The information theory formalism unifies the detection of the patterns of sexual selection and assortative mating for both discrete and quantitative traits.

A. Carvajal-Rodríguez*

Centro de Investigación Mariña (CIM), Departamento de Bioquímica, Genética e

Inmunología. Universidade de Vigo, 36310 Vigo, Spain.

*: A. Carvajal-Rodríguez. Departamento de Bioquímica, Genética e Inmunología.

Universidade de Vigo, 36310 Vigo, Spain. Phone: +34 986813828

email: acraaj@uvigo.es

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Abstract

Sexual selection plays a crucial role in modern evolutionary theory, offering valuable

insight into evolutionary patterns and species diversity. Recently, a comprehensive

definition of sexual selection has been proposed, defining a trait as subject to sexual

selection when variations in that trait are nonrandomly associated with variations in

access to gametes, a limiting resource. Previous research on discrete traits demonstrated

that non-random mating can be effectively quantified using the Kullback-Leibler

symmetrized (Jeffreys) divergence, capturing information acquired through mating

influenced by mutual mating propensities instead of random occurrences. This novel

theoretical framework allows for detecting and assessing the strength of sexual selection

and assortative mating.

In this study, we aim to achieve two primary objectives. Firstly, we demonstrate the

seamless alignment of the previous theoretical development, rooted in information

theory and mutual mating propensity, with the new sexual selection definition.

Secondly, we extend the theory to encompass quantitative traits. Our findings reveal

that sexual selection and assortative mating can be quantified effectively for quantitative

traits by measuring the information gain concerning the random mating pattern.

Specifically, the gain in information within the sexual selection pattern can be expressed

as a function of the sexual selection intensity index and the ratio of variances.

Additionally, the measure capturing the underlying information of assortative mating is

a function of the square of the correlation coefficient, taking values within the non-

negative real number set $[0, +\infty)$.

It is worth noting that the same divergence measure captures information acquired

through mating for both discrete and quantitative traits. This presents a remarkable

innovation as it provides a common context and simplifies the study of the patterns of

sexual selection.

Introduction

Sexual selection is a fundamental concept in modern evolutionary theory, offering

valuable insights into various evolutionary patterns and the diversity of species.

Originally proposed by Darwin in (1871), sexual selection describes the competition

among individuals of one sex to secure mates from the other sex. However, the concept

of sexual selection has been a subject of controversy since its inception (Andersson

1994; Prum 2012; Parker 2014; Parker and Pizzari 2015; Roughgarden et al. 2015;

Alonzo and Servedio 2019; Lehtonen 2022; Vries and Lehtonen 2023). There is

ongoing debate surrounding its precise definition (Fitze et al. 2011), and its role as a key

component of modern evolutionary biology has even faced challenges (Roughgarden et

al. 2006; Parker and Pizzari 2015; but see Shuker 2010).

Recently, Shuker and Kvarnemo (2021) put forth a comprehensive definition of sexual

selection, characterizing it as "any selection that arises from fitness differences

associated with nonrandom success in the competition for access to gametes for

fertilization". In essence, a trait falls under sexual selection when variations in that trait

are nonrandomly associated with variations in access to gametes, which serve as a

limiting resource. From an evolutionary perspective, this process of sexual selection

generates a discernible pattern of change in gene frequency, influencing the frequencies

of corresponding genotypes and phenotypes. It is important to emphasize the use of the

term "pattern of change" when discussing sexual selection, as it helps to clarify whether

we are referring to the dynamic process or to the recognizable pattern in genetic or

phenotypic changes.

Darwin identified two primary biological mechanisms of sexual selection: mate

competition and mate choice (Ng et al. 2019 and references therein). The mate choice

process refers to the effects of certain traits expressed in one sex that leads to the non-

random allocation of reproductive effort with members of the opposite sex (Edward

2015; Rosenthal 2017). Mate choice may produce a mating pattern where similar

phenotypes or genotypes mate with one another more or less frequently than would be

expected under a random mating pattern i.e. it produces a deviation from random

mating in mated individuals (Rolán-Alvarez and Caballero 2000; Carvajal-Rodríguez

2018b). When the observed mating pattern between similar phenotypes is more frequent

that the expected by chance, we refer to positive assortative mating and when it is less

frequent we refer to negative assortative mating. Sometimes in the literature the positive

assortative mating is just referred as assortative mating and the negative assortative

mating as disassortative mating.

Mate choice when it produces positive assortative mating can have important

evolutionary effects as it alters the variance of quantitative traits (Fisher 1918), can

generate sexual selection (de Cara et al. 2008; Carvajal-Rodríguez 2018b) and under

certain conditions, may facilitate local adaptation and speciation even in the presence of

gene flow (Kirkpatrick and Ravigné 2002; Coyne and Orr 2004; Gavrilets 2004;

Rettelbach et al. 2013; Cotto and Servedio 2017; Sachdeva and Barton 2017;

Fernández-Meirama et al. 2022; Yukilevich 2023).

When examining phenotypes involving discrete traits, assortative mating can be

identified through mating data using the pair sexual isolation statistic (Rolán-Alvarez

and Caballero 2000; Carvajal-Rodríguez and Rolan-Alvarez 2006). For quantitative

traits, assessing assortative mating can be achieved by computing the Pearson

correlation coefficient within mated pairs (Jiang et al. 2013). However, it is important to

note that this measure comes with known problems and limitations (Clancey et al.

2022; Fernández-Meirama et al. 2022).

In a previous study (Carvajal-Rodríguez 2018b), it was demonstrated that non-random

mating can be effectively quantified using the Jeffreys (Kullback-Leibler symmetrized)

divergence. This measure captures the information acquired when mating is influenced

by mutual mating propensities, rather than occurring randomly. By employing this

novel theoretical framework, it becomes possible to detect and assess the strength of

sexual selection and assortative mating (Carvajal-Rodríguez 2020; Estévez et al. 2020;

Lau et al. 2021). However, it should be noted that the original framework was designed

for discrete classes, and in this current work, we aim to extend its applicability to

continuous cases.

The objective of this work is twofold. Firstly, we aim to demonstrate that the previous

theoretical development, which is rooted in information theory and the concept of

mutual mating propensity (Carvajal-Rodríguez 2018b, 2020), seamlessly aligns with the

new sexual selection definition put forth by Shuker and Kvarnemo (2021). In this

context, sexual selection emerges as a consequence of variations in the mutual mating

propensity, without requiring specific assumptions about the genetic basis of the

involved traits. Secondly, we aim to illustrate that, for quantitative traits, non-random

mating is also measured by Jeffreys divergence, which implies that the pattern of sexual

selection, quantified as the difference between the phenotypic distribution of mated

individuals and the population distribution , can be expressed as a function of the sexual

selection intensity index (sensu Ng et al. 2019) and the variance ratio. Additionally, we

emphasize that the measure capturing the underlying information of assortative mating

is not the correlation itself, but rather a function of the square of the correlation

coefficient, that takes values within the non-negative real number set $[0, +\infty)$.

