

Supplementary Information

to *Evolutionarily stable payoff matrix in hawk–dove games*
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SI.1

Lemma 1. (Condition for symmetric ESP) *Let $(p_1^*, 1 - p_1^*) = \mathbf{p}^*$ be an ESS with respect to the payoff matrix \mathbf{A} defined in Eq. (2) of the main text. If*

$$\partial_1 a(m^*, m^*) = \partial_2 a(m^*, m^*) = 0$$

and

$$\partial_{11} a(m^*, m^*) < 0, \quad \partial_{22} a(m^*, m^*) > 0,$$

then (\mathbf{p}^*, m^*) is an ESP with respect to the hawk–dove game defined by the payoff function $a(m_1, m_2)$.

Proof. Using notation introduced in the main text, (\mathbf{p}^*, m^*) is an ESP if there exist an $\varepsilon_0 > 0$ and a $\delta > 0$ such that Eq. (3), that is

$$(1 - \varepsilon)\mathbf{p}^* \mathbf{A} \mathbf{p}^* + \varepsilon \mathbf{p}^* \mathbf{B} \mathbf{q} > (1 - \varepsilon)\mathbf{q} \mathbf{C} \mathbf{p}^* + \varepsilon \mathbf{q} \mathbf{D} \mathbf{q},$$

holds whenever $\mathbf{q} \neq \mathbf{p}^*$, $|m - m^*| < \delta$ and $0 < \varepsilon < \varepsilon_0$.

Since \mathbf{p}^* is an ESS with respect to \mathbf{A} , there is an ε^* such that

$$(1 - \varepsilon)\mathbf{p}^* \mathbf{A} \mathbf{p}^* + \varepsilon \mathbf{p}^* \mathbf{A} \mathbf{q} > (1 - \varepsilon)\mathbf{q} \mathbf{A} \mathbf{p}^* + \varepsilon \mathbf{q} \mathbf{A} \mathbf{q}$$

is true for every $\mathbf{q} \neq \mathbf{p}^*$ and $0 < \varepsilon < \varepsilon^*$ [cf. 1]. Therefore, if

$$(1 - \varepsilon)\mathbf{p}^* \mathbf{A} \mathbf{p}^* + \varepsilon \mathbf{p}^* \mathbf{B} \mathbf{q} \geq (1 - \varepsilon)\mathbf{p}^* \mathbf{A} \mathbf{p}^* + \varepsilon \mathbf{p}^* \mathbf{A} \mathbf{q}, \quad (\text{SI.1})$$

and

$$(1 - \varepsilon)\mathbf{q} \mathbf{A} \mathbf{p}^* + \varepsilon \mathbf{q} \mathbf{A} \mathbf{q} \geq (1 - \varepsilon)\mathbf{q} \mathbf{C} \mathbf{p}^* + \varepsilon \mathbf{q} \mathbf{D} \mathbf{q} \quad (\text{SI.2})$$

hold for any $0 < \varepsilon < \varepsilon_0$ with some appropriate $\varepsilon_0 \in (0, \varepsilon^*]$ in a neighbourhood of m^* , we are done.

Since ε is positive, inequality (SI.1) is equivalent to

$$\mathbf{p}^* \mathbf{B} \mathbf{q} \geq \mathbf{p}^* \mathbf{A} \mathbf{q},$$

which is satisfied if $a(m^*, m)$ has a local minimum at $m = m^*$. This occurs if

$$\partial_2 a(m^*, m^*) = 0 \quad \text{and} \quad \partial_{22} a(m^*, m^*) > 0.$$

Inequality (SI.2) can be rearranged as follows:

$$(1 - \varepsilon)\mathbf{q}[\mathbf{A} - \mathbf{C}]\mathbf{p}^* + \varepsilon\mathbf{q}[\mathbf{A} - \mathbf{D}]\mathbf{q} \geq 0.$$

Performing the multiplications yields

$$(1 - \varepsilon)q_1 p_1^* [a(m^*, m^*) - a(m, m^*)] + \varepsilon q_1^2 [a(m^*, m^*) - a(m, m)] \geq 0,$$

where $(q_1, 1 - q_1) = \mathbf{q}$. This inequality is true if $q_1 = 0$ or the left-hand side has a strict minimum at $m = m^*$. The latter holds if $q_1 \neq 0$ and its first-order derivative with respect to m is 0 and its second order derivative with respect to m is positive at $m = m^*$, that is, if

$$-(1 - \varepsilon)p_1^* \partial_1 a(m^*, m^*) - \varepsilon q_1 [\partial_1 a(m^*, m^*) + \partial_2 a(m^*, m^*)] = 0$$

and

$$-(1 - \varepsilon)p_1^* \partial_{11} a(m^*, m^*) - \varepsilon q_1 [\partial_{11} a(m^*, m^*) + 2\partial_{12} a(m^*, m^*) + \partial_{22} a(m^*, m^*)] > 0.$$

