1	Supplementary information for:				
2	The interplay between movement, dispersal and morphology in				
3	Tetrahymena ciliates				
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5	Frank Pennekamp ^{1,2} , Jean Clobert ³ & Nicolas Schtickzelle ¹				
6	¹ Earth and Life Institute & Biodiversity Research Centre, Université catholique de Louvain,				
7	Croix du Sud 4, L7.07.04, 1348 Louvain-la-Neuve, Belgium				
8	² Present address: Institute of Evolutionary Biology and Environmental Studies, University of				
9	Zurich, Winterthurerstrasse 190, 8057 Zurich, Switzerland				
10	³ Station d'Ecologie Théorique et Expérimentale, CNRS, 09200 Moulis, France				
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12 (1) Tetrahymena thermophila genotypes used and culture conditions

In this study, we used a set of 44 genetically distinct genotypes (synonyms: clonal lines, strains) 13 of the ciliate protist Tetrahymena thermophila that have a different history in terms of 14 geographical location at date of isolation (Pennekamp et al. 2014, see Table S1). We previously 15 showed they differ in several life history traits such as growth rate, maximum cell density and 16 survival under starvation conditions (Fjerdingstad et al. 2007, Pennekamp 2014). Cells only 17 reproduced clonally because all cells from one genotype share the same mating type, 18 preventing sexual reproduction through conjugation (Collins 2012); this ensures stability of the 19 genotype over the duration of our study. 20

After defrosting cells from stock cultures kept in suspended animation in liquid nitrogen 21 (Altermatt et al. 2015), the 44 genotypes were maintained under standard culture conditions 22 before and during the experiment: axenic liquid culture in a nutrient medium (broth consisting 23 of 2% Proteose peptone and 0.2% yeast extract [Becton Dickinson], diluted in ultrapure water), 24 25 kept at constant 27°C temperature in a light controlled incubator with a 14:10 hours light/dark cycle. Culture stocks were renewed every 10 days by inoculating a 2 mL sample of fresh 26 medium with 100 µL of culture and maintained in 2 mL multi-well plates (CELLSTAR® 27 multiwell plates, Ref. 662102 from Greiner BioOne, Belgium). All manipulations of axenic 28 cultures were conducted under sterile conditions in a laminar flow hood (Ultrasafe 218 S, 29 Faster, Italy). 30

32 (2) Experimental design with two patch dispersal systems

We quantified dispersal and movement behaviour of *T. thermophila* cells using the same
standardized two patch system developed in our previous work (Fjerdingstad et al. 2007,
Schtickzelle et al. 2009, Chaine et al. 2010, Pennekamp et al. 2014) (Fig. S1).

In addition to the strict control of all our *T. thermophila* culture conditions, two standardization 36 steps were performed prior to the experiment. First, a pre-culture of each genotype was started 37 from the stock by transferring 100 µL of culture into 2 mL of fresh nutritive medium (2% 38 Proteose peptone and 0.2% yeast extract [Becton Dickinson], diluted in ultrapure water) on a 39 24 well plate (CELLSTAR® multiwall plate, Ref. 662102 from Greiner BioOne, Belgium) and 40 allowed to grow exponentially for 4 days to synchronize populations to the logarithmic phase 41 of population growth (Collins 2012). Second, at the end of this synchronization phase, cell 42 43 density was estimated for each genotype, and new cultures, to be used for the experiment, were launched at an equal starting density of 10000 cells/mL in culture flasks (CELLSTAR® Cell 44 45 Culture Flask 50 mL, Ref. 690175 from Greiner BioOne, Belgium). These cultures grew for three days allowing them to reach sufficiently high cell densities for the experiment. 46

