLAB running on an Intel i5 3Ghz CPU. For tracking 400 ROIs/flies average tracking speed was about 5hz.

Data output

Data output is the same as described for our standard Buridan arenas.

Setup 3: Virtual flight simulator

Assay

We designed a novel, easy-to-assemble, and easy-to-use virtual flight simulator using

affordable 3D-printable and off-the-shelf parts (Fig. 5a-c) to quantify behavioral responses

during flight. For a detailed description of the setup and custom code, see supplemental Fig.

S9. In short, using a tethering station (for description, see [48]), flies were cooled down for

immobilization and glued to a steel pin (0.1mm diameter, 10mm length, www.entosphinx.cz)

using UV-cured glue (Bondic). Once tethered, single flies were put between two neodymium

magnets (magneto-tether; upper magnet: diameter 5mm, lower magnet: ring, outer diameter

10mm, inner diameter 5mm). Once the appropriate distance between the magnet is found

(about 20mm), the magnetic field allows the flies to rotate freely around their yaw-axis while

keeping the steel pin in place (Fig. 5b). The tip of the steel pin is inserted into a V-shaped

sapphire bearing (1mm diameter) to reduce friction. For re-initiation of flight, air puffs can be

delivered from below via a membrane pump. A panoramic LED matrix surrounding the

magneto-tether was used for Buridan stimulus presentation (RGB, 256x128 pixels, 120Hz

frame rate, 5kHz refresh rate, pixel size 1.4°, angular extent 360° horizontal x 100° vertical,

stripe width: 12°). The LED matrix controller is connected to a computer via HDMI, allowing an

easy and flexible stimulus creation and presentation (Novastar mctrl 660 pro, 120Hz, < 1ms

Pixel update latency). Flies were filmed from below under near-infrared illumination (Fig. 5c),

and each fly's heading was tracked in real-time (90 Hz) using a custom-written MATLAB code

(available for download from github.com/LinneweberLab/individuality-assays). Before starting

the experiment, each fly was tested for 120s with an optomotor stimulus (rotating stripe pattern,

120°/s, 40x CW and 40x CCW, respectively) to assess the flies' ability to rotate smoothly (Fig.

5d, left). After this, flies were presented with the Buridan stimulus with varying contrast (100%,

50%, 100% inverted; stripe width 30°), and their heading was tracked for 3x4 min. Flies were

then carefully separated from the steel pin and transferred into single vials for overnight

recovery before being re-tethered and tested the following day around the same time. For flight

vs. walking experiments, flies were tested for 10 min while flying under a 100% contrast

Buridan stimulus (stripe width 12° from a fly's perspective) before being gently separated from

the steel pin and being transferred into the LED Buridan assay (Type B), where they were

tested for 15min under a 100% contrast Buridan stimulus (stripe width 12° from a fly's

perspective).

<u>Automated tracking procedure</u>

The code for tracking a fly's heading in 360° can be downloaded from

github.com/LinneweberLab/individuality-assays. Flight tracking is done in real-time at 90fps

(see supplemental Fig. S9c). A timestamp is created for each incoming camera frame, and two

different thresholded images are computed with different Gaussian filter settings. A strong filter

setting blurs out the legs of the fly and allows for thresholding only the fly's body and calculating

a heading angle between 0° and 180° through ellipse fitting. By subtracting the thresholded

body pixels from a weaker filtered version of the same frame, the limbs and wings (visible when

the fly stops flying) can also be tracked separately if the experimenter wants. To calculate the

heading between 0° and 360°, the centroid of thresholded body pixels is set to (0,0) within a

fictive coordinate system (see supplemental Fig. S9c), and the body axis is aligned with the x-

axis. The head region is assigned based on the sum of pixels within the 4 axis quadrants

(smallest area of either I+IV or II+III, respectively). Once the heading is calculated, it is stored

together with the original video frame for additional offline data analysis, if desired. When a fly

stops flying (wings get detected in the video), a command is sent to a connected Arduino,

which controls a relay to switch on a membrane pump and provide an air puff to reinitiate flight.

During tracking, the thresholded video is displayed together with the computed heading to

allow the experimenter to assess tracking quality.

<u>Data output</u>

After tracking, raw heading data, a graphical overview image, and a table containing

computed behavioral parameters are saved automatically for each fly which can be used for

further data analysis. For a detailed explanation of all calculated parameters, see supplemental

table 1.

<u>Data analysis</u>

All data analysis was performed with MATLAB. For angular calculations, we used

algorithm presented in [68]. After a test for normality (Shapiro-Wilk-Test), behavioral data

between groups was compared using the paired T-test. For correlating individual data, we

calculated Pearson correlation coefficients. For walking data, an arena's center region (edge-

corrected area) was defined as the circular area around the center with a radius of 80% of the

arena radius. The maximum color value (red) for generating heatmaps was set to the 95%-

quantile of the count distribution to increase data visibility.

Results

The goal of our study was to understand to what extent animal individuality is influenced by situational changes in the environment, i.e., how much of an animal's individuality remains after one or more environmental features change. Based on our previous work on animal individuality [13], we started our analysis in the multiparametric two-stripe Buridan's paradigm (Fig. 1a, supplemental Fig. S1 for technical details) [13,67]. As expected, we measured interindividually variable, yet remarkably stable behavioral responses, both on the individual (Fig. 1b, Fig. 2a, supplemental Fig. S3a) and mean group level (supplemental Fig. S2a), when tested in the exact same situation with all environmental features kept the same. This was true for the five representative behavioral parameters (out of 20 parameters total, see supplemental Table T1 and supplemental Fig. S8), defined as falling into three behavioral traits: exploration (% of time walked, walking speed), attention (vector strength, angular velocity), and anxiety (centrophobicity).