Fitness and sexual selection

Fitness, defined as an organism's ability or propensity to survive and reproduce relative

to others in the same population drives natural selection by leading to changes in gene

frequencies (Hedrick 2005; Hartl 2020). Identifying natural selection relies on

recognizing the patterns of change it produces in frequencies.

Consider a population with various types, each representing a set of individuals with

specific properties (traits, genotypes, etc.). Let N_i be the number of individuals of type i,

and suppose a selection episode changes this number to N'_i. The absolute fitness, or

growth rate, of type i during this episode is denoted as $W_i = N'_i/N_i$. To compare different

types' fitness fairly, we compute the relative fitness, w_i, as the ratio of their absolute

fitness to the mean population fitness. Thus, for a given type i with frequency p_i, the

new frequency p'_i after a selection episode will be given by (Bürger 2000; Hansen 2018)

119              $p'_i = p_i w_i$ (1).

Recently, a simple theoretical framework, similar to the model presented in (1), has

been developed to describe the influence of mating competition and mate choice on the

emergence of sexual selection and assortative mating patterns (Carvajal-Rodríguez

2018b, 2019, 2020). This theoretical framework has enabled the development of new

statistical tests for detecting sexual selection and assortative mating. Moreover, by

linking variation in mating ability (i.e., gamete access for fertilization) to the resulting

frequency patterns, this framework facilitates the construction of predictive models and

supports multi-model inference based on observed data patterns (Carvajal-Rodríguez

2020).

In the following sections, I will briefly review and elucidate this theoretical framework,

demonstrating its compatibility with Shuker and Kvarnemo's (SK hereafter) definition

of sexual selection ("any selection that arises from fitness differences associated with

nonrandom success in the competition for access to gametes for fertilization"; Shuker

and Kvarnemo 2021).

Fitness for mating

We will focus on the change in frequencies due to fitness differences associated with

access to gametes for fertilization. First, we must specify the concept of individual

fitness for mating. Secondly, we will express for the discrete case, the SK definition of

sexual selection in terms of information and then, extend the definition to the

continuous case.

Let a female character X with a total of k_1 discrete possible trait values and a male

character Y with k_2 discrete values. Characters X and Y can be the same or not. We

define M_{ij} as the mutual ability or propensity to access gametes for fertilization between

a female with value i for X and a male with value j for Y. Let p_{1i} be the population

frequency of type i females and p_{2j} the frequency of type j males so that $q_{ij} = p_{1i}p_{2j}$ is the

probability of random pairing of gametes from type i females with gametes from type j

males for fertilization. The mean propensity is

$$M = \sum_{i=1}^{k1} \sum_{j=1}^{k2} q_{ij} M_{ij}$$

and the relative mutual propensity $m_{ij} = \dfrac{M_{ij}}{M}$.

The probability q'_{ij} of mating or gamete access for fertilization between any pair having

such a combination of traits is (Carvajal-Rodríguez 2018b)

$$q'_{ij} = q_{ij} m_{ij} \quad (2).$$

Equation (2) represents the expected frequencies of mating between types i and j as a

function of the expected frequencies per random gamete encounter and the relative

values of mutual propensity. The equation clearly resembles equation (1) for the change

in frequencies because of selection. In fact, the absolute value M_{ij} can be expressed in a

similar way to absolute fitness i.e., as the ratio of the number N'_{ij} of mating between

types i and j after selection over the expected number N_{ij} if there were no selection

(random mating).

$$M_{ij} = \dfrac{N'_{ij}}{N_{ij}}.$$

Our focus is on identifying the individual propensity to access gametes for fertilization,

rather than the mutual propensity of pairs. We refer to this individual propensity as the

fitness for mating. In this context, we define the absolute marginal propensity M_{Fi} of the

female of type i as

$$M_{Fi} = \sum_{j=1}^{k2} p_{2j} M_{ij}.$$

The relative marginal propensity or mating fitness m_{Fi} is obtained by dividing by the

mean propensity M

$$m_{Fi} = \frac{M_{Fi}}{M} \qquad (3).$$

Similarly, the relative marginal propensity or mating fitness of males of type j is

$$m_{Mj} = \sum_{i=1}^{k1} p_{1i} m_{ij} = \frac{M_{Mj}}{M}.$$

Thus, the relative mating fitness of a type is the average relative ability of individuals of

that type to access gametes for fertilization.

While different mutual propensities shape mate choice, different mating fitness within a

sex will produce a pattern of sexual selection (see Figure1 in Carvajal-Rodríguez 2020).

This pattern arises due to the unequal ability to access the gametes of the opposite sex

for fertilization. Consequently, this sexual selection aligns perfectly with SK's

definition.

Importantly, the concepts of mutual propensity and mating fitness, as defined, do not

presuppose any particular mating preference function.

Having formalized the concepts of mutual propensity and mating fitness it is then

possible to model mate competition and choice and the patterns of sexual selection and

assortative mating they produce (see Carvajal-Rodríguez 2018b, 2020). The pattern of

sexual selection is caused mainly by two biological mechanisms: mate competition and

mate choice (Lewontin et al. 1968; Arnold and Wade 1984; Endler 1986; Rolán-Alvarez

and Caballero 2000; Rolán-Alvarez et al. 2015).

The process of mate competition refers, in a broad sense, to access to mating through

courtship, intrasexual aggression and/or competition for limited reproductive resources

(Andersson 1994; Kokko et al. 2012; Wacker and Amundsen 2014). Mate competition

may generate a pattern of sexual selection (a change in frequencies of the trait under

study) in the sex that competes.

The process of mate choice is the behavioral propensity to mate with individuals having

certain traits and occurs whenever the effects of traits expressed in one sex leads to non-

random allocation of reproductive investment with members of the opposite sex

(Edward 2015 and references therein). Choice may be mediated by phenotypic

(sensorial or behavioral) properties that affect the propensity of individuals to mate with

certain phenotypes (Jennions and Petrie 1997).The observed pattern driven by mate

choice can be a change in trait frequency in the other sex (sexual selection) and/or

assortative mating.

The model in (2) facilitates the development of new statistical tests, based on

information theory, to detect patterns of sexual selection and assortative mating

(Carvajal-Rodríguez 2018b). Moreover, it enables to formalize the necessary and

sufficient conditions for random mating in terms of mating propensities. Once these

conditions are defined, we can generate different mating fitness scenarios that produce

the intended patterns of sexual selection and assortative mating, allowing the

application of powerful model inference techniques (Carvajal-Rodríguez 2020).

