

22 Supplementary Discussion

23 Cultural analogues of the Fifth Commandment

24 Cultural norms that promote the help of the parents are widespread in both western and eastern 25 culture. The Fifth Commandment (of the Hebrew and protestant Bible, the Fourth one, 26 according to the catholic numbering) states: "Honor your father and your mother, that your 27 days may be long in the land that the LORD your God is giving you." (Exodus 20:12)

28 From the interpretations of this commandment by the western churches we recall the following: 29 Sefer Ha-chinukh (mitzva 33) elaborates: "A person should realize that his father and mother 30 are the cause of his existence in this world; therefore it is appropriate that he render them all 31 the honor and do them all the service he can". St. Thomas Aquinas wrote: "Since we receive 32 nourishment from our parents in our childhood, we must support them in their old age." Martin 33 Luther said: "For he who knows how to regard them in his heart will not allow them to suffer 34 want or hunger, but will place them above him and at his side, and will share with them 35 whatever he has and possesses" (Luther, M. p. 29).

36 We also note that in China, to take care of elderly parents is also a moral rule: e.g. Confucius 37 declared: "In serving his parents, a filial son reveres them in daily life; he makes them happy 38 while he nourishes them; he takes anxious care of them in sickness …" (26)

39 Based on the above, we introduce the so-called Fifth Rule, which is a translation of the Fifth 40 Commandment into biological terms and is inherent in the above interpretations: "During your 41 reproductive period, give away from your resources to your post-fertile parents."

43 Supplementary Methods

44 In the following we investigate a biological model for intra-familiar help, give some general 45 results on it, and finally, we investigate the simplest model with numerical examples. 46

47 1. The survival at the carrying capacity

48 Here we propose a strictly Darwinian reasoning to see that the long-term growth rate is 49 maximized by natural selection: the number of offspring, in general, is much higher than the 50 carrying capacity, so only a part of the offspring and adults will survive. Let us consider random 51 survival, assuming that the survival probabilities of individuals do not depend on phenotypes 52 (in our case intergenerational help) and on the age of individuals. (Observe that this assumption 53 gives some advantage to the families in which the intergenerational help is less.)

54 Now let us consider two phenotypes A and B with respective long-term growth rates (i.e. 55 positive eigenvalues of the corresponding Leslie matrices) λ_A , $\lambda_B > 1$ with λ_A>λ_B. To see the 56 asymptotic frequency of phenotype B, we suppose that phenotypes A and B start from 57 respective initial densities $x(0)$ and $z(0)$. According to the original Darwinian view, we need 58 some density dependent selection to keep the total density of these two phenotypes at the 59 carrying capacity. Since in the considered selection situation there is no interaction between the 60 phenotypes and we assume that the phenotypes differ only in the demographic parameters, thus 61 we can suppose there is a uniform survival process, i.e. the survival rate corresponding to the 62 carrying capacity is the same for all individuals. Now the question arises which phenotype will 63 win in the struggle for existence on the long selection time scale?

64 Let us suppose that phenotypes A and B develop according to Leslie models having the 65 respective population vectors $x(t)$, $z(t)$, and matrices L_A , L_B , total densities

66
$$
||x(t)|| = \sum_{i} x_i(t), ||z(t)|| = \sum_{i} z_i(t)
$$
. Then the relative frequency of phenotype B tends to zero, as

67 it is shown below:

68 Indeed, let us suppose that the subpopulations start from initial states $x(0)$ and $z(0)$, 69 respectively, and the time unit is chosen in such a way that in unit time the total density of the 70 system always exceeds the carrying capacity K , in particular

71
$$
||L_A x(0)|| + ||L_B z(0)|| > K.
$$
 (SI 2)

72 Now, by the selection the total density of the system is reduced to K proportionally:

73
$$
x(1) = \frac{K}{\|L_A x(0)\| + \|L_B z(0)\|} L_1 x(0), \quad \text{(SI 3)}
$$