Modifying the visual stimulus within the Buridan arena (0, 1, and 2 stripes were tested, Fig. 1c, Fig. 2b, and supplemental Fig. S2b and S3b) significantly affected the mean group behavior across all three behavioral traits (supplemental Fig. S2b). Especially the population means of three parameters (% of time walked, vector strength, and centrophobicity) were strongly affected by the stimulus modification. Surprisingly, despite this change in population means, we observed only a minor effect on individuality (Fig. 1c and Fig. 2b, measured by a high Pearson correlation coefficient and a consistent rank within the distribution). The individuality for the traits exploration and anxiety persisted even upon modulation (variation of stripe number) and after removing visual cues from the arena. Only those parameters describing the fly's attention towards visual cues showed an unsurprising reduction in behavioral stability upon removal of the visual stimuli (Fig. 2b). Altogether, these results demonstrate that individuality persisted after stimulus modifications within the same arena, even when the modifications resulted in altered population responses.

To further characterize the surprising relationship between individuality and population means upon situational changes, we modified the homogeneity of background illumination in

the Buridan arena using two different diffuser materials (paper and acryl-based covers of

fluorescent ring lights) and compared these conditions to a newly designed Buridan LED arena

displaying the same visual stimulus (stripes of the same color, width, and height) (Fig. 1d, Fig.

2c and supplemental Fig. S2c and S3c). We measured no significant differences between the

population means of the two diffuser conditions and only minor group-based changes

compared to the LED Buridan (supplemental Fig. S2c). We then tested the correlation of

individual behavior across these conditions, as a measurement of behavioral stability,

expecting high correlations. Indeed, the different diffusers within the same arenas only had

minor effects on the behavior of individual flies (Fig. 1d). Conversely, the individual response

changed dramatically when the same fly was tested again in the LED arena (Fig. 1d). A fly's

attention and anxiety parameters quantified within the original arena did not allow for any

prediction for the LED arena (Fig. 2c). Significant correlations between the LED arena and the

original arenas were only measured for exploration parameters (Fig. 2c). This unpredictability

of individual fly behavior across situations was once again particularly surprising since group-

based performances remained similar (supplemental Fig. S2c). Finally, a detailed comparison

of males and females revealed only quantitative but no qualitative differences across all

parameters between the two sexes (supplemental Fig. S2 and supplemental Fig. S3).

Based on the surprising result that individuality [1] of visual orientation behavior is more

strongly influenced by the arena type (despite producing very similar population responses)

and, only to a lesser degree, dependent on the visual cue (despite strong differences in the

population response), we set out to characterize the contribution of environmental situations

to individuality even more quantitatively. We therefore built a multiplatform high-throughput

assay (Indytrax) to quantitatively study the behavior of the same individual flies in different

environmental situations (Fig. 3, technical details supplemental Fig. S4).

Using Indytrax, we compared the behavioral parameters of each fly in eight different

environmental situations (Fig. 4, supplemental Fig. S6, S7, S8). The flies were tested in two

independent arenas (Buridan and Y-maze) at two different temperatures (23°C and 32°C,

supplemental Fig. S5) and under both light and dark conditions. Our analysis showed that

temperature significantly affected exploration parameters on the population mean level. As expected, most animals tested were more explorative at higher than at lower temperatures (reviewed in: [69]) (supplemental Fig. S6). Despite these population changes, individuals kept their rank within the population independent of the changed baseline activity. For example, fly #1 was the least explorative, fly #2 exploration performance ranked medium, and fly #3 the highest (Fig. 4a). The graphical analysis of exploration parameters (Fig. 4b) confirmed that most individuals maintain their rank in the distribution under different temperatures and light-dark conditions, whereas the individual distribution rank changed between the different arena types. This relationship was less pronounced for attention parameters. The detailed correlative analysis between individuals underscores these observations (Fig. 4c and supplemental Fig. S7 and full analysis supplemental Fig. S8): Amongst the exploration parameters, the percentage of time walked significantly correlated between individuals (Pearson r=0.263-0.398) independently of temperature, illumination, and arena. Walking speed was highly correlated between individuals (Pearson r=0.81, 0.825) under temperature and illumination changes but uncorrelated in different arena types (Pearson r=0.091). Vector strength was the most correlated attention parameter. The individual correlations ranged from Pearson r=0.533 to -0197. Angular velocity significantly correlated for temperature and illumination (Pearson r=0.246, 0.354) but not in different arenas (Pearson r=0.087). The same is true for anxiety measured through centrophobicity (Pearson r=0.694, 0.740, 0.033). In summary, these data revealed that individuality measured as three traits is mostly unaffected by temperature and illumination but strongly affected by the arena type. Our analysis confirmed that even in situations with a large effect on the mean group behavior (like temperature), individual behavioral stability of idiosyncrasies remained high, while conversely, other environmental situations like the arena type had little impact on the mean but a rather massive effect on individual behavioral stability (Fig. S6). Qualitatively similar results leading to the same conclusions were obtained for both sexes (Fig. S6a, S7a) and two different genotypes (Fig. S6b, S7b)