The formalization of the mate choice and competition processes from the model

presented in (2) have been carried out only for discrete traits and here we extend the

mathematical framework to the continuous case. As a result, we will see that the pattern

of sexual selection can be expressed as a function of the sexual selection intensity index

and the ratio of variances, and the pattern of assortative mating is captured by a function

of the square of the Pearson correlation instead of the correlation per se.

Sexual selection and assortative mating in quantitative traits

Consider two continuous traits in a population, X in females and Y in males, and let be

the mating probability distribution Q'(x,y) with probability density $q'(x,y)$ representing

the mating probability for pairs with values in the infinitesimal interval ([$x,x+dx$],

[$y,y+dy$]). Alternatively consider the probability distribution $Q(x,y)$ of random mating

from the product of densities $q(x,y)=f(x)g(y)$ of X and Y, respectively.

Note that (2) still holds so that $q'(x,y)$ depends on the density $q(x,y)$ and the infinitesimal

mutual mating propensity m_{xy} which is the relative mutual propensity to access gametes

for fertilization between a female with value in the infinitesimal interval [$x,x+dx$] and a

male with value in the infinitesimal interval [$y,y+dy$].

In earlier work, Carvajal-Rodríguez (2018b) showed for discrete traits, that non-random

mating can be characterized as the divergence between the mating distribution, Q',

derived from mutual mating propensities, and the distribution of random mating, Q. For

the continuous case we have that the divergence between the two distributions Q' and Q

is given by

$$J = \int_{-a}^{+a} \int_{-a}^{+a} (q'(x,y) - q(x,y)) \log \frac{q'(x,y)}{q(x,y)} \, dx \, dy \qquad (4),$$

where $q(x,y)=f(x)g(y)$.

For simplicity and without lost of generality we assume that the range is the same for X

and Y so that, $X, Y \in [-a, a]$.

The distribution function of X in the population

$$F(x) = \int_{-a}^{x} f(u) \, du$$

and Y

$$G(y) = \int_{-a}^{y} g(v) \, dv,$$

where $f(u)$ and $g(v)$ are the respective probability densities.

From the probability density $q'(x,y)$, the marginal densities for X and Y in the matings

are

$$f_1(x) = \int_{-a}^{+a} q'(x, y) \, dy$$

$$f_2(y) = \int_{-a}^{+a} q'(x, y) \, dx$$

Sexual selection

The difference $q'(x,y)$-$q(x,y)$ in (4) can be partitioned as follows

$$q'(x, y) - q(x, y) = f_1(x) f_2(y) - q(x, y) + q'(x, y) - f_1(x) f_2(y),$$

similarly, the logarithm

$$\log \frac{q'(x, y)}{q(x, y)} = \log \left(\frac{f_1(x) f_2(y)}{q(x, y)} \frac{q'(x, y)}{f_1(x) f_2(y)} \right) = \log \frac{f_1(x) f_2(y)}{q(x, y)} + \log \frac{q'(x, y)}{f_1(x) f_2(y)}$$

Recall that the pattern of sexual selection is quantified by comparing the expected mate

types resulting from random mating among mated individuals with those calculated

using the population frequencies (Hartl and Clark 1997; Rolán-Alvarez and Caballero

2000). In the continuous scenario, this is formalized by the Jeffreys divergence

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$$J_{PSS}=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}(f_1(x)f_2(y)-q(x,y))\log\frac{f_1(x)f_2(y)}{q(x,y)}dxdy \qquad (5)$$

248   and because

249
$$\int\limits_{-a}^{+a}f_1(x)dx=\int\limits_{-a}^{+a}f_2(y)dy=1 \quad \text{and} \quad \int\limits_{-a}^{+a}f(x)dx=\int\limits_{-a}^{+a}g(y)dy=1$$

250   we can separate by sex

251    $J_{PSS}=J_{PS1}+J_{PS2}$ where

252
$$J_{PS1}=\int\limits_{-a}^{+a}(f_1(x)-f(x))\log\frac{f_1(x)}{f(x)}dx$$

253
$$J_{PS2}=\int\limits_{-a}^{+a}(f_2(y)-g(y))\log\frac{f_2(y)}{g(y)}dy.$$

Therefore, similar to the approach applied in the discrete case (Carvajal-Rodríguez

2018b), J_{PSS} quantifies sexual selection for quantitative traits as the change in phenotype

distribution resulting from differential mating. The information captured in J_{PSS} is the

sum of the sexual selection within each sex.

Normally distributed traits

Let assume that $q'(x,y)$ is a bivariate normal density and traits X and Y are normally

distributed. Then, the marginal density function for the trait in females within matings is

$f_1(x) \sim N(\mu_1,\sigma_1^2)$, and in the population is $f(x) \sim N(\mu_x,\sigma_x^2)$. The marginal density function

for the trait in males within matings is $f_2(y) \sim N(\mu_2,\sigma_2^2)$ and in the population is $g(y) \sim$

$N(\mu_y,\sigma_y^2)$ which are normally distributed and J_{PS1} and J_{PS2} can be expressed in closed

form as (Kullback 1997; Belov and Armstrong 2011)

$$J_{PS1}=\frac{1}{2}\left(\frac{\Phi_1{}^2+1}{\Phi_1}+\frac{\Phi_1+1}{\Phi_1}\frac{(\mu_1-\mu_x)^2}{\sigma_x{}^2}-2\right) \quad \text{where} \quad \Phi_1=\sigma_1{}^2/\sigma_x{}^2.$$

$$J_{PS2}=\frac{1}{2}\left(\frac{\Phi_2{}^2+1}{\Phi_2}+\frac{\Phi_2+1}{\Phi_2}\frac{(\mu_2-\mu_y)^2}{\sigma_y{}^2}-2\right) \quad \text{where} \quad \Phi_2=\sigma_2{}^2/\sigma_y{}^2.$$

with nJ_{PS1} (nJ_{PS2}) asymptotically distributed as χ^2 with 2 degrees of freedom under the

null hypothesis that the females (males) within matings are distributed with mean μ_x

(μ_y) and variance σ_x^2 (σ_y^2) where n is the sample size of matings.

Note that in the above expressions, the term for the difference of the means is the square

of the standardized sexual selection intensity (SSI) index (Erlandsson and Rolán-

Alvarez 1998; Ng et al. 2019; Caballero 2020) so that

$$J_{PS1}=\frac{1}{2}\left(\frac{\Phi_1{}^2+1}{\Phi_1}+\frac{\Phi_1+1}{\Phi_1}SSI_1{}^2-2\right)$$

$$J_{PS2}=\frac{1}{2}\left(\frac{\Phi_2{}^2+1}{\Phi_2}+\frac{\Phi_2+1}{\Phi_2}SSI_2{}^2-2\right) \qquad (6)$$

where SSI_1 is the SSI index for females and SSI_2 for males.