The first condition is fulfilled if $\partial_1 a(m^*, m^*) = \partial_2 a(m^*, m^*) = 0$, while the inequality is equivalent to

$$\partial_{11} a(m^*, m^*) < \frac{-\varepsilon q_1}{(1 - \varepsilon)p_1^*} [\partial_{11} a(m^*, m^*) + 2\partial_{12} a(m^*, m^*) + \partial_{22} a(m^*, m^*)].$$

Since the right-hand side tends to 0 as $\varepsilon \rightarrow 0$, the inequality holds if $\partial_{11} a(m^*, m^*) < 0$ and $0 < \varepsilon < \varepsilon_0$ for a small enough ε_0 . □

SI.2

Lemma 2. (Invasibility) *If $a(m, m^*)$ has a strict local maximum at $m = m^*$, in particular, if*

$$\partial_1 a(m^*, m^*) = 0 \quad \text{and} \quad \partial_{11} a(m^*, m^*) < 0, \quad \text{then} \quad (\text{SI.3})$$

$\mathbf{p}^* \mathbf{A} \mathbf{p}^* > \mathbf{q} \mathbf{C} \mathbf{p}^*$ for any \mathbf{q} distinct from $\mathbf{e}_2 = (0, 1)$ and $m \neq m^*$ close enough to m^* .

Proof. Assume that $m \neq m^*$. We first show that

$$\mathbf{p}^* \mathbf{A} \mathbf{p}^* > \mathbf{e}_1 \mathbf{C} \mathbf{p}^*. \quad (\text{SI.4})$$

Performing the multiplications, the inequality takes the following shape

$$\frac{1}{2} (2a(m^*, m^*)(p_1^*)^2 + v - (p_1^*)^2 v) > a(m, m^*) p_1^* + (1 - p_1^*) v.$$

Since $p_1^* = v / (v - 2a(m^*, m^*))$ (see Eq. (4) in the main text), the left-hand side is

$$\frac{v}{2} \frac{-2a(m^*, m^*)}{v - 2a(m^*, m^*)},$$

and the right-hand side is

$$v \frac{a(m, m^*) - 2a(m^*, m^*)}{v - 2a(m^*, m^*)}.$$

We easily infer that inequality (SI.4) holds if

$$a(m^*, m^*) - a(m, m^*) > 0,$$

which follows from condition (SI.3).

On the other hand, since \mathbf{p}^* is an ESS with respect to \mathbf{A} , it follows that $\mathbf{p}^* \mathbf{A} \mathbf{p}^* = \mathbf{e}_2 \mathbf{A} \mathbf{p}^*$, and it is easy to check that $\mathbf{e}_2 \mathbf{A} \mathbf{p}^* = \mathbf{e}_2 \mathbf{C} \mathbf{p}^*$. Therefore, $\mathbf{p}^* \mathbf{A} \mathbf{p}^* = \mathbf{e}_2 \mathbf{C} \mathbf{p}^*$. Hence, for $\mathbf{q} = q_1 \mathbf{e}_1 + q_2 \mathbf{e}_2$ with $q_1 > 0$, we immediately get

$$\mathbf{p}^* \mathbf{A} \mathbf{p}^* = \mathbf{q} \mathbf{A} \mathbf{p}^* = q_1 \mathbf{e}_1 \mathbf{A} \mathbf{p}^* + q_2 \mathbf{e}_2 \mathbf{A} \mathbf{p}^* > q_1 \mathbf{e}_1 \mathbf{C} \mathbf{p}^* + q_2 \mathbf{e}_2 \mathbf{C} \mathbf{p}^* = \mathbf{q} \mathbf{C} \mathbf{p}^*.$$

□

SI.3

An observation on the conservation of $p_1^*(m, m)c(m, m)$

If the payoff matrix is

$$\begin{pmatrix} a(m_1, m_2) & v \\ 0 & v/2 \end{pmatrix},$$

then

$$a(m_1, m_2) = \pi(m_1, m_2)v - (1 - \pi(m_1, m_2)) \frac{\pi(m_1, m_2)v - a(m_1, m_2)}{1 - \pi(m_1, m_2)}.$$

Hence, following the structure of $a(m_1, m_2)$ given in Eq. (1) of the main text, we can express the cost function $c(m_1, m_2)$ as

$$c(m_1, m_2) = \frac{\pi(m_1, m_2)v - a(m_1, m_2)}{1 - \pi(m_1, m_2)}, \quad (\text{SI.5})$$

from which

$$c(m, m) = \frac{\pi(m, m)v - a(m, m)}{1 - \pi(m, m)} = \frac{\frac{1}{2}v - a(m, m)}{\frac{1}{2}} = v - 2a(m, m).$$

Consequently, if ESS $\mathbf{p}^* = \mathbf{p}^*(m, m) = (p_1^*, 1 - p_1^*)$ is the ESS for a trait value m , then

$$p_1^* = p_1^*(m, m) = \frac{v}{v - 2a(m, m)} = \frac{v}{c(m, m)},$$

and so the product $p_1^*(m, m)c(m, m)$, that is, the per game average of the maximal possible cost of playing hawk in a monomorphic ESS population is always v independently of m .