Next, we drastically modified the behavioral situation while keeping the visual stimulus

as constant as possible. We, therefore, decided to investigate the influence of a fly's modality

of locomotion (walking versus flying) on individuality. To do this, we designed a new virtual

flight arena for tethered flight utilizing the exact same LED panels as were used in the LED

Buridan arena (Fig. 5a-c and supplemental S9 for technical details), to maximize the similarity

of the visual context. This flight simulator tracked a fly's heading in real-time in response to

dynamic or static panoramic visual stimuli with high resolution (1.4° pixel size). This allowed

us to quantitatively compare the visual decisions of both walking and flying flies in response to

standard optomotor and Buridan-like stimuli (Fig. 5d-e). We first tested individuality in flying

flies over time under the same and a set of modified visual stimuli (Fig. 6, supplemental Fig.

S10 and S11). Each fly was tested over two consecutive days under three different visual

conditions (dark stripes on a bright background with either 50% or 100% contrast, and an

inverted Buridan with bright stripes on a dark background and 100% contrast) (Fig. 6a).

Overall, the flies showed some individuality in visual attention across days and contrast. This

was also true for the inverted Buridan stimulus, in which the flies oriented themselves toward

the dark areas, similar to previous results [70]. To allow a quantitative comparison across

behavioral traits between walking and flight behavior, we derived an adapted set of parameters

for exploration and attention quantification (exploration parameters: # of pauses, Absolute

angular velocity; attention parameters: Vector strength, Angular velocity, and Median heading

axial, Fig. 6b). The resulting analyses revealed individuality in flying flies (Fig. 6c). Only the

inversion of the Buridan stimulus resulted in variable angles towards the black background on

different days. However, it is remarkable that on a given day, each fly selected an arbitrary

heading that was kept for the entire experimental length (Fig. 6a). Again, the quantitative

analysis between males and females did not reveal qualitative differences (Fig. S11).

After confirming that individuality also existed in flying flies, we tested how different

locomotion modalities affected a given animal's behavioral individuality. Behavioral parameters

of both walking and flying flies were quantified in response to the same panoramic LED screen

displaying an identical two-stripe Buridan stimulus (Fig. 7a, supplemental Fig. S12 and S13).

We expected one of three possible outcomes: 1. There is no stability of individual behavioral

traits across flying and walking, 2. Only some behavioral traits are similar across flying and

walking. 3. The fly's individuality remains stable across the same visual cue, independent of

the locomotive modality. Subsequent quantitative analysis indeed revealed a significant

correlation for angular velocity between flying and walking flies (Fig. 7b), indicating that a fly's

attention is, to some degree, preserved independent of the locomotor modality. In contrast, we

found no statistically significant correlations for any of the exploration parameters between

flying and walking. Finally, a more detailed analysis revealed mostly comparable results for

male and female data (supplemental Fig. S12 and S13).

In summary, the data presented here reveal a hierarchical influence of different

parameters within the environmental context on individuality (Fig. 8): First, we conclude that

time (at least within the frame of days) has a negligible influence on individual behavioral

responses (the basis of the definition of animal individuality in a single situation), since we

found stability of individual behavioral traits and parameters over time in both locomotor

modalities, walking and flying. Second, temperature had the second weakest impact on

individuality (although only tested for walking behavior). This contrasts the dramatically altered

population mean responses, a difference that can be explained by scaled individual responses

in which individuals keep the rank in the population. Third, the nature of the visual input

appears to influence the individual orientation responses but has little impact on the exploration

parameters. Finally, we identified the most critical situational environmental features that

determine individual responses as both the arena type and the locomotor modality (walking

vs. flying). Therefore, we conclude that the individuality of visually guided behaviors persists

specifically in some traits across situations, but the environmental context plays a prominent

role in defining individual behavioral responses.

Discussion

Our work presented here focused on the question of whether the combination of stable behavioral parameters ranging across different traits and therefore defining an animal's individuality are invariable across situations or whether they are re-shaped by different environmental contexts. This question is closely related to the questions in the human 'person-situation debate' [19]. Until now, in the neurosciences individuality across situations has been largely ignored since, the commonly used definition of animal individuality only refers to the repeatability of behavior over time in a single environmental situation [1]. The contrary is true for mainly field and observation-based sciences such as ethology and behavioral ecology that have since decades acquired observations for [28,29] or against [25,26] behavioral stability across situations, with some of the most convincing work probably found in hermit crabs [21,22], blue tits [23], and geese [24]. Our work establishes animal individuality across situations under several tightly controlled laboratory conditions, revealing that animal individuality shares remarkable similarities to what has been described in human personality research. In agreement with Vansteelandt, 2004 our results are in line with the concept of the personality triad, emphasizing the interplay of individual, situation, and behavior (in humans: person, situation, and behavior) [71].

Our results confirm that under the same environmental situation and stimulus, animals produce interindividually variable, but extremely stable behavioral responses when being retested [72]. Therefore, stable and individually variable behaviors likely result from individual brain properties that persist over long timescales (in this paper days as shown in previous papers months [13]), contrasting unstable random behavioral variations like sudden internal state variations. Hence, when presented with the same stimulus in the same situation, each individual responds variably but repetitively so, pointing towards a deterministic and hardwired basis of individual properties [10,13].

