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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 consistent over long time periods. The consistency of individually variable behaviors over time is often referred to as an animal's individuality, or if consisting of multiple traits 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 consistency 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, consistency of behavioral individuality always persisted in at least one of the traits. Furthermore, our quantification reveals a hierarchical order of environmental features influencing individuality. We confirmed this hierarchy using a generalized linear model. 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 consistency of these interindividual differences allows for defining these idiosyncrasies as animal individuality, also called animal temperament, behavioral syndrome, or if the entirety of trait differences between individuals is meant even animal personality (Andrew Sih et al., 2004; Dall et al., 2004; Gosling, 2008; Reale et al., 2007). Animal individuality has been described for numerous species and behaviors, including: mouse exploration behavior (Freund et al., 2013), great tit feeding preferences (Partridge, 1976), octopus threat responses and feeding behavior (Mather & Anderson, 1993), pea aphid threat responses (Schuett et al., 2011), as well as vinegar fly handedness (Ayroles et al., 2015; Buchanan et al., 2015), olfaction (Honegger et al., 2020), phototaxis (Kain et al., 2012), and object orientation (Linneweber et al., 2020). The temporal consistency of individuality allows for discriminating it from short-term behavioral changes, like internal state variations. Furthermore, temporal consistency of behavior allows modifications by learning (Kahsai & Zars, 2011) and behavior-based natural selection (Lapiedra et al., 2018). Historically, the 'nature vs. nurture' debate (i.e., genome vs. environment) has shaped our understanding of the origins of behavioral individuality (Galton, 1875). Stochastic developmental processes are a more recently found factor (Honegger & de Bivort, 2018; Linneweber et al., 2020), providing an additional layer to gene-environmental interactions (Traynor & Singleton, 2010). 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 (Reale et al., 2007), and the definition of human personality entails behavioral consistency across situations (Mischel, 1968). Similar to humans, animals will never re-encounter the exact same situation under natural conditions, and animal personality can be an important factor in determining the ecological success of an animal and, as a consequence, its entire population (Takola & Schielzeth, 2022). It is, therefore, crucial to understand how an animal's personality or individuality in specific traits 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 (Briffa et al., 2008;

Herborn et al., 2010; Kralj-Fišer et al., 2007; Mowles et al., 2012) and against (Coleman &

Wilson, 1998; D'Eath & Burn, 2002) individuality in animals across situations with many finding

mixed results (Boissy & Bouissou, 1995; Spoolder et al., 1996; Werkhoven et al., 2021). Here

we sought to systematically investigate the consistency of individual behavioral differences

across situations using laboratory-controlled variable environmental conditions.

There is ample evidence for the consistency of human personality and animal

individuality over time (Costa & McCrae, 1988; Weisbuch et al., 2010). Intuitively, humans

attribute consistent personalities to other humans (Ross, 1977). The consistency of human

behavior across situations, including environmental context, nevertheless remains highly

debated, with several studies arguing for (Costa & McCrae, 1988; Fleeson, 2001) and against

it (Hartshorne & May, 1930; Mischel, 1968). 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

(Mischel, 1968). Animal models for behavioral consistency 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

consistency 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 (Dorkenwald et al., 2024; Xu et al., 2020; Zheng et

al., 2018), an extensive collection of cell type-specific, binary expression systems (Luan,

Peabody, et al., 2006; Pfeiffer et al., 2010), and a similarly extensive collection of tools for

manipulating neuronal function, including neuronal silencers (Baines et al., 2001; Sweeney et

al., 1995), activators (Luan, Lemon, et al., 2006), and thermogenetic (Hamada et al., 2008)

and optogenetic (Klapoetke et al., 2014; Pulver et al., 2009) 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 (Bülthoff et al., 1982; Götz, 1980), arenas for tethered walking (Seelig et

al., 2010) or virtual flight (Mathejczyk & Wernet, 2020) have been used to study behavioral

individuality in vison-based tasks (Linneweber et al., 2020; Mathejczyk & Wernet, 2019). 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) (Hassan et al., 2000), also known as Lobula Columnar Neuron 14 (LC14)

(Otsuna & Ito, 2006), Buridan's paradigm has been used to show the importance of the central

complex for normal walking speed (Strauss et al., 1992) and for studying photoreceptor

function or development (Kiral et al., 2020; Strauss et al., 2001), and visual memory (Neuser

et al., 2008). Other circular arenas have been frequently used to study locomotory activity

without visual stimuli (Branson et al., 2009; Robie et al., 2017; Valente et al., 2007).

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 (Hirsch & Boudreau, 1958),

geotaxis (Hirsch, 1959), olfactory learning (Jiang et al., 2016; Quinn et al., 1974), and circadian

rhythms (Konopka & Benzer, 1971). The Ethoscope (Geissmann et al., 2017) and MARGO

(Werkhoven et al., 2019) 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 (Ayroles et al., 2015; Buchanan et al., 2015; Kain et al., 2012).

To investigate the influence of the environment on the consistency 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 (Buchanan et al., 2015), and a simple setup to quantify locomotor activity in large groups of flies (Branson et al., 2009). We used this tracker to determine the relative contribution of various environmental features (temperature, visual cues, arena shape) to the consistency 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 (Costa & McCrae, 1988; Fleeson, 2001). 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 consistency across situations in human subjects (Mischel, 1968).

Materials and Methods

<u>Animals</u>

All experiments were performed using one week old wild-type *Drosophila melanogaster*

(CantonS (Source: Bassem Hassan) and TopBanana (Warren et al., 2018)) 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

<u>Assay</u>

Two Buridan setups were used: type A uses fluorescent tubes for background

illumination (as described in (Colomb et al., 2012; Linneweber et al., 2020)). 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.