If we want to test only the mean value of the female trait X in the mating sample, we set

$\Phi_1=1$ so that

$$J_{PS1\,\Phi_1=1}=\frac{(\mu_1-\mu_x)^2}{\sigma_1{}^2}$$

and with a sample size of n matings we have $(nJ_{PS1\Phi=1})^{0.5} \sim t_{n-1}$. Similarly, for the males

we have $(nJ_{PS2\Phi=1})^{0.5} \sim t_{n-1}$ with

$$J_{PS2\,\Phi_2=1}=\frac{(\mu_2-\mu_y)^2}{\sigma_2{}^2}.$$

For large sample sizes both types of tests (χ^2 or t) work similar, for a smaller sample

size the χ^2 test might be slightly liberal and the t-test should be preferred.

Assortative mating

The assortative mating pattern corresponds to the deviation from random mating

distribution measured in mated individuals i.e. the divergence

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$$J_{PSI} = \int\limits_{-a}^{+a} \int\limits_{-a}^{+a} (q'(x,y) - f_1(x)f_2(y)) \log \frac{q'(x,y)}{f_1(x)f_2(y)} dxdy \quad (7).$$

Normally distributed traits

As before, suppose that $q'(x,y)$ is a bivariate normal density and that traits X and Y are

normally distributed. The divergence for assortative mating in (7) is the Jeffreys (or

Kullback-Leibler symmetrized) divergence between the joint distribution and the

product of the marginals and serves as a measure of the relationship between X and Y

that can be expressed as a function of the correlation coefficient (see Appendix A and

Kullback 1997) so,

295
$$J_{PSI} = \int\limits_{-a}^{+a} \int\limits_{-a}^{+a} (q'(x,y) - f_1(x)f_2(y)) \log \frac{q'(x,y)}{f_1(x)f_2(y)} dxdy = \frac{\rho^2}{1-\rho^2} \quad (8).$$

Thus, as long as the quantitative traits we are measuring are normally distributed, the

natural measure of assortative mating is a function of the square of correlation

coefficient that takes values within the set of nonnegative real numbers $[0, +\infty)$.

The statistical significance of $J_{PSI}>0$ can be tested since nJ_{PSI} is asymptotically

distributed as χ^2 with 1 degree of freedom (Kullback 1997). To distinguish between

positive and negative assortative mating, we only need to consider the sign of the

Pearson correlation coefficient ρ.

Finally, the total divergence in (4) can be expressed as the sum of sexual selection (5)

and assortative mating (7) divergences plus a factor E

305
$$J = J_{PS1} + J_{PS2} + J_{PSI} + E$$

306   where (see Appendix B)

$$E = \int_{-a}^{+a} \int_{-a}^{+a} (f_1(x)f_2(y) - q(x,y)) \log \frac{q'(x,y)}{f_1(x)f_2(y)} dxdy \neq 0 \Leftrightarrow J_{PSS} \neq 0 \wedge J_{PSI} \neq 0.$$

308   So the E component only appears when both sexual selection ($J_{PSS} \neq 0$) and assortative

309   mating ($J_{PSI} \neq 0$) are present.

Simulations

Detection of sexual selection

We will utilize the program MateSim (Carvajal-Rodríguez 2018a) to simulate the

mating distribution for quantitative traits using a logistic model (Xie et al. 2015). This

model represents an open-ended preference function, capable of generating a pattern of

sexual selection in both females and males.

The simulation procedure consists of two main steps. First a population of size N is

generated, comprising $N/2$ females and $N/2$ males. The individuals are randomly

sampled from a standard normal distribution with a mean of 0 and a standard deviation

of 1, $Z(0,1)$. Second, a sample of n ($n<N$) matings is obtained as the individuals interact

to form pairs for mating. The mating probability of a male j with phenotype Z_j of mating

with any female F is evaluated as follows:

$$P(F,j) = \exp(-1 + \alpha Z_j)/[1 + \exp(-1 + \alpha Z_j)]$$

and the same applies for a female i with phenotype Z_i to mate with any male M.

$$P(i,M) = \exp(-1 + \alpha Z_i)/[1 + \exp(-1 + \alpha Z_i)]$$

Therefore, the mating probability after encounter for a pair $i \times j$ will be

$$p_{ij} = P(i,M) \times P(F,j).$$

The parameter α (≥0) accounts for the strength of sexual selection, if α=0 there is

random mating, the higher the value of α, the greater the strength of sexual selection.

The tests in (6) are able of distinguishing between random mating and sexual selection

for different sample sizes and sexual selection intensities. Not surprisingly, the greater

the sexual selection and the size of the sample, the greater the power of detection and

the value of the information recovered by the J_{PS1} and J_{PS2} indexes (Table 1). For a large

sample size, $n=500$, the correlation between the strength of sexual selection α and the J

indexes was 0.99. For smaller sample size, $n=50$, the correlation was 0.97, indicating

that J_{PS1} and J_{PS2} indices are good proxies of the strength of sexual selection.

Table 1. Mating under a logistic model. The parameter α accounts for the intensity of
sexual selection. A value of $\alpha = 0$ corresponds to random mating. Mating is with
replacement. N is the population size and n the mating sample size. The quantitative
trait distribution is $Z(0,1)$. Number of runs: 10,000. % sig: The % of tests with p-value
≤0.05. The values in parentheses correspond to the average value of the statistic for the
total number of significant runs.

N	n	α	% sig. (J_{PS1})	%sig. (J_{PS2})
10^4	500	0	5.3 (0.016)	4.3 (0.016)
10^4	500	0.1	28.5.0 (0.019)	30.0 (0.019)
10^4	500	0.25	96.6 (0.039)	96.1 (0.038)
10^4	500	0.5	100 (0.126)	100 (0.128)
10^4	500	1	100 (0.40)	100 (0.41)
10^4	500	1.5	100 (0.710)	100 (0.701)
10^4	500	2	100 (0.981)	100 (0.985)
500	50	0	5.0 (0.146)	5.1 (0.141)
500	50	0.1	9.5 (0.155)	8.6 (0.15)
500	50	0.25	25.9 (0.168)	28.2 (0.179)
500	50	0.5	64.3 (0.215)	66.3 (0.219)

500	50	1	99.9 (0.586)	99.5 (0.476)
500	50	1.5	100 (0.845)	100 (0.918)
500	50	2	100 (1.107)	100 (1.044)

Detection of assortative mating: J_{PSI} versus correlation

We will utilize the program MateSim (Carvajal-Rodríguez 2018a) to simulate the

mating distribution for quantitative traits using a FND Gaussian function (Carvajal-

Rodríguez and Rolán-Alvarez 2014). The FND function with parameter $C>0$ generate

positive assortative mating patterns that can be measured by the Pearson correlation

coefficient ρ and also by J_{PSI} as given in (8).