The variability of the situational environmental context makes individual animal behavior in the wild much less predictable than that under laboratory conditions [73]. In opposition to the vast human literature on the 'person-situation debate' [19,31,33,71,74], the

importance of environmental context in shaping individual animal behavior remains descriptive

[24] and experimentally largely unexplored, with few exceptions [21,22,27,75]. Here we closed this

gap in knowledge by quantitatively demonstrating how situational environmental and locomotor

context changes individual behavior and individuality in animals. Our results demonstrate that

situational environmental context has a pronounced effect on individual behavioral responses,

be it an alternative stimulus representation, a different arena, or another locomotive modality.

Similar to the human literature [19], the correlation coefficients are lower across situations

than in the same situation or environmental context. From this, we can conclude that under

more natural conditions with varying contexts, the stability of individual traits will be reduced

but remain measurable; therefore, for each situational context, the brain computes the

appropriate behavior. This is like the situation-behavior profile of the individual in the

personality psychology literature [76,77]. Each time the individual is placed in the same situational

context, the same behavior will be evoked, in a different context, a modified behavior. But

despite these modifications, the individual remains stable. In line with this, a vast literature

demonstrates the importance of animal individual differences for ecological communities under

natural conditions reviewed in: [20,78].

Our quantitative analysis of the effect of environmental and locomotor context on

behavioral stability revealed a hierarchy. As expected from animal and human studies, the time

frame of days had little effect on the behavioral outcome [8,10,13,19,32,72,79]. The hierarchy revealed

by our experiments argues for differences in the relevance of environmental features for the

traits we measured, which aligns with Mischel's strong or weak situations [80]. Therefore, it is

unsurprising that exploration can be best measured across situations within experiments of

walking flies, and attention towards a stimulus can be measured across movement modalities.

Our quantitative analysis demonstrates that situational context can be the most significant

factor in determining the individual behavioral outcome, even in a per-design visual assay. This

argues that individual behavior is determined by the entirety of the situation consisting of both

the stimulus and the environmental context. For each situation, each individual manifests its

individual behavior.

In the future, it will be important to investigate which individual neurophysiological

responses change when tested under exactly the same stimulus, but different situations. A

systematic analysis of relevant circuit elements for different behaviors will allow for determining

the exact neuronal substrates underlying individual behavioral computations within a fly's or

other animals' brains. This future research on animal models will provide important new insight

into the 'person-situation debate'.

On a more practical level, the importance of the situational environmental context for

shaping individual behavioral outcomes is also very important when comparing the results of

different research groups when using similar behavioral paradigms. In many instances, it is

impossible to replicate all components of a specific assay, let alone the conditions of the

experimental room. Our results show that such details substantially impact individual

behavioral responses and contribute to errors in replicating the results from another lab [81,82].

Only meticulous and detailed reporting of as many environmental features as possible will

hopefully alleviate some of these problems in the future.

In conclusion, our work confirms that animals including flies have, like humans,

individuality across situations. Similar to humans, animal individuality across situations is also

marked by lower correlation coefficients than repetitions of a single situation. The individual is

stable across situations producing individualistic behavior through an interaction with the

situation. This involves a hierarchy of influence of different environmental features on the

predictability of behavior across situations.

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Author Contributions

Conceptualization, T.F.M., G.A.L. and M.F.W.; Methodology, T.F.M., G.A.L. and M.F.W.; Investigation, T.F.M., C.K., M.A.H., F.F., T.M.; Resources, M.F.W.; T.F.M., G.A.L. and M.F.W.; Supervision, G.A.L. and M.F.W.; Funding Acquisition, M.F.W. and G.A.L

Declaration of Interests

The authors declare no competing interests.

Figure 1:

 Figure legends

611 **Fig. 1:** Setups for studying individuality walking flies. **a** Overview of a classical (left, fluorescent tubes)
612 and LED (middle) Buridan assay (behavioral platform diameter = 120mm). A MATLAB script provides
613 single fly tracking and analysis of behavioral key components (right). **b-d** persistence of individual
614 walking patterns across time (**b**), different stripe numbers (**c**), and changes in background illumination
615 (**d**).

Figure 2:

617 **Fig. 2:** Individual traits of walking flies persist over time but depend on the visual scenery. **a-c**
618 Correlation of key behavioral parameters over time (**a**), different stripe numbers (**b**), and changes in
619 background illumination (**c**). Correlation coefficients are displayed in the upper left corner of each plot.
620 *=p<0.05, **=p<0.01, ***=p<0.001, for detailed fly numbers per experimental condition, see
621 supplemental Fig. S2.

Figure 3:

Fig. 3: A high-throughput assay for studying the influence of temperature, illumination, and arena
shape on individuality. **a** Setup overview. Flies are filmed from below against a near-infrared backlight
and are walking within a 5x5 array of small Buridan chambers (**b**) or one of up to 144 Y-maze arenas
(**c**). The setup is housed within a temperature- and humidity-controlled enclosure (600 x 450 x 820
mm). The provided MATLAB code corrects lens distortions, automatically detects the arenas, and
tracks each fly's position over time, allowing for generating high-throughput single-fly behavioral data
(**d**).