<u>Automated tracking procedure</u>

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

(Colomb et al., 2012)) or offline using a custom-written MATLAB code (type B, available for

download from github.com/LinneweberLab/Mathejczyk_2024_eLife_Individuality, 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.

<u>Data output</u>

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 (Colomb et al., 2012). 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 (Colomb et al., 2012).

Setup 2: IndyTrax multi-arena platform

<u>Assay</u>

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 (Colomb et al., 2012), 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/Mathejczyk_2024_eLife_Individuality. 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/Mathejczyk_2024_eLife_Individuality. 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€.

<u>Temperature- and humidity-controlled enclosure</u>

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 (Colomb et al., 2012) 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 (Werkhoven et al., 2019). 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).

<u>Automated tracking procedure</u>

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 (Mathejczyk & Wernet, 2020)), 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/Mathejczyk_2024_eLife_Individuality). 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).

Automated tracking procedure

The code for tracking a fly's heading in 360° can be downloaded from github.com/LinneweberLab/Mathejczyk_2024_eLife_Individuality. 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 and statistical analysis for all experiments</u>

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

algorithm presented in (Berens, 2009). 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. Flies that did

not move during the experiment were excluded from the analysis.

Generalized Linear Model (GLM)

To summarize and generalize the findings from all experiments in this study (using individual

rank differences upon changing one or more environmental parameters), a Generalized Linear

Model (GLM) was trained using the environmental parameters as predictors (0 when

environmental parameter was not changed and 1 if it was changed for any given experimental

group) and the resulting individual rank differences as the response variable. We employed

ridge regression to mitigate collinearity among predictors and to control for overfitting. A range

of regularization parameters (λ) was tested using 5-fold cross-validation to identify the optimal

value that minimized the mean squared error (MSE). Z-scoring was applied to the predictor

variables prior to model fitting for data normalization. To estimate coefficient variability and

assess significance, bootstrapping was applied. For each of the 1,000 bootstrap iterations,

predictors and response data were resampled with replacement, and the ridge regression

model was refit using the optimal λ. Coefficients and intercept values from each bootstrap

sample were stored, and their distributions were used to calculate the mean coefficients,

standard errors (SEs) and p-values for coefficients based on t-statistics derived from

bootstrapped SEs.

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 (Linneweber et al., 2020), we started our analysis in the multiparametric two-stripe Buridan's paradigm (Fig. 1a, supplemental Fig. S1 for technical details) (Colomb et al., 2012; Linneweber et al., 2020). As expected, we measured interindividually variable, yet remarkably consistent behavioral responses, both on the individual (Fig. 1b, Fig. 2a, supplemental Fig. S3a) and mean group level (average of all individuals in the same situation, 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 in three behavioral traits: exploration (% of time walked, walking speed), attention (vector strength, angular velocity), and anxiety (centrophobicity). The exploration trait is characterized by parameters describing activity. Parameters related to attention reflect heightened responses to visual cues, but unlike commonly used metrics such as angle or stripe deviations (Colomb et al., 2012; Linneweber et al., 2020), they can be calculated in the absence of visual cues. Lastly, the parameter centrophobicity, used as an indicator of the trait anxiety, is linked to the open-field test in mice, where the ratio of wall-to-open-field activity is frequently calculated as a measurement of anxiety (Carter & Shieh, 2015).

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 means of three parameters (% of time walked, vector strength, and centrophobicity) were strongly affected by the stimulus modification. Surprisingly, despite this change in group means, we observed only a minor effect on individuality when testing the same fly in different situations (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 consistency 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 group responses.

To further characterize the surprising relationship between individuality and group

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

group 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 consistency, 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, albeit presenting the same

stimulus (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 (Gosling, 2008) of visual orientation

behavior is more strongly influenced by the arena type (despite producing very similar group

responses) and, only to a lesser degree, dependent on the visual cue (despite strong

differences in the group 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 group mean level. As

expected, most animals tested were more explorative at higher than at lower temperatures

((reviewed in: Gibert et al., 2016)) (supplemental Fig. S6). Despite these group changes,

individuals kept their rank within the group 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 in 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 consistency of idiosyncrasies remained high, while

conversely, other environmental situations like the arena type had little impact on the mean

but a massive effect on individual behavioral consistency (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 (Han et al., 2021). 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 consistency 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 towards a visual object 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).

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 consistency 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 group mean responses, a

difference that can be explained by scaled individual responses in which individuals keep the

rank in the group. 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.

To verify the results of our multiple correlative analyses, we took advantage of a

generalized linear model (GLM, Fig. 9). We first analyzed the effect of time, temperature,

vision, and arena on the mean group responses analogous to supplemental Fig. S2, S6, S10,

S12, (Fig. 9a). Our reanalysis of the walking data confirmed the temperature dependency of

activity parameters (Fig. S6). It also revealed effects on the group means by vision and arena

particularly for walking speed. The orientation parameter vector strength was similarly affected

by temperature vision and arena (see also supplemental Fig. S2, S6), while the position

parameter centrophobicity was strongest affected by the arena (see also supplemental Fig.

S2, S6). For flight parameters, the reanalysis of mean data revealed a strong effect of vision

on the orientation parameter vector strength and the heading parameter median heading axial

(see also supplemental Fig. S10). Our reanalysis, therefore, confirmed our previous analysis

of mean group data.

The next step of our reanalysis was to compare the rank changes of our flies across

situations analogous to the correlation analysis we had done (Fig. 2, 4, 6, 7). Our reanalysis

revealed significant changes in activity parameters and the orientation parameter vector

strength. In these parameters, the effect on the rank of each individual increases from time

over temperature, vision, and arena to the behavioral state. Confirming the data we found in

Fig. 2, 4, 6, 7, and 8. Also, the lack of a significant effect of the behavioral state for angular

velocity confirms our correlation analysis of Fig. 7 and 8.