The simulation procedure is as before but the preference is evaluated following the FND

function

$$FND = e^{-\left(\frac{C^2 D^2}{s^2 D^2_{max}}\right)}.$$

If $C=0$ the mating is random, if $C>0$ there is mate choice that generates positive

assortative mating. The parameter D represents the phenotypic difference between male

and female in the pair. If the D is zero the preference is the maximum. The parameter s

is the tolerance to mate with a partner that does not correspond to the preference

optimum. i.e. $s= 0.08$ implies eight times more tolerance for non-optimal mating than

$s=0.01$. D_{max} is the maximum phenotypic difference in the population between any two

individuals, regardless of sex. For a phenotypic difference D/D_{max} between a pair, the

strength of the choice depends on the ratio between the choice and tolerance parameters,

C/s.

For a sample of n matings, the significance of ρ can be computed as $\dfrac{\rho\sqrt{(n-2)}}{\sqrt{(1-\rho^2)}}$ that

follows a t-distribution with n-2 degrees of freedom. We have already seen that nJ_{PSI} is

asymptotically distributed as χ^2 with 1 degree of freedom.

In Table 2 both types of tests are performed for mating distributions generated under

different choice and tolerance values. When the population and sample size are large,

10,000 and 500, respectively, from choice values of 0.2, both the correlation and the J_{PSI}

have greater than 97% power to detect assortative mating. The difference is that the

correlation value already reaches its maximum from choice values of 0.4.

When the population and sample sizes are smaller, specifically 500 and 50, respectively,

we observe a few noteworthy points. Firstly, it is not surprising that J_{PSI} tends to be

somewhat liberal in these cases, as the test nJ_{PSI} is only asymptotically distributed as χ^2,

and its behavior improves with higher sample sizes. Secondly, both methods

demonstrate approximately 40% power when the choice is 0.2 and the tolerance is 0.08.

However, they exhibit almost 100% power in other cases, such as when the tolerance is

0.01 or the choice is from 0.4 and beyond. The difference between ρ and J_{PSI} is again the

saturation of the value of ρ as the strength of choice increases.

Table 2. One sex is choosy. Choice implies preference for similarity. Mating is with
replacement. N is the population size and n the sample size of matings. The quantitative
trait distribution is $Z(0,1)$. Number of runs: 10,000. % sig: The % of tests with p-value
≤ 0.05. The values in parentheses correspond to the average value of the statistic for the
total number of significant runs.

N	n	*Choice*	s	% sig. (ρ)	%sig. ($\rho^2/(1-\rho^2)$)
10^4	500	0	--	5.1 (0.01)	5.2 (0.01)
10^4	500	0.2	0.08	97.9 (0.18)	97.9 (0.03)
10^4	500	0.4	0.08	100 (0.55)	100 (0.43)
10^4	500	0.8	0.08	100 (0.85)	100 (2.60)

10^4	500	1	0.08	100 (0.89)	100 (3.72)
10^4	500	0.2	0.01	100 (0.95)	100 (9.3)
10^4	500	0.4	0.01	100 (0.99)	100 (59.0)
10^4	500	0.8	0.01	100 (1.0)	100 (194.0)
10^4	500	1	0.01	100 (1.0)	100 (413.4)
500	50	0	--	5.1 (0.03)	6.2 (0.12)
500	50	0.2	0.08	38.6 (0.37)	41.7 (0.16)
500	50	0.4	0.08	99.95 (0.61)	99.97 (0.64)
500	50	0.8	0.08	100 (0.90)	100 (4.59)
500	50	1	0.08	100 (0.93)	100 (6.68)
500	50	0.2	0.01	100 (0.96)	100 (13.54)
500	50	0.4	0.01	100 (0.99)	100 (56.68)
500	50	0.8	0.01	100 (1.0)	100 (183.08)
500	50	1	0.01	100 (1.0)	100 (403.27)

381   The greater the sample size and the choice strength, the greater the power to detect

382   assortative mating. In Figure 1 the correlation between the choice strength (C/s) and

383   both statistics can be appreciated. For the ρ statistic the correlation with choice strength

384   is 0.6 for both large and small sample size, for J_{PSI} the correlation with the choice

385   strength is 0.97.

Figure 1. Continuous line N=10,000, n=500. Dashed line N=500, n=50.

The results were almost identical (Appendix C Table S1) when we consider a bias for the mating preference, i.e. the choosy males preferring females 1.5 times bigger than themselves (Lau et al. 2021).

We were also interested in evaluating the performance of the J_{PSI} index across different functions and mating scenarios. To achieve this, we employed the logistic function with α =2 (which implies sexual selection without mate choice) with mating involving replacement, as well as individual and mass mating without replacement (see Appendix C Table S2). In the case of large population and sample sizes, both ρ and J_{PSI} exhibited the expected false positive rate of 5%. However, with lower sample sizes, J_{PSI} showed a slightly liberal false positive rate of 6%.

We also simulate a mate choice scenario for a bimodal population where two more or less extreme phenotypes coexist (Appendix C Table S3). The power for the detection of assortative mating was similar to the unimodal case and the correlation of the statistics with the strength of choice was greater the larger the sample size, being about 0.58-0.65 for ρ and 0.95-0.97 for J_{PSI} (Figure 2).

404 Figure 2. Bimodal population with mixture proportion $\pi=0.5$, $\mu_2-\mu_1=3$ and $\sigma_1+\sigma_2 = 2$.
405

406 Lastly, we examined how being choosy in one or both sexes affects the ρ and J_{PSI}

407 indexes and found no differences (compare Table S4 in Appendix C with Table 1).

5. Example of application

Echinolittorina malaccana is a marine intertidal snail commonly found on rocky shores

throughout the Indo-West Pacific region (Reid 2007). The primary mechanism of mate

choice in this species appears to be male preference for females slightly larger than

themselves. Assortative mating for size, measured by pairwise correlation, has been

estimated to be around 0.5 (Ng et al. 2019; López-Cortegano et al. 2020). In a previous

study, size (shell length, a continuous trait) has been transformed into a qualitative trait

(discrete size classes) enabling the use of multimodel inference techniques. These

analyses provided deeper insights into mate choice and confirmed that the most suitable

model involves mate choice with bias, resulting in both assortative mating and sexual

selection for size (Lau et al. 2021).