Each dispersal system consisted of two standard 1.5 mL microtubes connected by a silicon pipe 47 (length: 17 mm; external diameter: 6 mm; Ref. 228-0709 from VWR, Belgium). The dispersal 48 49 system was filled prior to the experiments with 3 mL of the standard nutritive medium through one tube to ensure fluid transition between the tubes and a connecting pipe free of air bubbles. 50 The system was then closed by placing a clamp in the middle of the connecting pipe. To start 51 the experiment, cells were inoculated into the "start" tube of the system at a density of density 52 of 300000 cells / mL (i.e. 450000 cells for a 1.5 mL volume) and the tube content was 53 54 homogenized to encourage the cells to move freely throughout the start tube. After 30 minutes of acclimation to the new medium, the clamp closing the connecting pipe was removed and 55 cells could freely disperse between the two tubes for 6 h. After these 6 h, the pipe was clamped 56

again, and five independent samples taken from the "start" and "target" tubes after culture
homogenization. Each sample was loaded into the chamber of a counting slide (Precision cell
301890, Vacutest Kima, Italy).

61 (3) Reconstructing movement trajectories from videos

We developed for this study a workflow to extract movement trajectories from digital videos 62 in a standardized and automated fashion, which was later transformed into the R package 63 BEMOVI (Pennekamp et al. 2015). We customized the ParticleTracker plug-in for ImageJ 64 software, originally aimed at tracking intracellular movements of cell structure 65 (http://mosaic.mpi-cbg.de/ParticleTracker/). The ParticleTracker links each position of a given 66 cell, as recorded on every frame of the video, into a unique trajectory of time-stamped X and 67 Y coordinates, which are output as text files. The 25 frames per second acquisition speed for 68 our videos guarantees correct assembly of positions into individual trajectories even when 69 many cells are tracked simultaneously. ParticleTracker's feature point tracking algorithm is 70 described in detail in Sbalzarini & Koumoutsakos (2005) and has several powerful features: 1) 71 tracking of many individuals simultaneously is feasible and due to low computational 72 73 requirements longer video sequences may be analysed; 2) the plug-in deals with unrestricted viewing fields, i.e. cells may leave and enter the video because the viewing field is not 74 physically bounded; 3) in case the algorithm is unable to retain the identity of a given cell 75 because of collisions between tracked cells or with artefacts on the video such as dust, it acts 76 conservatively by terminating the current trajectories and assigning new trajectory identities 77 when cells are again separated. 78

Home-made SAS scripts (www.sas.com) were used to read the raw trajectory data extracted by ParticleTracker (Fig. S2). 49% of the recorded trajectories (258,592 out of 525,328) were discarded because they lasted less than 1 s or had a total net displacement (i.e. the bee line distance between the start and the end position of the trajectory) less than 50 μ m (corresponding to one body length of an average *Tetrahymena* cell); such trajectories correspond to nonmoving cells. For analysis with continuous time movement models (Fleming et al. 2014, Gurarie et al. 2017), which is highly computationally demanding, we then subsampled 23

- trajectories for each genotype x replicate x tube combination, resulting in 6072 trajectories
- sampled from a total of 266,736 trajectories of moving cells.

89 Figures

- 90 Figure S1: A two patch dispersal system made of two 1.5 mL microtubes connected by a
- silicon pipe, filled with nutritive medium, used to quantify dispersal rate from cells
- 92 inoculated in the start patch and allowed to move freely during 6 h.



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- 95 Figure S2: An illustration of the raw trajectory data extracted from videos
- 96 Different colours show different individual trajectories. The linearity differed among
- 97 trajectories with some being very linear and others more tortuous (see arrows). Some very
- 98 short (in time or space) trajectories correspond to non-moving cells.



Figure S3: A positive correlation between movement speed and linearity was found across
genotypes. Faster genotype moved more linear. The strength of the relationship did not differ
with dispersal status.



Figure S4: Relationships between movement and morphology differences between residents
and dispersers. Genotypes with relatively larger dispersers showed relatively faster and more
linear movements. Relatively more elongated dispersers showed relatively slower and less
linear movements.



112 Tables

113 Table S1: List of *Tetrahymena thermophila* strains used in this study, as well as the distributor,

114 isolator and geographic location from which the strain was obtained.