In the last step of our reanalysis, we generated a GLM with ridge regression to show

the effect of changing environmental situations on individual rank differences (Fig, 9c). For the

activity parameters, time, temperature, and vision significantly stabilized individual

phenotypical rank. In contrast, the arena and behavioral state significantly had a negative effect

on rank stability. For the orientation parameter vector strength, we could fully replicate our

previous findings that time had the strongest positive effect on consistency, followed by

temperature. Vision had no significant impact. In contrast, the arena had a small negative effect

consistency, while the behavioral state had a strong negative effect on behavioral consistency.

In the second orientation parameter (angular velocity), a significant destabilizing effect was

only found for the behavioral state. and for the positional parameter (centrophobicity), only

vision had a significant effect on behavioral consistency.

In summary, using two independent mathematical approaches, we have established a

hierarchy of different environmental situations on behavioral consistency. This data

demonstrates that temporal consistency is also in animal most prevalent. Like humans also

animals have significant consistency across situation, but similar to humans different situations

evoke different behavioral responses masking measurable behavioral consistency.

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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' (Mischel, 1968). 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 (Gosling, 2008).

The contrary is true for mainly field and observation-based sciences such as ethology and

behavioral ecology that have since decades acquired observations for (Boissy & Bouissou,

1995; Spoolder et al., 1996) or against (Coleman & Wilson, 1998; D'Eath & Burn, 2002)

behavioral consistency across situations, with some of the most convincing work probably

found in hermit crabs (Briffa et al., 2008; Mowles et al., 2012), blue tits (Herborn et al., 2010),

and geese (Kralj-Fišer et al., 2007). 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) (Vansteelandt & Van Mechelen, 2004).

Our results confirm that under the same environmental situation and stimulus, animals

produce inter-individually variable, but extremely consistent behavioral responses when being

retested (Bell et al., 2009). Therefore, consistent 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 (Linneweber et al., 2020)), contrasting 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 (Buchanan et al., 2015; Linneweber

et al., 2020).

The variability of the situational environmental context makes individual animal

behavior in the wild much less predictable than that under laboratory conditions (Briffa &

Greenaway, 2011). In opposition to the vast human literature on the 'person-situation debate'

(Diener & Larsen, 1984; Fleeson, 2001; Mischel, 1968; Vansteelandt & Van Mechelen, 2004;

Weisbuch et al., 2010), the importance of environmental context in shaping individual animal

behavior remains descriptive (Kralj-Fišer et al., 2007) and experimentally largely unexplored,

with few exceptions (Briffa et al., 2008; Mowles et al., 2012; Versace et al., 2020; Werkhoven

et al., 2021). 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 (Mischel, 1968), 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 consistency of individual

traits measured by behavioral parameters 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 (Mischel &

Shoda, 1995; Mischel & Shoda, 1998). 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 brain properties computing the

response remain the same. In line with lower but measurable behavioral consistency across

situations , a vast literature demonstrates the importance of animal individual differences for

ecological communities under natural conditions (reviewed in: Sih et al., 2012; Takola &

Schielzeth, 2022).

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

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

time frame of days had little effect on the behavioral outcome (Bell et al., 2009; Buchanan et

al., 2015; Linneweber et al., 2020; Mischel, 1968; Pantoja et al., 2016; Ross, 1977; Schuett et al., 2011). 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 (Mischel, 1973). 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 (Crabbe et al., 1999; Lithgow et al., 2017). 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

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

684   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

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

Figure 2:

714 **Fig. 2:** Individual traits of walking flies persist over time but depend on the visual scenery. **a-c**
715 Correlation of key behavioral parameters over time (n=60) (**a**), different stripe numbers (n=57) (**b**), and
716 changes in background illumination (n=54) (**c**). Correlation coefficients are displayed in the upper left
717 corner of each plot. *=p<0.05, **=p<0.01, ***=p<0.001, for detailed fly numbers per experimental
718 condition, see supplemental Fig. S2.

Figure 3:

a

IR/white backlight

Humidifier

Camera

Fan

Door

Heating plate

b

Acrylic cylinder

El-foil

Clear dome

Acrylic plate

c

Paper

Glass

Y-maze plates

Paper

Glass

d

Buridan

Y-maze

mm/s

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**). See Materials and Methods:Setup 2: IndyTrax multi-arena platform for details.

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
and vector strength for all tested flies depending on environmental conditions (n=400). Right: data
grouped by temperature, illumination, and arena. **c** Correlation of key behavioral parameters across
temperature, illumination, and arena type (n=400). For detailed fly numbers per experimental
condition, see supplemental Fig. S6.

Figure 5:

a
LED Matrix
Fly
Aluminium beams

b
Magnet
Sapphire bearing
Steel pin
Magnet

c
Magnet
Saphire bearing
850nm LED
850nm LED
Dichroic mirror
980nm Laser
Air Pump
IR-pass filter
Camera
Input
LED controller (via HDMI)
Output
Real-time tracking
Computer (Matlab)

d
Optomotor stimulus
Buridan stimulus

e

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:

754 **Fig. 6:** Individual traits of flying flies persist over time but depend on the visual scenery. **a** Angular
755 histograms show the persistence of individual heading choices over consecutive days and across
756 different contrast ratios. **b,c** Correlation of key behavioral parameters over time (n=30) (**b**) and
757 depending on changes in visual contrast (n=30) (**c**).