We will use the same onshore data from mating pairs of *E. malaccana* used in (Lau et

al. 2021) and available from (Ng et al. 2019; Rolán-Alvarez et al. 2021) to apply the

new indices J_{PSS} and J_{PSI} directly to the shell length trait (Table 3).

Table 3. Analysis of *Echinolittorina malaccana* data.

Test	N	n	J_{PS1} (pval)	J_{PS2} (pval)	ρ (pval)	J_{PSI} (pval)
Chi^2	876	228	0.11 ($<10^{-5}$)	0.03 (0.031)	0.57 ($<10^{-5}$)	0.49 ($<10^{-5}$)

423 Data from Ng et al 2019.
424
Sexual selection was observed in both sexes, although it appeared to be weaker in

males. This finding aligns with previous research, indicating that sexual selection in *E.*

malaccana females is likely driven by male preference for slightly larger females, while

in males, it may result from male-male competition (Ng et al. 2019; Lau et al. 2021).

As for assortative mating, both the Pearson correlation coefficient and the new J_{PSI}

index supported and validated previous findings.

6. Discussion

There is a vast amount of literature exploring mate choice, competition, assortative

mating, and sexual selection as influential forces in the evolution of traits that may not

always be directly favoured by natural selection (Darwin 1871; Andersson 1994;

reviewed in Lindsay et al. 2019). These concepts are traditionally grouped under the

term 'theory of sexual selection'. Sexual selection theory stands as one of the most active

areas of evolutionary research, engaging diverse fields and inference methodologies,

given its relevance in population genetics, evolutionary ecology, animal behavior,

sociology, and psychology. Two recent important reviews were published on the

concept and definition of sexual selection (Alonzo and Servedio 2019; Shuker and

Kvarnemo 2021). Alonzo and Servedio discuss the challenges in achieving a consensus

definition, while Shuker and Kvarnemo propose a new definition for sexual selection.

In this study, we demonstrate that the SK definition of sexual selection can be

effectively applied within a modeling framework that utilizes information theory to

describe mate choice and competition processes, leading to patterns of sexual selection

and assortative mating. Under the information theory formalism, these patterns are

described in terms of Jeffreys divergence.

Importantly, this novel framework, initially developed for discrete traits, is general and

doesn't assume a priori whether the choice is exercised by males, females, or both, nor

does it make assumptions about the sex that competes. Furthermore, it is agnostic

concerning the genetic basis or evolutionary mechanisms underlying complex mate

choice and competition processes. The informational framework starts with a concept of

mutual mating fitness or mutual mating propensity (sensu Carvajal-Rodríguez 2018b,

2020), where the marginal values correspond to individual mating fitness. These mating

fitnesses allow us to measure the pattern of sexual selection according to the SK

definition.

In addition, we have extended the informational framework to encompass quantitative

traits. For both discrete and quantitative traits, sexual selection is detected by comparing

the distribution of mating phenotypes with that of the overall population. Similarly,

assortative mating is detected by comparing the observed distribution in mated

individuals with the expected distribution under random mating of those same

individuals.

When the traits are normally distributed, the sexual selection indexes J_{PS1} (females) and

J_{PS2} (males) can be expressed as a function of the standardized sexual selection intensity

index and the ratio of sample and population variances. Simulations have shown that J_{PS}

indexes have a correlation of 0.99 with the strength of sexual selection.

Furthermore, the assortative mating index J_{PSI} possesses as much statistical power as ρ,

with the added advantage of being highly correlated with the strength of mate

preference, which is not generally true for the correlation coefficient (Clancey et al.

2022; Fernández-Meirama et al. 2022).

Finally, we applied the quantitative trait information indices developed in this work to a

previously published dataset on size assortative mating and sexual selection (Ng et al.

2019; Lau et al. 2021). We confirmed previous findings and validate the quantitative

approach for the new indices of sexual selection and assortative mating.

In conclusion, the application of information theory to describe patterns of sexual

selection and assortative mating (Carvajal-Rodríguez 2018b, 2019, 2020), has paved the

way for extending these developments to quantitative traits. This study demonstrates

that sexual selection and assortative mating can be effectively quantified as information

gain concerning the random mating pattern. Just like the previous developments for

discrete traits, the results obtained with quantitative traits indicate that employing

information theory to describe mating processes and patterns holds great promise as a

valuable approach to understanding both the causes and effects of sexual selection.

Declaration of competing interest

482   None

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Appendix

487 **A. Assortative mating with normally distributed traits**

488 The general formulation is

$$J_{PSI} = \int_{-a}^{+a} \int_{-a}^{+a} (q'(x,y) - r'(x,y)) \log \frac{q'(x,y)}{r'(x,y)} \, dx \, dy$$

490 where $r'(x,y) = f_1(x) f_2(y)$

491 If q' is bivariate normal and f_1 and f_2 are normally distributed we have

$$q'(x,y) = \frac{1}{|2\Pi \Sigma_1|^{0.5}} \exp\left[-\frac{1}{2}(z-\mu_1)' \Sigma_1^{-1}(z-\mu_1)\right]$$

493 where Σ_1 is the variance-covariance matrix (see below), $|\Sigma_1|$ is the determinant, z is the

494 column vector of the two variables x,y and z' is the row vector of x,y and μ_1 is the

495 column vector (μ_{1x}, μ_{1y}).

496 Similarly, the product of the two marginals is

$$r'(x,y) = \frac{1}{|2\Pi \Sigma_2|^{0.5}} \exp\left[-\frac{1}{2}(z-\mu_2)' \Sigma_2^{-1}(z-\mu_2)\right]$$

498 where μ_2 is the column vector (μ_{2x}, μ_{2y}).

499 Note that in our case, using males and females from matings (joint distribution versus

500 random mating distribution obtained from mating data) implies that mean and variances

501 are equal between distributions ($\mu_1 = \mu_2$, $\sigma_1 = \sigma_2$), i.e. $\mu_{1x} = \mu_{2x}$, $\mu_{1y} = \mu_{2y}$ and $\sigma_{1x} = \sigma_{2x}$, $\sigma_{1y} = \sigma_{2y}$.