Our name	Distributor and reference	Isolator and isolation date	Geographic isolation
7	ATCC 30306	D.L. Nanney, 1953	Woods Hole, MA
20	TSC SD01236	E. Orias	Laboratory created
4A	CCAP 1630/4A	Nanney & McCoy, 1976	Unknown
В	ATCC 30384 (B-18687)	Simon & Nanney, 1968	Laboratory created
С	TSC SD01216	Yuhua Shang, 2002	Laboratory created
Е	ATCC 205043	Simon & Nanney, 1986	McCurdy Pond, ME
F	TSC SD00086	P. J. Bruns	Laboratory created
G	TSC SD00112	Unknown	Unknown
Н	TSC SD00270	Unknown	Unknown
Ι	TSC SD01206	Unknown	Unknown
J	TSC SD00626	Unknown, 1976	Laboratory created
K	TSC SD01590	Unknown	Laboratory created
L	TSC SD01223	Unknown, 1979	Laboratory created
М	CCAP 1630/1M	Phelps, 1948/9	Unknown, AZ
Ν	CCAP 1630/1N	Phelps, 1948/9	Unknown, AZ
Ο	TSC SD01422 (NP1)	L. Rasmussen, 1968	Laboratory created
Р	CCAP 1630/1P	Phelps, 1948/9	Unknown, AZ
Q	CCAP 1630/1Q	Phelps, 1948/9	Unknown, AZ
R	TSC SD00703 (SB210)	E. Orias	Laboratory created
S	TSC SD01532	Unknown	Laboratory created
Т	TSC SD01538	Unknown	Laboratory created
U	CCAP 1630/1U (WH14)	Elliott, 1952	Woods hole, MA
D1	TSC SD01546	Doerder, 8/2002	CRWP, PA
D2	TSC SD01547	Doerder, 8/2002	CRWP, PA
D3	TSC SD01548	Doerder, 06/2003	SG29, PA
D4	TSC SD01549	Doerder, 06/2003	SG29, PA
D5	TSC SD01550	Doerder, 06/2003	SG29, PA
D6	TSC SD01551	Doerder, 06/2003	SG29, PA
D7	Doerder AK III	Doerder	Unknown
D8	TSC SD01553	Doerder, 07/2008	FS136NW, PA
D9	TSC SD01552	Doerder, 07/2008	Beaver Meadows, PA
D10	TSC SD01557	Doerder, 07/2009	Lake Warren, NH
D11	TSC SD01558	Doerder, 07/2009	IslandPond#1, NH
D12	TSC SD01556	Doerder, 08/2008	SG69-1, PA
D13	TSC SD01555	Doerder, 08/2008	SG69-4, PA
D14	TSC SD01554	Doerder, 08/2008	SG69-6, PA
D15	TSC SD01560	Doerder, 07/2009	Gregg Lake, NH
D16	TSC SD01559	Doerder, 07/2009	Gregg Lake, NH
D17	TSC SD01561	Doerder, 07/2009	Willard Pond, NH
D18	TSC SD01562	Doerder, 07/2009	Willard Pond, NH
D19	TSC SD01564	Doerder, 07/2009	Childs Bog, NH
D20	TSC SD01563	Doerder, 07/2009	Russell Reservoir, NH
D21	TSC SD01565	Doerder, 07/2009	Perkins Pond, NH
D22	TSC SD01566	Doerder, 07/2009	South Pond, VT

- **117** Model selection among and within genotypes:
- **Table S2: Model selection to identify whether dispersal status and morphological properties**
- 119 (predictors) explain variation in movement speed (response). The most parsimonious model is
- 120 shown in bold. K = number of parameters, AICc = Akaike information criterion value, delta =
- 121 difference with the lowest AIC value, weight = AIC weight.