74
$$
z(1) = \frac{K}{\|L_A x(0)\| + \|L_B z(0)\|} L_B z(0). \quad (SI 4)
$$

75 Indeed, obviously $||x(1)|| + ||z(1)|| = K$.

76 We emphasize that in this model we consider the "intrinsic" survival (described by the Leslie 77 matrices) and the survival under selection independently. However, this model can be formally 78 considered as a particular Leslie-type model depending on the total density of the system, where 79 each demographic parameter in the Leslie matrices L_A and L_B is multiplied by

80
$$
\frac{K}{\|L_A x(0)\| + \|L_B z(0)\|}.
$$

81 Similarly, for all $t= 1, 2, 3,...$ we get our kin demographic selection model for two different 82 phenotypes:

83

$$
x(t+1) = \frac{K}{\|L_{A}x(t)\| + \|L_{B}z(t)\|} L_{A}x(t)
$$

$$
z(t+1) = \frac{K}{\|L_{A}x(t)\| + \|L_{B}z(t)\|} L_{B}z(t)
$$
(SI 5)

85 Now, for the proportion of phenotype B we obtain

86
$$
\frac{\|z(t)\|}{\|x(t)\| + \|z(t)\|} = \frac{1}{\frac{\|x(t)\|}{\|z(t)\|} + 1}.
$$
 (SI 6)

87 Here

88
$$
\frac{\|x(t)\|}{\|z(t)\|} = \left(\frac{\lambda_{\mathbf{A}}}{\lambda_{\mathbf{B}}}\right)^t \frac{\|x(t)\|}{\|z(t)\|}.
$$
 (SI 7)

89 Since we can suppose that in both phenotypes the last two fecundities are positive, so the Perron-Frobenius theorem (see e.g. 28) implies that both $\frac{dP}{dt}$ $x(t)$ A (t) $\frac{\partial (t)_{\parallel}}{\partial t}$ and $\frac{\parallel 2(t)}{\parallel}$ $z(t)$ B (t) 90 Perron-Frobenius theorem (see e.g. 28) implies that both $\frac{\mathbb{P}^{(1)}[f]}{\lambda}$ and $\frac{\mathbb{P}^{(1)}[f]}{\lambda}$ tend to finite 91 positive limits as $t \to \infty$. In fact, the Leslie matrices can be cut at the last fertile age class, apply 92 the Perron-Frobenius theorem to these matrices, and then the convergence can be extended to the post fertile age groups by simple survival Therefore, $\lim_{n\to\infty} \frac{||x(t)||}{||x(t)||} = \infty$ $\rightarrow \infty |z(t)|$ (t) lim $z(t)$ $x(t)$ 93 the post fertile age groups by simple survival Therefore, $\lim_{t\to\infty} \frac{\|f(t)\|}{\|g(t)\|} = \infty$, implying

94
$$
\lim_{t \to \infty} \frac{\|z(t)\|}{\|x(t)\| + \|z(t)\|} = 0. \text{ (SI 8)}
$$

95 Thus if $\lambda_A > \lambda_B$, then the relative frequency of phenotype B tends to zero as t tends to infinity. 96 Observe that in our model, the fecundity of a phenotype is determined by a phenotype-97 dependent Leslie matrix, and the survival rates corresponding to the carrying capacity of 98 different phenotypes are the same, so the long-term growth rate of a phenotype determines the 99 fitness.

101 2. The general results

102 Consider the general $K \times K$ Leslie matrix, where the entries depend on the cost y spent to 103 grandparent support. Under the grandmother hypothesis, the grandmother support decreases the 104 fecundity and survival rate of fertile parents, but increases the survival rate of the grandmother, 105 who therefore increases the survival rate of pre-fertile grandchildren. Then the characteristic 106 equation is

107
$$
\lambda^{K} - \alpha_{k+1}(y) \prod_{i=1}^{k} \omega_{i}(y) \lambda^{K-(k+1)} - \alpha_{k+2}(y) \prod_{i=1}^{k+1} \omega_{i}(y) \lambda^{K-(k+2)} - \dots - \alpha_{K}(y) \prod_{i=1}^{K-1} \omega_{i}(y) = 0,
$$
 (SI 9)