502 Let

$$\Sigma_1 = \begin{pmatrix} \sigma_{xx} & \sigma_{xy} \\ \sigma_{xy} & \sigma_{yy} \end{pmatrix}, \quad \Sigma_1^{-1} = \frac{1}{S} \begin{pmatrix} \sigma_{yy} & -\sigma_{xy} \\ -\sigma_{xy} & \sigma_{xx} \end{pmatrix} \text{ with } S = \sigma_{xx}\sigma_{yy} - \sigma_{xy}^2$$

504 and

505
$$\Sigma_2 = \begin{pmatrix} \sigma_{xx} & 0 \\ 0 & \sigma_{yy} \end{pmatrix}, \quad \Sigma_2^{-1} = \frac{1}{\sigma_{xx}\sigma_{yy}} \begin{pmatrix} \sigma_{yy} & 0 \\ 0 & \sigma_{xx} \end{pmatrix}$$

506   In this case the Kullback-Leibler divergences are (c.f. Kullback 1997 (1.6) p. 190)

507
$$KL(q'\|r') = \frac{1}{2}\log\frac{|\Sigma_2|}{|\Sigma_1|} - 1 + \frac{1}{2}tr(\Sigma_1\Sigma_2^{-1}) = \frac{1}{2}\log\frac{\sigma_{xx}\sigma_{yy}}{S}$$

508   and

509
$$KL(r'\|q') = \frac{1}{2}\log\frac{|\Sigma_1|}{|\Sigma_2|} - 1 + \frac{1}{2}tr(\Sigma_2\Sigma_1^{-1}) = \frac{1}{2}\log\frac{S}{\sigma_{xx}\sigma_{yy}} - 1 + \frac{\sigma_{xx}\sigma_{yy}}{S}$$

510   where tr is the trace (sum of diagonal elements) of a square matrix.

511   Therefore,

512
$$J_{PSI} = KL(q'\|r') + KL(r'\|q') = \frac{\sigma_{xx}\sigma_{yy}}{S} - 1 = \frac{\sigma_{xy}^2}{\sigma_{xx}\sigma_{yy} - \sigma_{xy}^2} = \frac{\rho^2}{1-\rho^2}.$$

B. General formula for non-random mating information

Consider two continuous traits, X in females and Y in males from a population and let

the probability distribution Q'(x,y) with probability density $q'(x,y)$ that represents the

mating probability for pairs with values in the infinitesimal interval $([x,x+dx],[y,y+dy])$.

Consider alternatively the probability distribution $Q(x,y)$ of random mating with the

product of densities $q(x,y)=f(x)g(y)$ from X and Y, respectively. The divergence between

the two distributions Q' and Q is given by

521
$$J=\int_{-a}^{+a}\int_{-a}^{+a}(q'(x,y)-q(x,y))\log\frac{q'(x,y)}{q(x,y)}dxdy, \quad \text{where } q(x,y)=f(x)g(y) \quad \text{(B1)}.$$

522   The difference $q'(x,y)-q(x,y)$ can be partitioned as follows

523
$$q'(x,y)-q(x,y)=q'(x,y)-f_1(x)f_2(y)+f_1(x)f_2(y)-q(x,y),$$

524   where $f_1(x)$ and $f_2(y)$ are the marginal densities for X and Y

525
$$f_1(x)=\int_{-a}^{+a}q'(x,y)dy$$

526
$$f_2(y)=\int_{-a}^{+a}q'(x,y)dx \qquad \text{(B2)}.$$

527   Similarly, the logarithm

528
$$\log\frac{q'(x,y)}{q(x,y)}=\log(\frac{q'(x,y)}{f_1(x)f_2(y)}\frac{f_1(x)f_2(y)}{q(x,y)})=\log\frac{q'(x,y)}{f_1(x)f_2(y)}+\log\frac{f_1(x)f_2(y)}{q(x,y)}.$$

529   Now note that the densities $f(x)$ and $g(y)$ depend on the population distribution, while

530   the marginal densities $f_1(x)$ and $f_2(y)$ depend on the mating distribution so they are not

531   necessarily equal. Then we have

532
$$J_{PSS}=\int_{-a}^{+a}\int_{-a}^{+a}(f_1(x)f_2(y)-q(x,y))\log\frac{f_1(x)f_2(y)}{q(x,y)}dxdy$$

and

$$J_{PSI}=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}(q'(x,y)-f_1(x)f_2(y))\log\frac{q'(x,y)}{f_1(x)f_2(y)}dxdy$$

therefore, as in the discrete case, the measure of non-random mating can be divided into

$$J=J_{PSS}+J_{PSI}+E \qquad (B3).$$

The component E is

$$E=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}(A+B)dxdy=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}Adxdy+\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}Bdxdy=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}Adxdy$$

$$A=(f_1(x)f_2(y)-q(x,y))\log\frac{q'(x,y)}{f_1(x)f_2(y)}$$

$$B=(q'(x,y)-f_1(x)f_2(y))\log\frac{f_1(x)f_2(y)}{q(x,y)}.$$

We will show that

$$\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}Bdxdy=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}[(q'(x,y)-f_1(x)f_2(y))\log\frac{f_1(x)f_2(y)}{q(x,y)}]dxdy=0 \qquad (B4)$$

so that,

$$E=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}(f_1(x)f_2(y)-q(x,y))\log\frac{q'(x,y)}{f_1(x)f_2(y)}\neq0\Leftrightarrow J_{PSS}\neq0\wedge J_{PSI}\neq0 \qquad (B5).$$

Now, first recall that $q(x,y)=f(x)g(y)$ and then

$$\log\frac{f_1(x)f_2(y)}{q(x,y)}=\log\frac{f_1(x)}{f(x)}+\log\frac{f_2(y)}{g(y)}.$$

Also,

$$\int\limits_{-a}^{+a}f_1(x)dx=\int\limits_{-a}^{+a}\int\limits_{-a}^{+a}q'(x,y)dxdy=\int\limits_{-a}^{+a}f_2(y)dy=1$$

so that,

$$\int_{-a}^{+a}\int_{-a}^{+a}[f_1(x)f_2(y)\log\frac{f_1(x)}{f(x)}]dxdy=\int_{-a}^{+a}f_1(x)\log\frac{f_1(x)}{f(x)}dx\int_{-a}^{+a}f_2(y)dy=\int_{-a}^{+a}f_1(x)\log\frac{f_1(x)}{f(x)}dx$$

551   (B6),

552   and similarly

$$\int_{-a}^{+a}\int_{-a}^{+a}[f_1(x)f_2(y)\log\frac{f_2(y)}{g(y)}]dxdy=\int_{-a}^{+a}f_2(y)\log\frac{f_2(y)}{g(y)}dy \quad (B7).$$

554   Also using the marginal density definitions from (A2)

$$\int_{-a}^{+a}\int_{-a}^{+a}[q'(x,y)\log\frac{f_1(x)}{f(x)}]dxdy=\int_{-a}^{+a}\log\frac{f_1(x)}{f(x)}dx\int_{-a}^{+a}q'(x,y)dy=\int_{-a}^{+a}f_1(x)\log\frac{f_1(x)}{f(x)}dx$$