Model	K	AICc	delta	weight
speed ~ disp_status + shape + size + disp_status:shape + 1	6	2846.68	0	0.69
speed ~ disp_status + shape + size + disp_status:shape + disp_status:size + 1	7	2848.72	2.04	0.25
speed ~ disp_status + shape + disp_status:shape + 1	5	2851.37	4.68	0.07
speed \sim disp_status + shape + size + 1	5	2861.1	14.41	0
speed ~ disp_status + shape + size + disp_status:size + 1	6	2863.19	16.51	0
speed \sim disp_status + size + 1	4	2863.27	16.58	0
speed \sim disp_status + shape + 1	4	2865.25	18.56	0
speed \sim disp_status + size + disp_status:size + 1	5	2865.35	18.66	0
speed \sim disp_status + 1	3	2866.73	20.05	0
speed ~ size + 1	3	2881.54	34.86	0
speed \sim shape + size + 1	4	2883.29	36.6	0
speed ~ 1	2	2884.02	37.33	0
speed \sim shape + 1	3	2885.86	39.18	0

- 123 Table S3: Model selection to identify whether dispersal status and morphological properties
- 124 (predictors) explain variation in tau, which describes movement linearity (response). The most
- 125 parsimonious model is shown in **bold**. K = number of parameters, AICc = Akaike information
- 126 criterion value, delta = difference with the lowest AIC value, weight = AIC weight.

Model	К	AICc	delta	weight
tau ~ disp_status + shape + disp_status:shape + 1	5	-1129.43	0	0.58
$tau \sim disp_status + shape + size + disp_status:shape + 1$	6	-1128.16	1.27	0.31
tau ~ disp_status + shape + size + disp_status:shape + disp_status:size + 1	7	-1126.06	3.37	0.11
$tau \sim disp_status + shape + 1$	4	-1114.82	14.6	0
$tau \sim disp_status + shape + size + 1$	5	-1113.47	15.96	0
tau ~ disp_status + shape + size + disp_status:size + 1	6	-1111.54	17.89	0
tau ~ disp_status + 1	3	-1106.99	22.44	0
tau ~ disp_status + size + 1	4	-1105.33	24.1	0
tau ~ disp_status + size + disp_status:size + 1	5	-1103.44	25.99	0
tau ~ 1	2	-1079.59	49.84	0
$tau \sim shape + 1$	3	-1079.1	50.32	0
$tau \sim size + 1$	3	-1077.73	51.7	0
$tau \sim shape + size + 1$	4	-1077.29	52.14	0

128 Model selection within genotypes:

- 129 Table S4: Model selection to identify whether differences in morphology (predictors) explain
- 130 differences in speed (response). The most parsimonious model is shown in **bold**. K = number of
- 131 parameters, AICc = Akaike information criterion value, delta = difference with the lowest AIC
- 132 value, weight = AIC weight.

Model	К	AICc	delta	weight
<pre>speed_diff ~ shape_diff + size_diff + shape_diff:size_diff + 1</pre>	5	129.54	0	0.46
<pre>speed_diff ~ shape_diff + size_diff + 1</pre>	4	130.32	0.78	0.31
speed_diff ~ size_diff + 1	3	131.99	2.45	0.13
speed_diff ~ shape_diff + 1	3	133.43	3.89	0.07
speed_diff ~ 1	2	134.87	5.33	0.03

133

- 134 Table S5: Model selection to identify whether differences in morphology (predictors) explain
- 135 differences in tau (response), which describes movement linearity. The most parsimonious
- 136 model is shown in **bold**. K = number of parameters, AICc = Akaike information criterion value,
- 137 delta = difference with the lowest AIC value, weight = AIC weight.

Model	К	AICc	delta	weight
<pre>tau_diff ~ shape_diff + size_diff + shape_diff:size_diff + 1</pre>	5	290.04	0	0.62
tau_diff ~ shape_diff + size_diff + 1	4	291.51	1.47	0.3
tau_diff ~ shape_diff + 1	3	294.22	4.18	0.08
tau_diff ~ size_diff + 1	3	307.8	17.76	0
tau_diff ~ 1	2	309.7	19.65	0

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