Figure 7:

a

10 min   15 min

Flight   Walking

Fly 1

Fly 2

All flies
n=73

b

Walking

Flight

Exploration

% of time walked Walking   0.031
Flight
# of pauses

Walking speed (mm/s) Walking   -0.061
Flight
Abs ang velocity (°/min)

Attention

Vector strength Walking   0.005
Flight
Vector strength

Angular velocity (°/s) Walking   0.316**
Flight
Angular velocity (°/s)

760   **Fig. 7:** Visual attention persists across different behavioral states. **a** Experimental overview. Flies were
761   untethered after flying within the flight simulator for 10 min (left) and walking for 15 min (right) under
762   the same Buridan stimulus (stripe size=11°). Single fly data (middle) and group data (bottom) reveal
763   stripe fixation behavior when walking and during flight. **b** Correlation of key behavioral parameters
764   depending on the behavioral state (flight=x-axis, walking=y-axis, n=73).

Figure 8:

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

Absolute individual difference walking

Absolute individual difference flight

Absolute individual rank difference walking + flight

Regression model individual rank difference walking + flight

Fig. 9: A generalized linear model to demonstrate individuality across situations. **a** The difference of
individual behavioral parameters for walking (left) and flight (right) when changing environmental
conditions (time, temperature, vision, arena). Significance: *p<0.05, **p<0.01, ***p<0.001. **b** The
difference of rank (in %) within the tested groups (walking and flight combined) when changing
environmental conditions. Smaller rank differences indicate a lower influence of the environmental
parameter on the persistence of individuality. Significance: *p<0.05, **p<0.01, ***p<0.001. **c**
Generalized linear model (ridge regression) showing the influence of changing environmental
parameters on individual rank differences within the tested group (individuality) using the data in **b**.
Significance: *p<0.05, **p<0.01, ***p<0.001. Non-significant environmental parameters are shaded in
grey.

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/Mathejczyk_2024_eLife_Individuality 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: Group responses over time, varying stripe numbers and background
illumination. **a-c** Boxplots illustrating key behavioral parameters for the whole group (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 figure S4:

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/Mathejczyk_2024_eLife_Individuality 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 group (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 Fig. S6: Sex- and genotype-specific group responses over varying temperature,
illumination and arena shape in walking flies. **a** Boxplots illustrating key behavioral parameters for the
whole tested CantonS group (left, blue=23°C, red=32°C) and separately for females/males
(green/magenta) over varying temperature, illumination and arena shape. Animal numbers are
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:

852   **Supplemental Fig. S8:** Complete correlation matrix across all behavioral parameters and contexts.
853   Correlation coefficients with p<0.05 are color coded. A high resolution version of the data can be found
854   under https://github.com/LinneweberLab/Mathejczyk_2024_eLife_Individuality/blob/main/S8.pdf.

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/Mathejczyk_2024_eLife_Individualityfor 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:

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

Supplemental figure S11:

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

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

Supplemental figure S13:

a

Behavioral state

891   **Supplemental Fig. S13:** Attention persists across different behavioral states in males but not females.
892   **a** Correlation of key behavioral parameters for females and males depending on the behavioral state.
893   Magenta=females, green=males. Correlation coefficients are displayed in the upper left corner of each
894   plot. *=p<0.05, **=p<0.001, ***=p<0.0001, for detailed fly numbers per experimental condition, see
895   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 dat a 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

References

Andrew Sih, Alison M Bell, J Chadwick Johnson, & Robert E Ziemba. (2004). Behavioral Syndromes: An Integrative Overview. *The Quarterly Review of Biology*, *79*(3), 241-277. https://doi.org/10.1086/422893

Ayroles, J. F., Buchanan, S. M., O'Leary, C., Skutt-Kakaria, K., Grenier, J. K., Clark, A. G., Hartl, D. L., & de Bivort, B. L. (2015). Behavioral idiosyncrasy reveals genetic control of phenotypic variability. *Proceedings of the National Academy of Sciences*, *112*(21), 6706-6711. https://doi.org/doi:10.1073/pnas.1503830112

Baines, R. A., Uhler, J. P., Thompson, A., Sweeney, S. T., & Bate, M. (2001). Altered Electrical Properties in DrosophilaNeurons Developing without Synaptic Transmission. *The Journal of Neuroscience*, *21*(5), 1523-1531. https://doi.org/10.1523/jneurosci.21-05-01523.2001

Bell, A. M., Hankison, S. J., & Laskowski, K. L. (2009). The repeatability of behaviour: a meta-analysis. *Animal Behaviour*, *77*(4), 771-783. https://doi.org/https://doi.org/10.1016/j.anbehav.2008.12.022

Berens, P. (2009). CircStat: A MATLAB Toolbox for Circular Statistics. *2009*, *31*(10), 21. https://doi.org/10.18637/jss.v031.i10

Boissy, A., & Bouissou, M. F. (1995). Assessment of individual differences in behavioural reactions of heifers exposed to various fear-eliciting situations. *Applied Animal Behaviour Science*, *46*(1), 17-31. https://doi.org/https://doi.org/10.1016/0168-1591(95)00633-8

Branson, K., Robie, A. A., Bender, J., Perona, P., & Dickinson, M. H. (2009). High-throughput ethomics in large groups of Drosophila. *Nat Methods*, *6*(6), 451-457. https://doi.org/10.1038/nmeth.1328

Briffa, M., & Greenaway, J. (2011). High In Situ Repeatability of Behaviour Indicates Animal Personality in the Beadlet Anemone Actinia equina (Cnidaria). *PLoS One*, *6*(7), e21963. https://doi.org/10.1371/journal.pone.0021963

Briffa, M., Rundle, S. D., & Fryer, A. (2008). Comparing the strength of behavioural plasticity and consistency across situations: animal personalities in the hermit crab Pagurus bernhardus. *Proc Biol Sci*, *275*(1640), 1305-1311. https://doi.org/10.1098/rspb.2008.0025

Buchanan, S. M., Kain, J. S., & de Bivort, B. L. (2015). Neuronal control of locomotor handedness in Drosophila. *Proc Natl Acad Sci U S A*, *112*(21), 6700-6705. https://doi.org/10.1073/pnas.1500804112