108 and its unique positive root is obtained as the root of equation

109
$$
q(\lambda) = \frac{\alpha_{k+1}(y) \prod_{i=1}^{k} \omega_i(y)}{\lambda^{k+1}} + \frac{\alpha_{k+2}(y) \prod_{i=1}^{k+1} \omega_i(y)}{\lambda^{k+2}} + \dots + \frac{\alpha_{K}(y) \prod_{i=1}^{K-1} \omega_i(y)}{\lambda^{K}} = 1.
$$
 (SI 10)

110 It is easy to see, that if any of the numerators (i.e. the average numbers of offspring produced 111 by an individual of the corresponding age classes) in these fractions is changed to a greater one, 112 then the curve of the 'hyperbolic' function q shifts upwards, implying that the positive solution 113 λ of this equation also will be greater. Therefore, if in a population where within the families 114 grandparents are not supported, a new type emerges which supports grandparents, and all 115 mentioned numerators increase, then Fifth Rule as behaviour type will propagate. If all these 116 numerators decrease then this type will die out. Those mathematical cases when some of the 117 numerators increase, others decrease, would need further mathematical discussions.

118

119 Observe that equation $q(\lambda) = 1$ can be written as

120
$$
\prod_{i=1}^{k} \omega_i(y) \left(\frac{\alpha_{k+1}(y)}{\lambda^{k+1}} + \frac{\alpha_{k+2}(y)\omega_{k+1}(y)}{\lambda^{k+2}} + \dots + \frac{\alpha_{K}(y)\prod_{j=k+1}^{K-1}\omega_j(y)}{\lambda^{K}} \right) = 1.
$$
 (SI 11)

Here factor $\prod_{i=1}^{k}$ i $\partial_i(y)$ 1 121 Here factor $\iint \omega_i(y)$ measures how much child care by grandmothers increases the survival

122 of the children. Roughly speaking, factor

123
$$
\frac{\alpha_{k+1}(y)}{\lambda^{k+1}} + \frac{\alpha_{k+2}(y)\omega_{k+1}(y)}{\lambda^{k+2}} + \dots + \frac{\alpha_{k}(y)\prod_{j=k+1}^{K-1}\omega_{j}(y)}{\lambda^{K}}
$$
 (SI 12)

124 measures, in an implicit way, to what extent the support to grandparents by the fertile age class 125 decreases their own fecundity and survival rates. In this sense, the predictions of our model are 126 in harmony with the cost-benefit approach saying that a trait will propagate if it eventually 127 increases the fitness.

128

129 Now the question arises how the demographic parameters may depend on y. The 130 following assumptions are at hand: 1. The survival rate of grandparents is a saturation function 131 of y strictly increasing at the beginning, and remains constant after. 2. Based on the grandmother 132 hypothesis, the survival rate of grandchildren strictly increases with the survival rate of 133 grandparents (which on its term depend on y). The grandmother hypothesis is the worst case 134 when two trade-offs may exist. 3. The parents' fecundity entries of the Leslie matrix (135 $\alpha_{k+1},...,\alpha_{k}$ are strictly decreasing functions of y. 4. The survival rates of parents ($\omega_{k+1},...,\omega_{k}$) 136 are strictly decreasing functions of v .

137 These assumptions allow the Fifth Rule to win or lose the struggle for existence, 138 depending on whether the long-term growth rate of the family increases or decreases.

139

140 Under assumption 1 there is a threshold for the support to grandparents, above which 141 the survival of grandparents does not increase, and therefore the survival of grandchildren

142 either, but the fecundity and/or the survival of fertile parents still decrease. Over this threshold, 143 the support to grandparents has no evolutionary advantage.

144

145 Finally, we remark that the above reasoning can be applied not only to the grandmother 146 hypothesis, since either the mother hypothesis or the embodied capital model alone can ensure 147 the support to grandparents. For example, if any of the above two hypotheses implies the 148 increase of at least one of the numerators in (3), while the rest of the numerators do not decrease, 149 then the dominant eigenvalue, i.e. the asymptotic growth rate will increase. Of course, if in 150 addition to the fact that the grandmother increases the survival of her grandchildren and the 151 survival and fertility of her daughter, the hypothesis of the embodied capital model also holds 152 (the grandmother also increases the adult age survival and fertility of her grandchildren), then not only $\prod_{i=1}^k$ i $\hat{y}_i(y)$ 1 153 not only $\left| \right| \omega_i(y)$ but also factor (4) can synergically increase the asymptotic growth rate of

154 the family.