Figure 4:

Fig. 4: The influence of temperature, illumination, and arena shape on individuality in walking flies. **a**
Individuality in walking patterns across different temperatures, illumination, and arena type. Diameter
circular arena = 49mm, arm length from center Y-maze = 13.5mm. **b** changes in the % of time walked
for all tested flies depending on environmental conditions. **c** Changes in the % of time walked grouped
by temperature, illumination, and arena. **d** Correlation of key behavioral parameters across
temperature, illumination, and arena type. For detailed fly numbers per experimental condition, see
supplemental Fig. S6.

Figure 5:

Fig. 5: A virtual flight simulator for studying individuality in flying flies. **a** Setup overview (225 x 225 x 600 mm). Flies are glued to a steel pin and placed in a magneto-tether surrounded by an LED matrix for stimulus presentation. **b** Close-up photo of a tethered flying fly. **c** Schematic setup overview. Flies are filmed from below through an IR-pass filter against near-infrared back illumination. A connected pump provides air puffs to reinitiate flight if necessary. We provide a mounting port for an optional IR-laser for heat punishment via a dichroic mirror. **d** A MATLAB script controls stimulus presentation (optomotor or Buridan stimulus) with an angular extent of 360° azimuthally and 100° vertically and provides simultaneous real-time tracking of flight-heading over 360° (**e**). Before an experiment, each fly's ability to rotate was tested using a 120°/s CW or CCW stripe rotation (optomotor stimulus), resulting in measured rotational flight movements close to the rotational velocity of the stripe pattern (**e**, upper left) and uniform distribution of angles (**e**, lower left). This was followed by each fly flying under the Buridan stimulus for 10min (**d**, right) and an analysis of heading choices (**e**, upper right) and their angular distribution (**e**, lower right).

Figure 6:

Fig. 6: Individual traits of flying flies persist over time but depend on the visual scenery. **a** Angular histograms show the persistence of individual heading choices over consecutive days and across different contrast ratios. **b,c** Correlation of key behavioral parameters over time (**b**) and depending on changes in visual contrast (**c**). For detailed fly numbers per experimental condition, see supplemental Fig. S10.

Figure 7:

Fig. 7: Visual attention persists across different behavioral states. **a** Experimental overview. Flies were
untethered after flying within the flight simulator for 10 min (left) and walking for 15 min (right) under
the same Buridan stimulus (stripe size=11°). Single fly data (middle) and population data (bottom)
reveal stripe fixation behavior when walking and during flight. **b** Correlation of key behavioral
parameters depending on the behavioral state (flight=x-axis, walking=y-axis). For detailed fly numbers
per experimental condition, see supplemental Fig. S12.

Figure 8:

672   **Fig. 8:** A hierarchy of environmental and behavioral contexts based on their influence on individuality.
673   **a** Overview summarizing correlation coefficients ($p < 0.05$) for key behavioral parameters depending on
674   behavioral context, time, temperature, and visual input. **b** Quantification of the influence of behavioral
675   and environmental contexts on the persistence of individuality.

Supplemental figure S1:

a

■ Bought
■ 3D printed

b

850nm LED

c Computer (Matlab)

"SingleBuridan Tracker.mat"

Before tracking
- Auto-Detect circular ROI - Compute maximum intensity projection

Loop for each frame

- Subtract background
- Threshold and track X/Y coordinates of fly
    >> If fly is lost: assume previous position
    >> If more than one fly: choose blob
        closest to fly position in previous frame

After recording
Post processing for all recorded frames: - Convert to mm - Calculate parameters - For each ROI: Save .txt with all data + save figure

 Supplemental figure legends

Supplemental Fig. S1: Construction of Buridan assays. **a** Two different types of Buridan assays
(fluorescent tubes left, flexible LED matrix right) and the parts necessary for construction. Part
numbers are referenced in supplemental table T2 and for 3D printed parts under
github.com/LinneweberLab/individuality-assays for replication. Black numbers indicate off-the-shelve
parts; blue numbers indicate 3D-printed parts. **b** construction of Buridan bottom plates. An 850nm LED
strip is wrapped around a translucent acrylic disk (11), providing homogeneous near-infrared
background illumination from below. **c** schematic illustration of the tracking procedure. The provided
MATLAB script automatically reads video frames, tracks the X- and Y-coordinates of individual flies
over time, and saves data for further analysis.

Supplemental Fig. S2: Population responses over time, varying stripe numbers and background
illumination. **a-c** Boxplots illustrating key behavioral parameters for the whole population (blue) and
separately for females/males (green/magenta) over time (**a**), varying stripe numbers (**b**) and
background illumination (**c**). Animal numbers are indicated next to each plot.

Supplemental figure S3:

Supplemental Fig. S3: Sex-specific differences in individual traits. **a-d** Correlation of key behavioral
parameters for females and males over time (**a**), different stripe numbers (**b**), changes in background
illumination (**c**), and depending on the usage of a clear dome around the arena (**d**). Magenta=females,
green=males. Correlation coefficients are displayed in the upper left corner of each plot. *=$p<0.05$,
**=$p<0.01$, ***=$p<0.001$, for detailed fly numbers per experimental condition, see supplemental Fig.
S2.