Bülthoff, H., Götz, K. G., & Herre, M. (1982). Recurrent inversion of visual orientation in the walking fly,Drosophila melanogaster. *Journal of comparative physiology*, *148*(4), 471-481. https://doi.org/10.1007/BF00619785

Carter, M., & Shieh, J. (2015). Chapter 2 - Animal Behavior. In M. Carter & J. Shieh (Eds.), *Guide to Research Techniques in Neuroscience (Second Edition)* (pp. 39-71). Academic Press. https://doi.org/https://doi.org/10.1016/B978-0-12-800511-8.00002-2

Coleman, K., & Wilson, D. S. (1998). Shyness and boldness in pumpkinseed sunfish: individual differences are context-specific. *Animal Behaviour*, *56*(4), 927-936. https://doi.org/https://doi.org/10.1006/anbe.1998.0852

Colomb, J., Reiter, L., Blaszkiewicz, J., Wessnitzer, J., & Brembs, B. (2012). Open source tracking and analysis of adult Drosophila locomotion in Buridan's paradigm with and without visual targets. *PLoS One*, *7*(8), e42247. https://doi.org/10.1371/journal.pone.0042247

Costa, P. T., Jr., & McCrae, R. R. (1988). Personality in adulthood: a six-year longitudinal study of self-reports and spouse ratings on the NEO Personality Inventory. *J Pers Soc Psychol*, *54*(5), 853-863. https://doi.org/10.1037//0022-3514.54.5.853

Crabbe, J. C., Wahlsten, D., & Dudek, B. C. (1999). Genetics of Mouse Behavior: Interactions with Laboratory Environment. *Science*, *284*(5420), 1670-1672. https://doi.org/doi:10.1126/science.284.5420.1670

D'Eath, R., & Burn, C. (2002). INDIVIDUAL DIFFERENCES IN BEHAVIOUR: A TEST OF 'COPING STYLE' DOES NOT PREDICT RESIDENT-INTRUDER AGGRESSIVENESS IN PIGS. *Behaviour*, *139*(9), 1175-1194. https://doi.org/https://doi.org/10.1163/15685390260437326

Dall, S. R. X., Houston, A. I., & McNamara, J. M. (2004). The behavioural ecology of personality: consistent individual differences from an adaptive perspective. *Ecology Letters*, *7*(8), 734-739. https://doi.org/https://doi.org/10.1111/j.1461-0248.2004.00618.x

Diener, E., & Larsen, R. J. (1984). Temporal stability and cross-situational consistency of affective, behavioral, and cognitive responses. *J Pers Soc Psychol*, *47*(4), 871-883. https://doi.org/10.1037//0022-3514.47.4.871

Dorkenwald, S., Matsliah, A., Sterling, A. R., Schlegel, P., Yu, S.-c., McKellar, C. E., Lin, A., Costa, M., Eichler, K., Yin, Y., Silversmith, W., Schneider-Mizell, C., Jordan, C. S., Brittain, D., Halageri, A., Kuehner, K., Ogedengbe, O., Morey, R., Gager, J.,…The FlyWire, C. (2024). Neuronal wiring diagram of an adult brain. *Nature*, *634*(8032), 124-138. https://doi.org/10.1038/s41586-024-07558-y

Fleeson, W. (2001). Toward a structure- and process-integrated view of personality: traits as density distribution of states. *J Pers Soc Psychol*, *80*(6), 1011-1027. https://www.ncbi.nlm.nih.gov/pubmed/11414368

Freund, J., Brandmaier, A. M., Lewejohann, L., Kirste, I., Kritzler, M., Krüger, A., Sachser, N., Lindenberger, U., & Kempermann, G. (2013). Emergence of Individuality in Genetically Identical Mice. *Science*, *340*(6133), 756-759. https://doi.org/doi:10.1126/science.1235294

Galton, F. (1875). The history of twins, as a criterion of the relative powers of nature and nurture. *Fraser's Magazine*, *12*(71), 566-576.

Geissmann, Q., Garcia Rodriguez, L., Beckwith, E. J., French, A. S., Jamasb, A. R., & Gilestro, G. F. (2017). Ethoscopes: An open platform for high-throughput ethomics. *PLOS Biology*, *15*(10), e2003026. https://doi.org/10.1371/journal.pbio.2003026

Gibert, J. P., Chelini, M.-C., Rosenthal, M. F., & DeLong, J. P. (2016). Crossing regimes of temperature dependence in animal movement. *Global Change Biology*, *22*(5), 1722-1736. https://doi.org/https://doi.org/10.1111/gcb.13245

Gosling, S. D. (2008). Personality in Non-human Animals. *Social and Personality Psychology Compass*, *2*(2), 985-1001. https://doi.org/https://doi.org/10.1111/j.1751-9004.2008.00087.x

Götz, K. G. (1980). Visual Guidance in Drosophila. In O. Siddiqi, P. Babu, L. M. Hall, & J. C. Hall (Eds.), *Development and Neurobiology of Drosophila* (pp. 391-407). Springer US. https://doi.org/10.1007/978-1-4684-7968-3_28

Hamada, F. N., Rosenzweig, M., Kang, K., Pulver, S. R., Ghezzi, A., Jegla, T. J., & Garrity, P. A. (2008). An internal thermal sensor controlling temperature preference in Drosophila. *Nature*, *454*(7201), 217-220. https://doi.org/10.1038/nature07001

Han, R., Wei, T.-M., Tseng, S.-C., & Lo, C.-C. (2021). Characterizing approach behavior of Drosophila melanogaster in Buridan's paradigm. *PLoS One*, *16*(1), e0245990. https://doi.org/10.1371/journal.pone.0245990