155 Finally, we note that the "altriciality" hypothesis can also be handled in terms of a 156 linear model with a matrix structured differently from the Leslie matrices (since the survivals 157 of children also depend on the age of their mothers). Thus, only a generalization our model 158 could deal with the development of menopause based on altriciality. In our opinion, our Fifth 159 Rule may be derived on the bases of "altriciality" hypothesis, but in such a future model the 160 formation of multi-generation families should also be included, since "altriciality" hypothesis 161 itself does not need the convivence of several generations.

163 3. Two-age-class model

164 For a deeper insight, in this simplest case, we will calculate first when the menopause can 165 evolve, second, when the Fifth Rule is evolutionary successful, third, using numerical 166 examples we demonstrate that convex benefit and concave cost functions promote the 167 evolution of intra-familiar help.

168

169 Consider the simplest case with one child age class and one fertile age class. Then the 170 Leslie matrix is

$$
\begin{pmatrix} 0 & \alpha_0 \\ \omega_1 & 0 \end{pmatrix},
$$

172 where ω_1 is the survival rate of children and α_0 is the fecundity of fertile parents. The survival 173 rate from fertile age to post-fertile age is ω_2 , and ω_3 denotes the probability that a post-fertile 174 individual still lives (without the support by a fertile individual) when child care is needed. Now 175 the fitness is $\lambda = \sqrt{\alpha_0 \omega_1}$.

176

177 3.1 Grandmother hypothesis

178 Now the question arises: When is the menopause adaptive? Consider the case when fertile 179 individuals do not support grandmothers. We consider the following two cases: (i) Suppose 180 that grandmothers do not help in child care, but their survival linearly reduces their own 181 fecundity, i.e. $\omega_2(s) := s$ and $\alpha(s) := \alpha_0(1-s)$, where $s \in [0,1)$ is the cost spent on survival to 182 post-fertile age (Fig. 3 depicts the situation). The fitness of the population is the long-term 183 growth rate which can be calculated from the characteristic equation of the Leslie matrix:

184 $\lambda(s) = \sqrt{\alpha_0 (1-s) \omega_1}$, and the optimal strategy is not to spend on own survival to post-fertile 185 age. (ii) Suppose that grandmothers help in child care (Figure S 2.b). Let $\omega_2(s) = \overline{\omega}_2 s$, with 186 some $\overline{\omega}_2$, $\alpha(s) = \alpha_0(1-s)$ and $\omega_1(s) = \omega_1 + \alpha_{21}P(s)$, where $s \in [0,1)$, ω_1 is a "basic" 187 survival rate, and the probability that a grandmother is alive when her help needed is 188 $P(s) = \overline{\omega}_2 \omega_3 s$, i.e. we count only the help of those grandmothers who survive to the upper 189 boundary of the third age class and do not count those who reach 'grandmother age' (reach 190 the third class) but die before the upper boundary of age, and a_{21} denotes the efficiency of the 191 grandmother's grandchild care. Clearly $a_{21}\overline{\omega}_2\omega_3 < 1$ and $\omega_1(s) = \overline{\omega}_1 + a_{21}\overline{\omega}_2\omega_3 s$, thus the 192 fitness is

193
$$
\lambda(s) = \sqrt{\alpha_0 \left(1 - s\right) \left(\overline{\omega}_1 + a_{21} \overline{\omega}_2 \omega_3 s\right)},
$$
 (SI 13)

which is maximal at $_{21}\omega_2\omega_3$ $21\omega_2\omega_3$ ω_1 $2a_{21}\overline{\omega}_2\omega_3$ $\omega_2\omega_2-\omega_1$ a $s^* = \frac{a_1}{a_2}$ 194 which is maximal at $s^* = \frac{a_{21}\overline{\omega}_2 \omega_3 - \overline{\omega}_1}{2}$. Therefore, if the effect of grandchild care on the

195 grandchild's survival is greater than his/her survival rate without this care, i.e. $a_{21}\overline{\omega}_2 \omega_3 > \overline{\omega}_1$, 196 then menopause is evolutionarily successful.