Supplemental Fig. S4: Construction of a high-throughput assay for studying the influence of
temperature, illumination, and arena shape on individuality. **a** Overview of parts necessary for
constructing the assay. Part numbers are referenced in supplemental table T2 and for 3D printed parts
under github.com/LinneweberLab/individuality-assays for replication. Black numbers indicate off-the-
shelve parts; blue numbers indicate 3D-printed parts. **b** Schematic illustration of the tracking
procedure. The provided MATLAB script automatically reads video frames, detects single arenas,
tracks the X- and Y-coordinates of individual flies over time, and saves data for further analysis. **c**
Exemplary temperature- and humidity curves were measured within the center arena during heat-up to
25 and 30°C, respectively. Stable plateaus are reached after ~20min. **d** The camera was calibrated
using the MATLAB calibration toolbox. The provided MATLAB script uses these camera parameters to
correct lens distortion artifacts during tracking. **e** Screenshots of the tracking procedure using the
multi-Buridan assay. Arenas are detected automatically, and individual flies get thresholded after
background subtraction, resulting in reliable trajectories of each individual fly. **f** Same as **e**, but for the
Y-maze assay. The box on the right shows an enlarged depiction of a single arena. **g** Same as **e**, but
for linear assays, demonstrating the usability of the setup with other arena shapes. The box on the
right shows an enlarged depiction of a single arena.

Supplemental figure S5:

Supplemental Fig. S5: Preliminary data of a fly population (TopBanana) illustrating orientational
responses. Flies walk between the 2 stripes and show randomly distributed headings in darkness,
independent of temperature. **a** Velocity- (left, transparency added to tracks), transition- (middle), and
angular plots (right, angles from moving periods only, non-axial) for vertical stripes positioned at 0°
(top), 90° (middle) or in complete darkness (bottom), respectively for all tested flies at 25°C. Scale bar
= 10mm. **b**) Same as **a**) but at 30°C. Fly numbers are indicated in the velocity plots.

Supplemental figure S6:

730   **Supplemental Fig. S6:** Sex- and genotype-specific population responses over varying temperature,
731   illumination and arena shape in walking flies. **a** Boxplots illustrating key behavioral parameters for the
732   whole tested CantonS population (left, blue=23°C, red=32°C) and separately for females/males
733   (green/magenta) over varying temperature, illumination and arena shape. Animal numbers are
734   indicated next to each plot. **b** Same as **a**, but for all tested TopBanana flies.

Supplemental figure S7:

Supplemental Fig. S7: Sex- and genotype-specific differences based on the influence of temperature,
illumination, and arena shape on the persistence of individual traits in walking flies. **a** Correlation of
key behavioral parameters for females and males depending on temperature, illumination, and arena.
Magenta=females, green=males. Correlation coefficients are displayed in the upper left corner of each
plot. *=$p<0.05$, **=$p<0.01$, ***=$p<0.001$, for detailed fly numbers per experimental condition, see
supplemental Fig. S6. **b** same as **a** but sorted by genotype (red=CantonS, blue=TopBanana).

Supplemental figure S8:

745   **Supplemental Fig. S8:** Complete correlation matrix across all behavioral parameters and contexts.

746   Correlation coefficients with p<0.05 are color coded.

Supplemental figure S9:

a

■ Bought
■ 3D printed

b

90 Hz real-time
video-stream

Loop for each
frame and each ROI

- Threshold fly at 2 gaussian filter setting to separate
  the body from the legs/wings
- Fit ellipse around fly body
- Calculate heading based on head-tail asymetry
- If fly stops flying (wings become visible in quadrants
  II and III) --> trigger air pump to reinitiate flight
- Save original camera frame to disk
- Display currect frame and plot heading in user
  window until tracking is done

After
recording

Post processing for all recorded frames:
- Downsample to 90hz
- Calculate parameters
- Save .txt with all data + save figure

c

II   I

III   IV

Supplemental Fig. S9: Construction of a virtual flight simulator for studying the persistence of
individual traits in flying flies. **a** Overview of parts necessary for constructing the assay. Part numbers
are referenced in supplemental table T2 and for 3D printed parts under
github.com/LinneweberLab/individuality-assays for replication. Black numbers indicate off-the-shelve
parts; blue numbers indicate 3D-printed parts. **b** schematic illustration of the tracking procedure. The
provided MATLAB script works in real-time at ~90Hz on a 4GHz processor. The code automatically
reads camera frames, computes the heading angle over time, and saves data for further analysis. **c**
Illustration of the tracking procedure. The camera frame is thresholded using 2 different Gaussian filter
settings to separately threshold the body (yellow) and legs/wings (dark blue). An ellipse is fitted around
the thresholded body pixels, and its main axis is used to measure the distribution of body pixels within
4 virtual quadrants whose origin is the centroid of the fly. The head region can be determined because
it appears smaller in the camera images than the abdominal region of the fly.

Supplemental figure S10: Population responses over time and different visual contrasts in flying flies.
Boxplots illustrating key behavioral parameters for the whole population (blue) and separately for
females/males (green/magenta) over time and varying stripe contrasts (**b**). Animal numbers are
indicated next to each plot.

Supplemental figure S11:

769 **Supplemental Fig. S11:** Sex-specific differences of individual traits in flying flies persist over time but
770 depend on the visual scenery. **a** Correlation of key behavioral parameters for females and males over
771 time and dependent on stimulus contrast. Magenta=females, green=males. Correlation coefficients are
772 displayed in the upper left corner of each plot. *=p<0.05, **=p<0.001, ***=p<0.0001, for detailed fly
773 numbers per experimental condition, see supplemental Fig. S10.