SI.4

Examples of local extrema in $c(m, m)$ at symmetric ESP

In our example in the main text, both $c(m, m)$ and $a(m, m)$ are constant. Here, we show that this is not the general situation and a strict maximum or minimum can occur at $m = m^*$. To find such a function, we rotate the graph of function $a(m_1, m_2)$ defined by Eq. (7) of the main text around (m^*, m^*) . If

$$u_\varphi(m_1, m_2) = m^* + \cos(\varphi)(m_1 - m^*) + \sin(\varphi)(m_2 - m^*)$$

and

$$w_\varphi(m_1, m_2) = m^* - \sin(\varphi)(m_1 - m^*) + \cos(\varphi)(m_2 - m^*),$$

then the graph of

$$a_\varphi(m_1, m_2) := a(u_\varphi(m_1, m_2), w_\varphi(m_1, m_2))$$

is the graph of $a(m_1, m_2)$ after a rotation of angle φ around (m^*, m^*) . Let us denote the corresponding cost function calculated according to (SI.5) by $c_\varphi(m_1, m_2)$.

For example, if $m^* = 5/8$ and

$$\begin{aligned} \varphi_1 = 30^\circ, \text{ then } u_{\varphi_1}(m_1, m_2) &= \frac{\sqrt{3}}{2}m_1 + \frac{m_2}{2} + \frac{5(1 - \sqrt{3})}{16} \\ \text{and } w_{\varphi_1}(m_1, m_2) &= -\frac{m_1}{2} + \frac{\sqrt{3}}{2}m_2 + \frac{5(3 - \sqrt{3})}{16}, \end{aligned}$$

and the cost $c_{\varphi_1}(m, m)$ is minimal, and so $a_{\varphi_1}(m, m)$ is maximal at $m = m^*$.

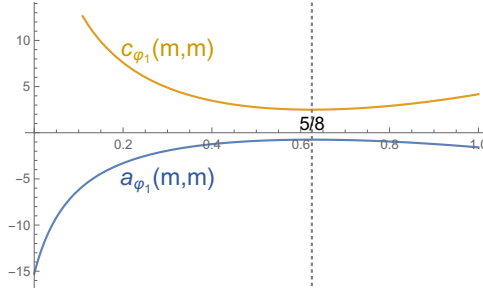


Fig. SI.1 The graphs of $c_{\varphi_1}(m, m)$ and $a_{\varphi_1}(m, m)$

While if

$$\begin{aligned} \varphi_2 = -30^\circ, \text{ then } u_{\varphi_2}(m_1, m_2) &= \frac{\sqrt{3}}{2}m_1 - \frac{m_2}{2} + \frac{5(3 - \sqrt{3})}{16} \\ \text{and } w_{\varphi_2}(m_1, m_2) &= \frac{m_1}{2} + \frac{\sqrt{3}}{2}m_2 + \frac{5(1 - \sqrt{3})}{16}, \end{aligned}$$

and the cost $c_{\varphi_2}(m, m)$ is maximal, and so $a_{\varphi_2}(m, m)$ is minimal at $m = m^*$.

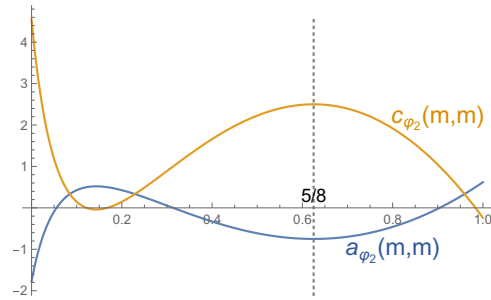


Fig. SI.2 The graphs of $c_{\varphi_2}(m, m)$ and $a_{\varphi_2}(m, m)$

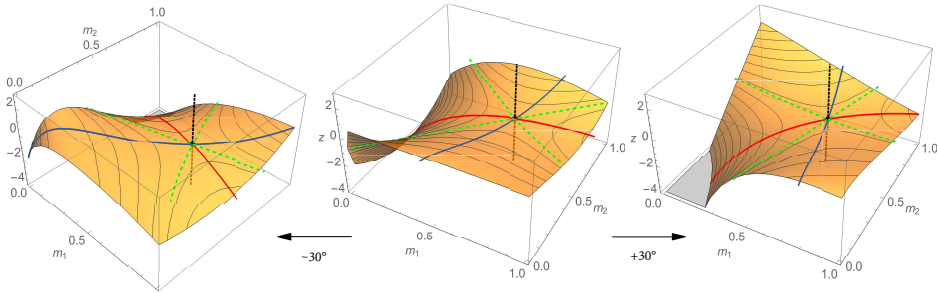


Fig. SI.3 The figure in the middle is the graph of the $a(m_1, m_2)$ defined by Eq. (7) of the main text. The figures on the left and the right show the graphs of the functions derived from $a(m_1, m_2)$ by rotating its graph by -30° and 30° around the dashed black line, respectively. The dashed green curves make up the level line corresponding to the value $a(m^*, m^*)$, and their intersection is the point on the graph corresponding to (m^*, m^*) . When the graph of $a(m_1, m_2)$ is rotated by -30° ($+30^\circ$), then the blue (red) curve ends up above the points with (m, m) coordinates.