Phoshorelays provide tunable signal processing capabilities for the cell

Supplementary Text

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This document provides the proofs and explanations of the analytical results described in the main text. The results are summarized in the first two sections and the claims are proven in the two appendices.

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1 The kinase HK is monofunctional

1.1 Reactions, equations and steady states

The model consists of four proteins: a histidine kinase HK, a receiver protein REC, a His-containing phosphotransfer protein Hpt, and a response regulator RR. Each of these proteins can be either phosphorylated (in which case we write $X_p$, where $X$ is one of the four proteins HK, REC, Hpt, and RR) or unphosphorylated (in which case we write $X$).

**Reactions.** The minimal set of reactions that the system has, consists of the autophosphorylation reaction of HK and the forward phosphorelays:

$$
\begin{align*}
&\text{HK} \xrightarrow{k_s} \text{HK}_p \\
&\text{HK}_p + \text{REC} \xrightarrow{k_2} \text{HK} + \text{REC}_p \\
&\text{REC}_p + \text{Hpt} \xrightarrow{k_3} \text{REC} + \text{Hpt}_p \\
&\text{Hpt}_p + \text{RR} \xrightarrow{k_4} \text{Hpt} + \text{RR}_p.
\end{align*}
$$

We study extensions of this minimal reaction mechanism obtained by adding reverse phosphotransfer reactions involving HK, REC, Hpt and RR, as well as hydrolysis reactions of REC$_p$ and RR$_p$:

$$
\begin{align*}
&\text{HK} + \text{REC}_p \xrightarrow{k_{2r}} \text{HK}_p + \text{REC} \\
&\text{REC} + \text{Hpt}_p \xrightarrow{k_{3r}} \text{REC}_p + \text{Hpt} \\
&\text{Hpt} + \text{RR}_p \xrightarrow{k_{4r}} \text{Hpt}_p + \text{RR}.
\end{align*}
$$

By setting some of the rate constants $k_{sr}$ or $k_{hr}$ to zero, we obtain 32 different topologies involving different combinations of reverse phosphorelay and hydrolysis reactions (Table 1 and Table S1). The minimal set of reactions is always part of the system, meaning that $k_2, k_3, k_4, k_s \neq 0$ for all topologies.

**Ordinary differential equations.** We model the protein concentrations in the system using ordinary differential equations (ODEs). To simplify the notation, we define:

$$
\begin{align*}
&x_1 = [\text{HK}] \\
&x_2 = [\text{HK}_p] \\
&x_3 = [\text{REC}] \\
&x_4 = [\text{REC}_p] \\
&x_5 = [\text{Hpt}] \\
&x_6 = [\text{Hpt}_p] \\
&x_7 = [\text{RR}] \\
&x_8 = [\text{RR}_p].
\end{align*}
$$

The dynamics of the concentrations in time is modeled with a system of ODEs:

$$
\begin{align*}
\dot{x}_1 &= -k_s x_1 - k_2 r x_1 x_4 + k_2 x_2 x_3 \\
\dot{x}_2 &= k_s x_1 + k_2 r x_1 x_4 - k_2 x_2 x_3 \\
\dot{x}_3 &= -k_2 x_2 x_3 + k_{hr_1} x_4 + k_2 r x_1 x_4 + k_3 x_4 x_5 - k_3 r x_3 x_6 \\
\dot{x}_4 &= k_2 x_2 x_3 - k_{hr_1} x_4 - k_2 r x_1 x_4 - k_3 x_4 x_5 + k_3 r x_3 x_6 \\
\dot{x}_5 &= -k_3 x_4 x_5 + k_{3r} x_3 x_6 + k_4 x_6 x_7 - k_{4r} x_5 x_8 \\
\dot{x}_6 &= k_3 x_4 x_5 - k_{3r} x_3 x_6 - k_4 x_6 x_7 + k_{4r} x_5 x_8 \\
\dot{x}_7 &= -k_4 x_6 x_7 + k_{4r} x_5 x_8 \\
\dot{x}_8 &= k_4 x_6 x_7 - k_{4r} x_5 x_8 - k_{4r} x_5 x_8.
\end{align*}
$$