Hartshorne, H., & May, M. A. (1930). *Studies in the Nature of Character: Studies in deceit*. McMillian. https://books.google.de/books?id=cZi7xwEACAAJ

Hassan, B. A., Bermingham, N. A., He, Y., Sun, Y., Jan, Y. N., Zoghbi, H. Y., & Bellen, H. J. (2000). atonal regulates neurite arborization but does not act as a proneural gene in the Drosophila brain. *Neuron*, *25*(3), 549-561. https://doi.org/10.1016/s0896-6273(00)81059-4

Herborn, K. A., Macleod, R., Miles, W. T. S., Schofield, A. N. B., Alexander, L., & Arnold, K. E. (2010). Personality in captivity reflects personality in the wild. *Animal Behaviour*, *79*(4), 835-843. https://doi.org/10.1016/j.anbehav.2009.12.026

Hirsch, J. (1959). Studies in experimental behavior genetics. II. Individual differences in geotaxis as a function of chromosome variations in synthesized Drosophila populations. *J Comp Physiol Psychol*, *52*(3), 304-308. https://doi.org/10.1037/h0043498

Hirsch, J., & Boudreau, J. C. (1958). Studies in experimental behavior genetics. I. The heritability of phototaxis in a population of Drosophila melanogaster. *J Comp Physiol Psychol*, *51*(6), 647-651. https://doi.org/10.1037/h0039498

Honegger, K., & de Bivort, B. (2018). Stochasticity, individuality and behavior. *Curr Biol*, *28*(1), R8-R12. https://doi.org/10.1016/j.cub.2017.11.058

Honegger, K. S., Smith, M. A.-Y., Churgin, M. A., Turner, G. C., & de Bivort, B. L. (2020). Idiosyncratic neural coding and neuromodulation of olfactory individuality in <i>Drosophila</i>. *Proceedings of the National Academy of Sciences*, *117*(38), 23292-23297. https://doi.org/doi:10.1073/pnas.1901623116

Jiang, H., Hanna, E., Gatto, C. L., Page, T. L., Bhuva, B., & Broadie, K. (2016). A fully automated Drosophila olfactory classical conditioning and testing system for behavioral learning and memory assessment. *Journal of Neuroscience Methods*, *261*, 62-74. https://doi.org/https://doi.org/10.1016/j.jneumeth.2015.11.030

Kahsai, L., & Zars, T. (2011). Learning and Memory in Drosophila: Behavior, Genetics, and Neural Systems. In N. Atkinson (Ed.), *International Review of Neurobiology* (Vol. 99, pp. 139-167). Academic Press. https://doi.org/https://doi.org/10.1016/B978-0-12-387003-2.00006-9

Kain, J. S., Stokes, C., & de Bivort, B. L. (2012). Phototactic personality in fruit flies and its suppression by serotonin and <i>white</i>. *Proceedings of the National Academy of Sciences*, *109*(48), 19834-19839. https://doi.org/doi:10.1073/pnas.1211988109

Kiral, F. R., Linneweber, G. A., Mathejczyk, T., Georgiev, S. V., Wernet, M. F., Hassan, B. A., von Kleist, M., & Hiesinger, P. R. (2020). Autophagy-dependent filopodial kinetics restrict synaptic partner choice during Drosophila brain wiring. *Nature Communications*, *11*(1), 1325. https://doi.org/10.1038/s41467-020-14781-4

Klapoetke, N. C., Murata, Y., Kim, S. S., Pulver, S. R., Birdsey-Benson, A., Cho, Y. K., Morimoto, T. K., Chuong, A. S., Carpenter, E. J., Tian, Z., Wang, J., Xie, Y., Yan, Z., Zhang, Y., Chow, B. Y., Surek, B., Melkonian, M., Jayaraman, V., Constantine-Paton, M.,…Boyden, E. S. (2014). Independent optical excitation of distinct neural populations. *Nat Methods*, *11*(3), 338-346. https://doi.org/10.1038/nmeth.2836

Konopka, R. J., & Benzer, S. (1971). Clock Mutants of Drosophila melanogaster. *Proceedings of the National Academy of Sciences*, *68*(9), 2112-2116. https://doi.org/10.1073/pnas.68.9.2112

Kralj-Fišer, S., Scheiber, I. B. R., Blejec, A., Moestl, E., & Kotrschal, K. (2007). Individualities in a flock of free-roaming greylag geese: Behavioral and physiological consistency over time and across situations. *Hormones and Behavior*, *51*(2), 239-248. https://doi.org/https://doi.org/10.1016/j.yhbeh.2006.10.006

Lapiedra, O., Schoener, T. W., Leal, M., Losos, J. B., & Kolbe, J. J. (2018). Predator-driven natural selection on risk-taking behavior in anole lizards. *Science*, *360*(6392), 1017-1020. https://doi.org/doi:10.1126/science.aap9289

Linneweber, G. A., Andriatsilavo, M., Dutta, S. B., Bengochea, M., Hellbruegge, L., Liu, G., Ejsmont, R. K., Straw, A. D., Wernet, M., Hiesinger, P. R., & Hassan, B. A. (2020). A neurodevelopmental origin of behavioral individuality in the Drosophila visual system. *Science*, *367*(6482), 1112-1119. https://doi.org/10.1126/science.aaw7182

Lithgow, G. J., Driscoll, M., & Phillips, P. (2017). A long journey to reproducible results. *Nature*, *548*(7668), 387-388. https://doi.org/10.1038/548387a

Luan, H., Lemon, W. C., Peabody, N. C., Pohl, J. B., Zelensky, P. K., Wang, D., Nitabach, M. N., Holmes, T. C., & White, B. H. (2006). Functional dissection of a neuronal network required for cuticle tanning and wing expansion in Drosophila. *J Neurosci*, *26*(2), 573-584. https://doi.org/10.1523/JNEUROSCI.3916-05.2006