197

198

199 3.2 The Fifth Rule

200 Now the question arises: When is Fifth Rule adaptive? It requires us to support our elderly,

201 which is possible only if the menopause has already become evolutionarily fixed, i.e. for fixed

202 $s \in [0,1)$, let $\omega_2 := \overline{\omega}_2 s$ and $\alpha := \alpha_0(1-s)$. Let $y \in [0,1]$ denote the cost spent on the survival of

- 203 post-fertile parents, and suppose that the negative effect of y on fecundity is linear: $\alpha(1 y)$,
- 204 the children survival is $\omega_1(y) := \overline{\omega}_1 + aP(y)$, where $P(y) := \omega_2(\omega_3 + by)$ and b indicates how

205 efficiently the support to post-fertile parents by fertile individuals increases post-fertile 206 survival, so $\omega_1(y) = \overline{\omega_1} + a\omega_2(\omega_2 + by)$ (Fig. 4 depicts the situation). Now we have to 207 maximize the fitness which can be calculated from the characteristic equation of the Leslie 208 matrix, it is given by the following function in y .

209
$$
\lambda(y) = \sqrt{\alpha(1-y)(\overline{\omega}_1 + a\omega_2(\omega_3 + by))},
$$
 (SI 14)

which attains its maximum at $y^* = \frac{a\omega_2(b-\omega_3)-\overline{\omega_1}}{2}$. The latter is po 2 $2(\theta - \omega_3)$ ω_1 $2ab\omega_2$ ω_{2} ω_{2} ω_{3} ω_{1} ab $a\omega_{2}(b)$ \mathcal{Y} $-\omega_1$)-210 which attains its maximum at $y^* = \frac{a\omega_2(\theta - \omega_3) - \omega_1}{2}$. The latter is positive if

211 $a\omega_2(b-\omega_3)-\overline{\omega}_1>0$. This condition is satisfied e.g., if the efficiency of the support to post-

212 fertile parents is sufficiently large compared to the basic post-fertile survival rate.

213

214

215 3.3 A general multiplicative coevolution model

216 Now we set up a model combining the model of grandmother hypothesis and the 217 model of the Fifth Rule. Our study will be based on two biological preconditions: First, since 218 one can help a grandmother only if she is alive, for the development of the Fifth Rule, the 219 existence of menopause is needed. Second, if a fertile mother gave away all her resources to 220 help the survival of her mother, her fecundity would be zero. As before, let s be the cost a 221 fertile female spends on her own survival to post-fertile age, and γ the cost a fertile female 222 spends on the survival of post-fertile parents. Based on the first precondition, unlike the 223 additive approach of sections 3.2 and 3.3, we express the effect of strategies s and y on the 224 demographic parameters in multiplicative form, considering the following strategy-dependent 225 Leslie matrix:

226
$$
L(s,y) := \begin{pmatrix} 0 & \alpha_0 \alpha(s,y) \\ \overline{\omega}_1 P(s,y) & 0 \end{pmatrix}, \quad (SI 15)
$$

246 [0,1] \times [0,1]. Indeed, maximization of $\lambda(s, y)$ is equivalent to the maximization of

247
$$
z(s, y) = \overline{\omega}_1 \alpha_0 p(s) q(y) \beta(s) \gamma(y) \quad (s, y \in [0, 1]).
$$
 (SI 18)

248 The first order necessary condition for the maximum attained at an interior point is

248 The first order necessary condition for the maximum attained at an interior point is
\n249
$$
\frac{\partial}{\partial s} z(s, y) = q(y) \gamma(y) (\beta(s) p'(s) + p(s) \beta'(s)) = 0, \text{ (SI 19)}
$$
\n250
$$
\frac{\partial}{\partial y} z(s, y) = p(s) \beta(s) (\gamma(y) q'(y) + q(y) \gamma'(y)) = 0 \text{ (SI 20)}
$$
\n251 Since *p*, *q*, *β*, *γ* are all positive in the interval (0,1), the above necessary condition is