556   (B8)

557   and

$$\int_{-a}^{+a}\int_{-a}^{+a}[q'(x,y)\log\frac{f_2(y)}{g(y)}]dxdy=\int_{-a}^{+a}f_2(y)\log\frac{f_2(y)}{g(y)}dy \quad\quad (B9).$$

559   Finally, noting that (B6) equals (B8) and (B7) equals (B9) and rearranging and

560   substituting in (B4) we obtain

$$\int_{-a}^{+a}\int_{-a}^{+a}Bdxdy=\int_{-a}^{+a}\int_{-a}^{+a}[(q'(x,y)-f_1(x)f_2(y))\log\frac{f_1(x)f_2(y)}{q(x,y)}]dxdy=(A6)+(A7)-(A8)-(A9)=0$$

C. Simulation results tables

563 Table S1. One sex is choosy. Choice implies preference for similarity with a bias of 1.5.
564 Mating is with replacement. N is the population size and n the sample size of matings.
565 The quantitative trait distribution is $Z(0,1)$. Number of runs: 10,000. % sig: The % of
566 tests with p-value ≤ 0.05. The values in parentheses correspond to the average value of
567 the statistic for the total number of significant runs.

N	n	*Choice*	s	% sig. (ρ)	%sig. ($\rho^2/(1-\rho^2)$)
10^4	500	0	--	4.8 (0.004)	4.9 (0.01)

10^4	500	0.2	0.08	99.9 (0.22)	99.9 (0.05)
10^4	500	0.4	0.08	100 (0.54)	100 (0.41)
10^4	500	0.8	0.08	100 (0.87)	100 (3.19)
10^4	500	1	0.08	100 (0.89)	100 (3.99)
500	50	0	--	5 (-0.025)	5.9 (0.12)
500	50	0.2	0.08	44.7 (0.37)	48.2 (0.17)
500	50	0.4	0.08	100 (0.67)	100 (0.90)
500	50	0.8	0.08	100 (0.84)	100 (2.54)
500	50	1	0.08	100 (0.90)	100 (4.43)
500	50	0.2	0.01	100 (0.97)	100 (64.72)
500	50	0.4	0.01	100 (0.99)	100 (41.97)
500	50	0.8	0.01	100 (1)	100 (115.87)
500	50	1	0.01	100 (1)	100 (3935.1)

569 Table S2. Mating under a logistic model with $\alpha=2$. There is no preference for similarity.
570 N is the population size and n the sample size of matings. The quantitative trait
571 distribution is $Z(0,1)$. Number of runs: 10,000. % sig: The % of tests with p-value
572 ≤ 0.05. The values in parentheses correspond to the average value of the statistic for the
573 total number of significant runs.

N	n	Mating	% sig. (ρ)	%sig. ($\rho^2/(1-\rho^2)$)
10^4	500	replacement	5.3 (0.003)	5.4 (0.01)
10^4	500	individual	4.9 (0.003)	5.0 (0.01)
10^4	500	mass-encounter	4.7 (0.004)	4.8 (0.01)
500	50	replacement	5.0 (-0.01)	6.0 (0.12)
500	50	individual	5.2 (0.02)	6.3 (0.12)
500	50	mass-encounter	5.4 (0.03)	6.4 (0.12)

Table S3. Bimodal populations with mixture proportion $\pi=0.5$, $\mu_2-\mu_1=3$ and $\sigma_1+\sigma_2 = 2$.
One sex is choosy. Choice implies preference for similarity. Mating is with
replacement. N is the population size and n the sample size of matings. The quantitative
trait distribution is $Z(0,1)$. Number of runs: 10,000. % sig: The % of tests with p-value
≤ 0.05. The values in parentheses correspond to the average value of the statistic for the
total number of significant runs.

N	n	*Choice*	s	% sig. (ρ)	%sig. ($\rho^2/(1-\rho^2)$)
10^4	500	0	--	5.1 (0.004)	5.2 (0.01)
10^4	500	0.2	0.08	90 (0.15)	90.1 (0.03)
10^4	500	0.4	0.08	100 (0.43)	100 (0.22)
10^4	500	0.8	0.08	100 (0.82)	100 (1.98)
10^4	500	1	0.08	100 (0.89)	100 (3.79)
10^4	500	0.2	0.01	100 (0.95)	100 (9.99)
10^4	500	0.4	0.01	100 (0.99)	100 (48.67)
10^4	500	0.8	0.01	100 (1.0)	100 (242.35)
10^4	500	1	0.01	100 (1.0)	100 (397.16)
500	50	0	--	5.1 (-0.01)	6.1 (0.12)
500	50	0.2	0.08	43.2 (0.38)	46.2 (0.18)
500	50	0.4	0.08	99.7 (0.59)	99.7 (0.58)
500	50	0.8	0.08	100 (0.93)	100 (6.58)
500	50	1	0.08	100 (0.94)	100 (7.74)
500	50	0.2	0.01	100 (0.97)	100 (18.57)
500	50	0.4	0.01	100 (0.99)	100 (47.17)
500	50	0.8	0.01	100 (1.0)	100 (177.0)
500	50	1	0.01	100 (1.0)	100 (324.56)

582 Table S4. Both sexes are choosy. Mating with replacement. N is the population size and
583 n the sample size of matings. Choice implies preference for similarity. The quantitative
584 trait distribution is $Z(0,1)$. Number of runs: 1,000. % sig: The % of tests with p-value

585 ≤0.05. The values in parentheses correspond to the average value of the statistic for the
586 total number of significant runs.

N	n	Choice	s	% sig. (ρ)	%sig. ($\rho^2/(1-\rho^2)$)
10^4	500	0	--	4.4 (0.03)	5.4 (0.06)
10^4	500	0.2	0.08	24.6 (0.25)	26.1 (0.07)
10^4	500	0.4	0.08	100 (0.47)	100 (0.3)
10^4	500	0.8	0.08	100 (0.75)	100 (1.39)
10^4	500	0.2	0.01	100 (0.93)	100 (7)
10^4	500	0.4	0.01	100 (0.98)	100 (26)
10^4	500	0.8	0.01	100 (1)	100 (106)
500	50	0	--	4.5 (0.017)	5.7 (0.12)
500	50	0.2	0.08	36.3 (0.36)	38.5 (0.16)
500	50	0.4	0.08	99.8 (0.58)	99.9 (0.57)
500	50	0.8	0.08	100 (0.81)	100 (2.1)
500	50	0.2	0.01	100 (0.96)	100 (12.7)
500	50	0.4	0.01	100 (0.99)	100 (56.7)
500	50	0.8	0.01	100 (1.0)	100 (206)

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