Luan, H., Peabody, N. C., Vinson, C. R., & White, B. H. (2006). Refined spatial manipulation of neuronal function by combinatorial restriction of transgene expression. *Neuron*, *52*(3), 425-436. https://doi.org/10.1016/j.neuron.2006.08.028

Mathejczyk, T. F., & Wernet, M. F. (2019). Heading choices of flying Drosophila under changing angles of polarized light. *Sci Rep*, *9*(1), 16773. https://doi.org/10.1038/s41598-019-53330-y

Mathejczyk, T. F., & Wernet, M. F. (2020). Modular assays for the quantitative study of visually guided navigation in both flying and walking flies. *Journal of Neuroscience Methods*, *340*, 108747. https://doi.org/https://doi.org/10.1016/j.jneumeth.2020.108747

Mather, J., & Anderson, R. C. (1993). Personalities of octopuses (Octopus rubescens). *Journal of Comparative Psychology*, *107*, 336-340.

Mischel, W. (1968). *Personality and assessment*. Wiley.

Mischel, W. (1973). Toward a cognitive social learning reconceptualization of personality. *Psychological Review*, *80 4*, 252-283.

Mischel, W., & Shoda, Y. (1995). A cognitive-affective system theory of personality: reconceptualizing situations, dispositions, dynamics, and invariance in personality structure. *Psychological Review*, *102*(2), 246-268. https://doi.org/10.1037/0033-295x.102.2.246

Mischel, W., & Shoda, Y. (1998). RECONCILING PROCESSING DYNAMICS AND PERSONALITY DISPOSITIONS. *Annual Review of Psychology*, *49*(1), 229-258. https://doi.org/10.1146/annurev.psych.49.1.229

Mowles, S. L., Cotton, P. A., & Briffa, M. (2012). Consistent crustaceans: the identification of stable behavioural syndromes in hermit crabs. *Behavioral Ecology and Sociobiology*, *66*(7), 1087-1094. https://doi.org/10.1007/s00265-012-1359-7

Neuser, K., Triphan, T., Mronz, M., Poeck, B., & Strauss, R. (2008). Analysis of a spatial orientation memory in Drosophila. *Nature*, *453*(7199), 1244-1247. https://doi.org/10.1038/nature07003

Otsuna, H., & Ito, K. (2006). Systematic analysis of the visual projection neurons of Drosophila melanogaster. I. Lobula-specific pathways. *J Comp Neurol*, *497*(6), 928-958. https://doi.org/10.1002/cne.21015

Pantoja, C., Hoagland, A., Carroll, E. C., Karalis, V., Conner, A., & Isacoff, E. Y. (2016). Neuromodulatory Regulation of Behavioral Individuality in Zebrafish. *Neuron*, *91*(3), 587-601. https://doi.org/10.1016/j.neuron.2016.06.016

Partridge, L. (1976). Individual differences in feeding efficiencies and feeding preferences of captive great tits. *Animal Behaviour*, *24*(1), 230-240. https://doi.org/https://doi.org/10.1016/S0003-3472(76)80119-4

Pfeiffer, B. D., Ngo, T. T., Hibbard, K. L., Murphy, C., Jenett, A., Truman, J. W., & Rubin, G. M. (2010). Refinement of tools for targeted gene expression in Drosophila. *Genetics*, *186*(2), 735-755. https://doi.org/10.1534/genetics.110.119917

Pulver, S. R., Pashkovski, S. L., Hornstein, N. J., Garrity, P. A., & Griffith, L. C. (2009). Temporal dynamics of neuronal activation by Channelrhodopsin-2 and TRPA1 determine behavioral output in Drosophila larvae. *J Neurophysiol*, *101*(6), 3075-3088. https://doi.org/10.1152/jn.00071.2009

Quinn, W. G., Harris, W. A., & Benzer, S. (1974). Conditioned Behavior in Drosophila melanogaster. *Proceedings of the National Academy of Sciences*, *71*(3), 708-712. https://doi.org/10.1073/pnas.71.3.708

Reale, D., Reader, S. M., Sol, D., McDougall, P. T., & Dingemanse, N. J. (2007). Integrating animal temperament within ecology and evolution. *Biol Rev Camb Philos Soc*, *82*(2), 291-318. https://doi.org/10.1111/j.1469-185X.2007.00010.x

Robie, A. A., Hirokawa, J., Edwards, A. W., Umayam, L. A., Lee, A., Phillips, M. L., Card, G. M., Korff, W., Rubin, G. M., Simpson, J. H., Reiser, M. B., & Branson, K. (2017). Mapping the Neural Substrates of Behavior. *Cell*, *170*(2), 393-406 e328. https://doi.org/10.1016/j.cell.2017.06.032

Ross, L. (1977). The Intuitive Psychologist And His Shortcomings: Distortions in the Attribution Process11The author gratefully acknowledges the assistance of Teresa Amabile, Daryl Bern, Phoebe Ellsworth, Baruch Fischhoff, David Greene, Larry Gross, Mark Lepper, Richard Nisbett, Julia Steinmetz, Amos Tversky, and Philip Zimbardo, all of whom have contributed useful comments and suggestions concerning both the manuscript and the research and conceptual analysis it reports. The preparation of this chapter and reported research were supported by National Institute of Mental Health Research Grant MH 24134. In L. Berkowitz (Ed.), *Advances*

in Experimental Social Psychology (Vol. 10, pp. 173-220). Academic Press. https://doi.org/https://doi.org/10.1016/S0065-2601(08)60357-3