250
$$
\frac{\partial}{\partial y}z(s,y) = p(s)\beta(s)(\gamma(y)q'(y) + q(y)\gamma'(y)) = 0 \quad \text{(SI 20)}
$$

251 Since p, q, β , γ are all positive in the interval (0,1), the above necessary condition is 252 equivalent to

253
$$
\varphi(s) := \beta(s) p'(s) + p(s) \beta'(s) = 0, \quad (S1 21)
$$

254
$$
\psi(y) = \gamma(y)q'(y) + q(y)\gamma'(y) = 0. \quad (S1 22)
$$

255 From conditions b), c) and d) we obtain $\varphi(0) > 0$, $\varphi(1) < 0$, hence there is an $s^* \in (0,1)$ with

 $256 \quad \varphi(s^*) = 0$. It is easy to check that conditions b), c), d) and e) also imply $\varphi'(s) < 0$, and hence φ is

257 strictly decreasing, therefore s^* is its unique zero in the interval. (0,1). Similar straightforward

258 checking shows that $\psi(y)$ also has a unique zero y^{*} in the interval (0,1). Hence (s^*, y^*) is a

259 unique stationary point of function \overline{z} in the interior of the unit square.

260 Now, for a second order sufficient condition for the maximum of function z , we calculate its 261 Hessian: tions b), c) and d) we obtain $\varphi(0) > 0$, $\varphi(1) < 0$, hence there is an $s^* \in (0,1)$ with
is easy to check that conditions b), c), d) and e) also imply $\varphi'(s) < 0$, and hence φ is
reasing, therefore s^* is its uni

262

263
$$
H(s,y) = \begin{pmatrix} q(y)\gamma(y)(\beta(s)p''(s) + 2p'(s)\beta'(s) + p(s)\beta''(s)) & (\beta(s)p'(s) + p(s)\beta'(s))(y'(y) + q(y)\gamma'(y)) \\ (\beta(s)p'(s) + p(s)\beta'(s))(y(y)q'(y) + q(y)\gamma'(y)) & p(s)\beta(s)(y(y)q''(y) + 2q'(y)\gamma'(y) + q(y)\gamma''(y)) \end{pmatrix},
$$

$$
264 \t\t(SI 23)
$$

265 Observe that from $\varphi(s^*) = 0$ and $\psi(y^*) = 0$, we obtain

 0 () () () () 2 () () () () () () () () 2 () () () () 0 (,) p s s y q y q y y q y y ^q ^y ^y ^s ^p ^s ^p ^s ^s ^p ^s ^s ^H ^s ^y 266

267 From conditions a)-e), we easily get

268
$$
q(y^*)\gamma(y^*)\big(\beta(s^*)p''(s^*)+2p'(s^*)\beta'(s^*)+p(s^*)\beta''(s^*)\big)<0 \qquad (\text{SI } 24)
$$

269
$$
p(s^*)\beta(s^*)\left(\gamma(y^*)q''(y^*)+2q'(y^*)\gamma'(y^*)+q(y^*)\gamma''(y^*)\right)<0
$$
, (SI 25)

270 implyig $tr H(s^*, y^*) < 0$ and $det H(s^*, y^*) > 0$, i.e. $H(s^*, y^*)$ is negative definite. Therefore (s^*, y^*) 271 is a strict local maximum point. Since (s^*, y^*) is the unique stationary point, it is also a strict 272 global maximum point in the interior of the unit square.

273 Finally, we note that, if Hessian $H(s, y)$ is negative definite in the interior of the unit 274 square, then function λ is globally strictly concave, and therefore (s^*, y^*) is a global 275 maximum point of λ . In the terminology of fitness landscapes, in the sense of any reasonable 276 strategy dynamics the species will evolve into the evolutionarily optimal behavior (s^*, y^*) . 277