We note that

$$
\dot{x}_1 + \dot{x}_2 = 0, \quad \dot{x}_3 + \dot{x}_4 = 0, \quad \dot{x}_5 + \dot{x}_6 = 0, \quad \dot{x}_7 + \dot{x}_8 = 0.
$$
As a consequence $x_i + x_{i+1}$ is constant for $i = 1, 3, 5, 7$ and the system has four conserved amounts:

\[
\begin{align*}
HK_{\text{tot}} &= x_1 + x_2, & \text{REC}_{\text{tot}} &= x_3 + x_4, & Hpt_{\text{tot}} &= x_5 + x_6, & RR_{\text{tot}} &= x_7 + x_8,
\end{align*}
\]

where $HK_{\text{tot}}, \text{REC}_{\text{tot}}, Hpt_{\text{tot}}$, and $RR_{\text{tot}}$ are positive constants given by the initial concentrations of the system. To ease the writing, we change the notation to

\[
\begin{align*}
\mathcal{H} &:= HK_{\text{tot}}, & \mathcal{C} &:= \text{REC}_{\text{tot}}, & \mathcal{T} &:= Hpt_{\text{tot}}, & \mathcal{R} &:= RR_{\text{tot}}.
\end{align*}
\]

**Steady-state equations.** The steady states of the system are found by setting the derivatives, $\dot{x}_i$, of the concentrations to zero, that is, $\dot{x}_i = 0$. By equating the right-hand side of the ODEs to zero we obtain a system of polynomial equations in the concentrations $x_i$.

Due to the existence of conserved amounts, some equations are redundant. For instance, the first and second steady-state equations are

\[
0 = -k_3 x_1 - k_2 r x_1 x_4 + k_2 x_2 x_3, \quad 0 = k_4 x_1 + k_2 r x_1 x_4 - k_2 x_2 x_3.
\]

One equation is minus the other, and hence, if one of them is fulfilled then so is the other. This happens because $x_1 + x_2$ is conserved. In total, four of the steady-state equations are redundant and must be replaced by the conservation equations. The steady states of the system are thus given as the solutions to the following system of equations:

\[
\begin{align*}
\mathcal{H} &= x_1 + x_2 & (S2) & 0 &= k_4 x_1 + k_2 r x_1 x_4 - k_2 x_2 x_3 & (S6) \\
\mathcal{C} &= x_3 + x_4 & (S3) & 0 &= k_2 x_2 x_3 - k_3 x_4 x_5 + k_3 r x_3 x_6 & (S7) \\
\mathcal{T} &= x_5 + x_6 & (S4) & 0 &= k_3 x_4 x_5 - k_3 r x_3 x_6 - k_4 x_6 x_7 + k_4 r x_5 x_8 & (S8) \\
\mathcal{R} &= x_7 + x_8 & (S5) & 0 &= k_4 x_6 x_7 - k_4 x_5 x_8 & (S9)
\end{align*}
\]

Only solutions with non-negative solutions are meaningful, that is, all concentrations must be positive or zero. Therefore, a steady state will always refer to a non-negative steady state.

If the rate constants and the total amounts are known, then we solve this system of equations to find the steady states. This can be done using mathematical software such as Matlab, Maple, or Mathematica.

**Rearrangement of the steady-state equations.** The set of equations (S2)-(S9) can be replaced by another system of equations that is easier to interpret. We change equations (S6)-(S9) by linear combinations of them. This process does not change the set of solutions to the system. Specifically, we replace:

(S8) by (S8)+(S9), \quad (S7) by (S7)+(S8)+(S9), \quad (S6) by (S6)+(S7)+(S8)+(S9),

and leave (S9) as it is. This results in the following equivalent system of equations:

\[
\begin{align*}
\mathcal{H} &= x_1 + x_2 & (S10) & 0 &= k_4 x_1 - k_3 x_4 - k_4 x_8 & (S14) \\
\mathcal{C} &= x_3 + x_4 & (S11) & 0 &= k_2 x_2 x_3 - k_3 x_4 x_5 - k_3 x_3 x_6 - k_4 x_2 x_8 & (S15) \\
\mathcal{T} &= x_5 + x_6 & (S12) & 0 &= k_3 x_4 x_5 - k_4 x_3 x_6 - k_4 x_2 x_8 & (S16) \\
\mathcal{R} &= x_7 + x_8 & (S13) & 0 &= k_4 x_6 x_7 - k_4 x_5 x_8 - k_4 x_2 x_8 & (S17)
\end{align*}
\]

where (S2)-(S5) are identical to (S10)-(S13), and (S17) is identical to (S9).
1.2 Steady-state relations and signal-response curve

We proceed to find an analytical expression of the signal-response curve. In this system, the signal is taken to be the value of the rate constant $k_s$, and the response is the steady-state value of the phosphorylated response regulator ($x_8$) corresponding to $k_s$ (with all the other rate constants and total amounts fixed).