Supplemental figure S12:

776   **Supplemental Fig. S12:** Population responses of orientational parameters across different behavioral
777   states. Boxplots illustrating orientational parameters under different behavioral states for the whole test
778   population (blue) and separately for females/males (green/magenta). Animal numbers are indicated
779   next to each plot.

Supplemental figure S13:

782 **Supplemental Fig. S13:** Attention persists across different behavioral states in males but not females.
783 **a** Correlation of key behavioral parameters for females and males depending on the behavioral state.
784 Magenta=females, green=males. Correlation coefficients are displayed in the upper left corner of each
785 plot. *=$p<0.05$, **=$p<0.001$, ***=$p<0.0001$, for detailed fly numbers per experimental condition, see
786 supplemental Fig. S12.

Supplemental table T1: Description of automatically computed output parameters

Parameter	Description
roinr	Arena number for fly identification.
sex	Male = 0, female = 1.
eucldistsum	Total distance walked in mm.
perctimemoving	% of time spent walking.
numberpauses	Number of pauses (min pause length = 1s).
medianpausedurationsec	Median pause duration in s.
meanspeedmov	Mean walking speed in mm/s.
mediananglecentermoving	Median walking angle (heading direction) between 0° and 360° in the center region of the arena as seen from the animal.
mediananglecenteraxialmoving	Median walking angle (heading direction) between 0° and 180° (axial) in the center region of the arena as seen from the animal. Axial transformation of angular data was achieved by subtracting 180° from all angles larger than 180°. Axial data is more robust against cancelling-out of opposing angles in median calculations, e.g. when a fly is walking back and forth between stripes.
stddeviation	Standard deviation of angular data in ° as seen from the animal. Low angular standard deviation indicates a smaller range of chosen heading directions. Contains only data points where the animal was walking in the center region of the arena.
rlength	Resulting vector length of angular data mapped between 0° and 360°. A vector length of 0 would indicate uniform distribution of heading angles. A vector length of 1 would indicate the animl was only moving in a perfectly straight path. Contains only data points where the animal was walking in the center region of the arena.
rlengthax	Resulting vector length of angular data mapped between between 0° and 180° (axial) . A vector length of 0 would indicate uniform distribution of heading angles. A vector length of 1 would indicate the animl was only moving in a perfectly straight path. Contains only data points where the animal was walking in the center region of the arena. Axial data is more robust against cancelling-out of opposing angles in median calculations, e.g. when a fly is walking back and forth between stripes.
medianangvelmovalldegs	Median angular velocity (angular speed of azimuthal rotation) when walking in °/s. Median angular velocity is also a measure of handedness. Positive values indiceate right turns and negative values indicate left turns as seen from the animal.
medianangvelmovcenterdegs	Median angular velocity (angular speed of azimuthal rotation) when walking in the center region of the arena in °/s. Median angular velocity is also a measure of handedness. Positive values indiceate right turns and negative values indicate left turns as seen from the animal.
medianangvelmovouterdegs	Median angular velocity (angular speed of azimuthal rotation) when walking in the edge region of the arena in °/s. Median angular velocity is also a measure of handedness. Positive values indiceate right turns and negative values indicate left turns as seen from the animal.
absturninganglesum	Sum of all walking heading changes in °. Divided by distance walked this can be used as a measure for „meandering" (heading change per mm walked in °/mm).
absturninganglesumcenter	Sum of all walking heading changes within the center region of the arena in °. Divided by distance walked this can be used as a measure for „meandering" (heading change per mm walked in °/mm).
centrophobindex	Centrophobicity index between 0 and 1. 0 indicates an animal being in the center area 100% of the time. 1 indicates an animal being in the edge area 100% of the time.
pecflymorethanone	% of time with more than two objects detected per arena. This parameter indicates false animal detections (e.g dust) and and is useful for validating tracking quality or finding tracking errors.
percflylost	% of time the tracker lost an animal. This parameter is useful for validating tracking quality or finding tracking errors.
stripedeviationmedian	Applies to Buridan data. Medain angular distance to the closest (in terms of azimuthal rotation) stripe when walking in the center of the arena in °.
turnsnumber	Applies to Y-maze data. Number of turns. One turn is defined as an animal (measured at center point) walking from one Y-maze arm 3mm into another.
turnsbias	Ratio of right VS left turns. Values larger than 0.5 indicate right-handedness. Values smaller than 0.5 indicate left-handedness.