278 3.4 Numerical Examples

279 In this section, by numerical study, we illustrate the effect of different (linear, convex and 280 concave) trade-offs on the level of the optimal backward help (v^*) . We calculated the 281 maximal long-term growth rate (fitness) of various populations as a function of y from the 282 characteristic equation of the corresponding Leslie matrix. The value of ν that gives the 283 highest long-term growth rate termed as the optimal backward help (y^*) . We also calculated 284 the number of offspring and the offspring survival given the optimal v^* . We investigated the 285 effects of different cost-benefit parameters on the evolvability of backward help (y) . Life-286 history parameters are based on the figures from Mace $[1]$. It is possible to generate all the 287 possible combinations of cost-benefit trade-offs by setting the appropriate cost, benefit 288 parameters to zero (c, d, h) . Also, convex or concave cost-benefit functions can be achieved 289 by setting the appropriate parameters (c, d, h) to smaller or to greater than one (see Table S1) 290 for a summary of parameters). We used the following general Leslie matrix (see Fig. 5 for a 291 schematic description):

292
\n
$$
\begin{pmatrix}\n0 & \alpha(1-y)^c + a_{12}\omega_2(1-y)^d(\omega_3 + b(1-(1-y)^b)) \\
0 & \alpha(1-y)^c + a_{12}\omega_2(1-y)^d(\omega_3 + b(1-(1-y)^b))\n\end{pmatrix}
$$
\n294
\n295

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295

296

302 benefits for both the number of offspring and for the survival of them, in the second case they

303 give benefit only for the number of offspring; in the third case they only give benefit for the

304 survival of the offspring and finally, in the last case, they do not provide any benefit. This last

305 case is not interesting for us, thus it will not be investigated any further.

306 In the same way, four possible combinations exist in terms of the cost functions: (i) c ,

307 $d > 0$; (ii) $c > 0$, $d = 0$, (iii) $c = 0$, $d > 0$; and (iv) c, $d = 0$. In the first case helping

308 grandmothers imposes a cost on both the parents' reproductive output and on the parents'

- 309 survival, in the second case only on the number of offspring, in the third only on the survival
- 310 of the parent, and finally, in the last case it imposes no cost at all. Just as before, this last case

311 is not interesting for us, thus it will not be investigated any further. See Table S2 for

312 investigated parameter combinations.

313

314 Table S2. The investigated parameter combinations (see Figures S1-S7 for the corresponding 315 results).

	Grandparental help	Shape of the cost	Shape of the benefit	Figure:
		function	function	
1.	$a_{12}, a_{21} > 0$	$d=1$	$h=1$	S ₁
2.		$d=1$	$h=2$	S ₂
3.		$d = 0.5$	$h=2$	S ₃
	4. $a_{12}=0$, $a_{21}>0$	$d=1$	$h=1$	S ₄
5.		$d = 0.5$	$h=2$	S ₅
6.	$a_{12} > 0$, $a_{21} = 0$	$d=1$	$h=1$	S ₆
7.		$d = 0.5$	$h=2$	S7

316

317 3.5 Illustrative numerical examples: results

318 IT evolves the most readily when the grandparental help increases both the survival of 319 the offspring and the number of offspring (Figure S1-S3). Linear cost and benefit functions do 320 not favour the evolution of IT (Figs. S1, S4, S6, $d=1$, $h=1$); conversely, convex benefit and 321 concave cost functions promote the evolution of IT (Fig. S2, S3, S5, S7, $d=0.5$, $h=2$). It is 322 possible to find cost parameters (c, d) where IT evolves even if the efficacy parental transfer 323 and grandparental help (a_{21} and b respectively) is low (Figs. S2, S3). Conversely, it is possible

- 324 to find (high) a_{21} , b parameters where IT evolves even if it imposes a high cost on the survival
- 325 of the parents or on the number of offspring $(d \text{ and } c,$ respectively, see Figs. S1, S2).

Figure S1. Numerical example for the Fifth Rule, grandmothers increase both offspring 334 survival and the number of offspring $(a_{12}, a_{21} > 0)$. a) Dominant eigenvalues (fitness); b) 335 corresponding y^* value; c) offspring number as a function of y^* ; d) survival of offspring as a 336 function of y^* ; all four subfigures plotted as a function of the effectiveness of IT (b) and the 337 efficacy of the grandparental help on the survival of the offspring (a_{21}) . Parameters: $a_2=6$, 338 $\omega_1=0.45, \omega_2=0.62, \omega_3=0.25, d=1, h=1, c=0.2, 0.6, 1; a_{12}=10.$