Schuett, W., Dall, S. R. X., Baeumer, J., Kloesener, M. H., Nakagawa, S., Beinlich, F., & Eggers, T. (2011). Personality variation in a clonal insect: The pea aphid, Acyrthosiphon pisum. *Developmental Psychobiology*, *53*(6), 631-640. https://doi.org/https://doi.org/10.1002/dev.20538

Seelig, J. D., Chiappe, M. E., Lott, G. K., Dutta, A., Osborne, J. E., Reiser, M. B., & Jayaraman, V. (2010). Two-photon calcium imaging from head-fixed Drosophila during optomotor walking behavior. *Nat Methods*, *7*(7), 535-540. https://doi.org/10.1038/nmeth.1468

Sih, A., Cote, J., Evans, M., Fogarty, S., & Pruitt, J. (2012). Ecological implications of behavioural syndromes. *Ecology Letters*, *15*(3), 278-289. https://doi.org/https://doi.org/10.1111/j.1461-0248.2011.01731.x

Spoolder, H. A. M., Burbidge, J. A., Lawrence, A. B., Simmins, P. H., & Edwards, S. A. (1996). Individual behavioural differences in pigs: Intra- and inter-test consistency. *Applied Animal Behaviour Science*, *49*(2), 185-198. https://doi.org/10.1016/0168-1591(96)01033-7

Strauss, R., Hanesch, U., Kinkelin, M., Wolf, R., & Heisenberg, M. (1992). No-Bridge of Drosophila Melanogaster: Portrait of a Structural Brain Mutant of the Central Complex. *Journal of Neurogenetics*, *8*(3), 125-155. https://doi.org/10.3109/01677069209083444

Strauss, R., Renner, M., & Götz, K. (2001). Task-specific association of photoreceptor systems and steering parameters in Drosophila. *Journal of Comparative Physiology A*, *187*(8), 617-632. https://doi.org/10.1007/s003590100234

Sweeney, S. T., Broadie, K., Keane, J., Niemann, H., & O'Kane, C. J. (1995). Targeted expression of tetanus toxin light chain in Drosophila specifically eliminates synaptic transmission and causes behavioral defects. *Neuron*, *14*(2), 341-351. https://doi.org/10.1016/0896-6273(95)90290-2

Takola, E., & Schielzeth, H. (2022). Hutchinson's ecological niche for individuals. *Biology & Philosophy*, *37*(4), 25. https://doi.org/10.1007/s10539-022-09849-y

Traynor, B. J., & Singleton, A. B. (2010). Nature versus Nurture: Death of a Dogma, and the Road Ahead. *Neuron*, *68*(2), 196-200. https://doi.org/https://doi.org/10.1016/j.neuron.2010.10.002

Valente, D., Golani, I., & Mitra, P. P. (2007). Analysis of the trajectory of Drosophila melanogaster in a circular open field arena. *PLoS One*, *2*(10), e1083. https://doi.org/10.1371/journal.pone.0001083

Vansteelandt, K., & Van Mechelen, I. (2004). The personality triad in balance: Multidimensional individual differences in situation–behavior profiles. *Journal of Research in Personality*, *38*(4), 367-393. https://doi.org/https://doi.org/10.1016/j.jrp.2003.08.001

Versace, E., Caffini, M., Werkhoven, Z., & de Bivort, B. L. (2020). Individual, but not population asymmetries, are modulated by social environment and genotype in Drosophila melanogaster. *Scientific Reports*, *10*(1), 4480. https://doi.org/10.1038/s41598-020-61410-7

Warren, T. L., Weir, P. T., & Dickinson, M. H. (2018). Flying Drosophila melanogaster maintain arbitrary but stable headings relative to the angle of polarized light. *J Exp Biol*, *221*(Pt 9). https://doi.org/10.1242/jeb.177550

Weisbuch, M., Slepian, M. L., Clarke, A., Ambady, N., & Veenstra-Vanderweele, J. (2010). Behavioral Stability Across Time and Situations: Nonverbal Versus Verbal Consistency. *J Nonverbal Behav*, *34*(1), 43. https://doi.org/10.1007/s10919-009-0079-9

Werkhoven, Z., Bravin, A., Skutt-Kakaria, K., Reimers, P., Pallares, L. F., Ayroles, J., & de Bivort, B. L. (2021). The structure of behavioral variation within a genotype. *Elife*, *10*, e64988. https://doi.org/10.7554/eLife.64988

Werkhoven, Z., Rohrsen, C., Qin, C., Brembs, B., & de Bivort, B. (2019). MARGO (Massively Automated Real-time GUI for Object-tracking), a platform for high-throughput ethology. *PLoS One*, *14*(11), e0224243. https://doi.org/10.1371/journal.pone.0224243

Xu, C. S., Januszewski, M., Lu, Z., Takemura, S.-y., Hayworth, K. J., Huang, G., Shinomiya, K., Maitin-Shepard, J., Ackerman, D., Berg, S., Blakely, T., Bogovic, J., Clements, J., Dolafi, T., Hubbard, P., Kainmueller, D., Katz, W., Kawase, T., Khairy, K. A.,…Plaza, S. M. (2020). A Connectome of the

Adult *Drosophila* Central Brain. *bioRxiv*, 2020.2001.2021.911859. https://doi.org/10.1101/2020.01.21.911859

Zheng, Z., Lauritzen, J. S., Perlman, E., Robinson, C. G., Nichols, M., Milkie, D., Torrens, O., Price, J., Fisher, C. B., Sharifi, N., Calle-Schuler, S. A., Kmecova, L., Ali, I. J., Karsh, B., Trautman, E. T., Bogovic, J. A., Hanslovsky, P., Jefferis, G., Kazhdan, M.,…Bock, D. D. (2018). A Complete Electron Microscopy Volume of the Brain of Adult Drosophila melanogaster. *Cell*, *174*(3), 730-743 e722. https://doi.org/10.1016/j.cell.2018.06.019