Supplemental table T2: Part list for setup replication

LED Buridan Setup

Nr	Description	Amount
1	Flir Firefly S FFY-U3-04S2M-C + Edmund optics UC Series Fixed Focal Length Objective (8mm, Stock #33-307)	1
2	Acrylic plate white opaque matt 500 x 410 x 3 mm (30mm circular hole in center)	1
3	Clear acrylic cylinder (300mm diameter, 300mm height) + diffuser paper or satinice acrylic cylinder (300mm diameter, 300mm height)	1
4	Fluorscent tubes (Osram, L 40w, 640C circular cool white) + Osram QT-M 1626–42 controller for AC flicker rate >1kHz	4
5	Aluminium profiles black (45 x 45 mm) P45N10LS www.aluprofile24.de:	
	400mm + thread M12 (min 100mm) profile	4
	480mm profile	2
	485mm profile	1
	+ Construction accessories www.aluprofile24.de:	
	Wi45x45 angle	10
	IWiN10 angle	20
	BuMuM8 nut screw	20
	FußD40DG m12x100 feet	4
	+ BitFenix Spectre PRO 230 mm mounted between aluminium rails on 2 opposing sides of the setup for cooling the LED matrix	8
6	6x RGB LED matrix (128 x 64 pixel) www.ledall.de 1x Hdmi controller Novastar mctrl 660 pro + 1x Novastar A8S receiving card 6x	1
	+ Famotec locating ring 3mm inner diameter (EAN: 4053199955619) + grub screw M3 x 20mm for fixing LED matrix to the LED cylinder	150
7	Transparent acrylic dome 120mm diameter (www.architekturbedarf.de) coated with Sigmacote on the inside	1
8	PLEXIGLAS® WH10 DC white snow satinice disk 119 x 119 x 10 mm + 850nm LED strip wrapped around edge for backlight illumintaion	1
9	Acrylic plate white opaque matt 500 x 500 x 3 mm (120mm circular hole in center)	1
10	PLEXIGLAS® WH10 DC white snow satinice 500 x 500 x 3 mm	1
11	PLEXIGLAS® WH10 DC white snow satinice disk 119 x 119 x 10 mm + 850nm LED strip (www.ledxess.de) wrapped around edge for backlight illumintaion	1
12	PLEXIGLAS® WH10 DC white snow satinice 500 x 500 x 10 mm (300mm circular hole in center)	2

IndyTrax Setup

Nr	Description	Amount
13	120x12mm EEC0381B1-000U-A99 Sunon	4
14	880nm infrared backlight https://www.amazon.de/LED-Lichttherapie-Dunkelrotes-Infrarot-LED-Lichttherapie-Panel-Hautverj%C3%BCngung-Druckentlastung/dp/B07ZQ	1
15	Server cabinet https://www.serverschrank24.de/15-he-serverschrank-wandgehause-mit-glastur-bxtxh-600-x-450-x-820-mm.html	1
16	Translucent acrylic cylinder https://www.kus-kunststofftechnik.de/plexiglas-rohr/plexiglasxt-satiniert-o-60-mm-laenge-waehlbar-von-100-bis-1400-mm/	25
16	Electroluminescent foil A7 white	50
16	Transparent dome 50mm https://www.dekoweltcreativ.de/acrylkugel-transparent-teilbar-stueck/p-3330.html	25
16	Plexiglas GS white 70% 3mm WH10 49mm diameter https://www.plattenzuschnitt24.de/Plexiglas-GS-weiss-WH10-GT-3mm.html	50
17	19" drawer https://www.it-budget.de/19-Fachboden-ausziehbar-Vollauszug-Tragkraft-30-kg-350-mm-schwarz	2
18	Camera Point Grey BFLY-PGE-23S6C-C mounted on Rack chassis 24563-194nVent – Schroff	1
18	Objective LM8HC3HOLE 1" 8mm/F1.4	1
19	BitFenix Spectre PRO 230 mm	2
20	Carbon heater https://thermowelt.de/Artikel/carbonheater1-1GN-220V.html	1
Controllers	Humidity controller TXG top-messtechnik.com TMT-HC-210 controlls 40x40mm Sunon EE40101S1-1000U-999 Axiallüfter 12 V/DC placed on (4)	1
	PID temperature controller pohltechnic.com RT4-121-Tr21Sd controlls (20)	1
	VOLTCRAFT LSP-1165 for controlling IR LED panel (14) and fans (18, 19)	2
	TT COMMANDER FT for cooling down enclosure, if necessary (controlls enclosure fans (13))	1
	Inverter https://www.el-light.shop/epages/17778431.sf/de_DE/?ObjectPath=/Shops/17778431/Products/InF1S150 controlls el-foil (16) brightness	1

LED flight simulator

Nr	Description	Amount
21	Aluminium profiles black (30 x 30 mm) P30N8 www.aluprofile24.de:	
	600mm profile	4
	165mm profile	10
	200mm profile	2
	+ Construction accessories www.aluprofile24.de:	
	PADK30N8B cap	4
	IWIN8B angle	25
	BuMuM6 nut screw	10
	WI30N8B	10
	HaKoM6 (12mm)	20
22	optional servo motor for dorsal stimulus presentation (Dynamixel MX-12W) + ball bearing 6811-ZZ (www.kugellager-express.de)	1
23	4x RGB LED matrix (128 x 64 pixel) www.ledall.de + 1x Hdmi controller Novastar mctrl 660 pro + 1x Novastar A8S receiving card	1
	+ Famotec locating ring 3mm inner diameter (EAN: 4053199955619) + grub screw M3 x 20mm for fixing LED matrix to the LED cylinder	150
24	Neodynium cylinder magnet (5mm diameter, 20mm height) + white plastic cap + 1mm diameter v-shaped sapphire bearing in center for friction reduction	1
25	Neodynium ring magnet (10mm diameter, 5mm height, inner diameter 5mm) + white plastic cap	1
26	850nm near infrared LED array (https://www.amazon.de/POFET-Infrarot-IR-LED-850-nm-Platinenmodul-f%C3%BCr-das-Kamera%C3%Bcberwachungssystem)	1
27	Flir Firefly S FFY-U3-04S2M-C + Edmund optics UC Series Fixed Focal Length Objective (12mm, Stock #33-307)	1
Air puffs	Arduino UNO Rev3 + relais module + Schego membrane pump 300l/h	1

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