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Figure S2. Numerical example for the Fifth Rule: grandmothers increase both offspring 346 survival and the number of offspring $(a_{12}, a_{21} > 0)$. a) Dominant eigenvalues (fitness); b) 347 corresponding y^* value; c) offspring number as a function of y^* ; d) survival of offspring as a 348 function of y^* ; all four subfigures plotted as a function of the effectiveness of IT (b) and the 349 efficacy of the grandparental help on the survival of the offspring (a_{21}) . Parameters: $a_2=6$, 350 $\omega_1=0.45$, $\omega_2=0.62$, $\omega_3=0.25$, $d=1$, $h=2$, $c=0.2$, 0.6, 1; $a_{12}=10$.

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Figure S3. Numerical example for the Fifth Rule: grandmothers increase both offspring 358 survival and the number of offspring $(a_{12}, a_{21} > 0)$. a) Dominant eigenvalues (fitness); b) 359 corresponding y* value; c) offspring number as a function of y^* ; d) survival of offspring as a 360 function of y^* ; all four subfigures plotted as a function of the effectiveness of IT (b) and the 361 efficacy of the grandparental help on the survival of the offspring (a_{21}) . Parameters: $a_2=6$, 362 $\omega_1=0.45$, $\omega_2=0.62$, $\omega_3=0.25$, $d=0.5$, $h=2$, $c=0.2$, 0.6, 1; $a_{12}=10$.

Figure S4. Numerical example for the Fifth Rule, grandmothers only increase offspring 371 survival $(a_{12}= 0, a_{21} > 0)$. a) Dominant eigenvalues (fitness); b) corresponding y^* value; c) 372 offspring number as a function of y^* ; d) survival of offspring as a function of y^* ; all four 373 subfigures plotted as a function of the effectiveness of $IT(b)$ and the efficacy of the 374 grandparental help on the survival of the offspring (a_{21}) . Parameters: $a_2=6$, $\omega_1=0.45$, $\omega_2=0.62$, 375 $\omega_3=0.25, d=1, h=1, c=0.2, 0.6, 1; a_{12}=0.$

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382 Figure S5. Numerical example for the Fifth Rule, grandmothers only increase offspring 383 survival ($a_{12}=0$, $a_{21}>0$). a) Dominant eigenvalues (fitness); b) corresponding y^* value; c) 384 offspring number as a function of y^* ; d) survival of offspring as a function of y^* ; all four 385 subfigures plotted as a function of the effectiveness of IT (b) and the efficacy of the 386 grandparental help on the survival of the offspring (a_{21}) . Parameters: $a_2=6$, $\omega_1=0.45$, $\omega_2=0.62$, 387 $\omega_3=0.25, d=0.5, h=2, c=0.2, 0.6, 1; a_{12}=0.$ 388

392 Figure S6. Numerical example for the Fifth Rule, grandmothers only increase offspring 393 number ($a_{12} > 0$, $a_{21} = 0$). a) Dominant eigenvalues (fitness); b) corresponding y^* value; c) 394 offspring number as a function of y^* ; d) survival of offspring as a function of y^* ; all four 395 subfigures plotted as a function of the effectiveness of IT (b) and the efficacy of the 396 grandparental help on the fecundity of the parent (a_{12}). Parameters: $a_2=6$, $\omega_1=0.45$, $\omega_2=0.62$, 397 $\omega_3=0.25, d=1, h=1, c=0.2, 0.6, 1; a_{21}=0.$ 398

Figure S7. Numerical example for the Fifth Rule, grandmothers only increase offspring 405 number ($a_{12} > 0$, $a_{21} = 0$). a) Dominant eigenvalues (fitness); b) corresponding y^* value; c) 406 offspring number as a function of y^* ; d) survival of offspring as a function of y^* ; all four 407 subfigures plotted as a function of the effectiveness of $IT(b)$ and the efficacy of the 408 grandparental help on the fecundity of the parent (a_{12}) . Parameters: $a_2=6$, $\omega_1=0.45$, $\omega_2=0.62$, 409 $\omega_3=0.25, d=0.5, h=2, c=0.2, 0.6, 1; a_{21}=0.$

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