In order to find an analytical relation between $k_s$ and $x_8$, we establish an analytical relation between each concentration $x_i$ and $x_8$ at steady state. For some topologies, however, the signal-response curve is constant. We start by discussing when this is the case and then proceed to study the remaining topologies.

**Topologies with constant signal-response curve.** In the following cases, the signal-response curve is constant and equals $RR_{tot}$ (i.e. $x_8 = R$ at steady state, for any non-zero value of $k_s$). Hence, the signal-response curve cannot have hyperbolic or sigmoidal shape, or show ultrasensitivity in these cases:

- $k_{h1} = k_{h2} = 0$.
- $k_{h2} = k_{3r} = 0$.
- $k_{h2} = k_{4r} = 0$.

The topologies that belong to at least one of the cases above are: **topologies 9-13, 15** and **topologies 17-24**. All the other topologies have non-constant signal-response curves and all concentrations at steady state are non-zero. In particular, if $k_{h2} \neq 0$, that is, if there is hydrolysis at $RR_p$, then the signal-response curve is not constant.

These claims are proven in Appendix A.1.

**Steady-state analytical relations.** We have expressed each concentration $x_i$ at steady state as a function of $x_8$. This implies that if the value of $x_8$ at steady state is known, then so are the values of $x_1, \ldots, x_7$.

We assume that if $k_{h2} = 0$ then either $k_{h1} \neq 0$, $k_{3r} \neq 0$ or $k_{4r} \neq 0$. That is, we consider only the cases for which the signal-response curve is not constant. In particular, there are no zero concentrations at steady state and hence all steady states must be positive at each concentration.

The following lists an iterative way to find the steady-state values of all variables, once the value of $x_8$ is known. These expressions are derived in Appendix A.2 using equations (S10)-(S13) and (S15)-(S17).

<table>
<thead>
<tr>
<th>Expression</th>
<th>Behavior as function of $x_8$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$x_7 = \bar{R} - x_8$</td>
<td>$x_7$ decreases in $x_8$</td>
</tr>
<tr>
<td>$x_5 = \frac{k_{1}T}{x_7-k_{h2}x_8}{k_{4}x_7+k_{4}x_8}$</td>
<td>$x_5$ decreases in $x_8$</td>
</tr>
<tr>
<td>$x_6 = \bar{T} - x_5$</td>
<td>$x_6$ increases in $x_8$</td>
</tr>
<tr>
<td>$x_3 = \frac{k_{3}C}{x_5-k_{h2}x_8}{k_{3}x_5+k_{3r}x_6}$</td>
<td>$x_3$ decreases in $x_8$</td>
</tr>
<tr>
<td>$x_4 = \bar{C} - x_3$</td>
<td>$x_4$ increases in $x_8$</td>
</tr>
<tr>
<td>$x_1 = \frac{k_{1}P}{x_3-k_{h2}x_4-k_{h2}x_8}{k_{2}x_3+k_{2r}x_4}$</td>
<td>$x_1$ decreases in $x_8$</td>
</tr>
<tr>
<td>$x_2 = \bar{P} - x_1$</td>
<td>$x_2$ increases in $x_8$</td>
</tr>
</tbody>
</table>

(S18)
By plugging in iteratively the explicit expressions in \( x_8 \) we obtain that each variable is expressed at steady state as the quotient of two polynomials in \( x_8 \). The steady-state values are all positive if and only if \( x_8 \) belongs to the interval \( I = (0, \alpha) \), where \( \alpha \) is the first positive root of the degree-2 polynomial:

\[
q_2(x_8) := a_2 x_8^2 + a_1 x_8 + a_0
\]

with

\[
\begin{align*}
a_2 &= k_{h2}(k_4 - k_{4r}) (k_{h1} + k_4 T) + k_{h2}^2 (k_3 - k_{3r}) + k_{h2} (k_3 k_4 - k_{3r} k_{4r}) T \\
a_1 &= -k_{h1} (k_{3r} C (k_{h2} + k_{4r} T) + k_{h2} k_4 R) - k_{h2} (k_4 R + k_3 C) + k_3 k_4 T R - k_2 k_3 k_4 H C T \\
a_0 &= k_2 k_3 k_4 H C T R
\end{align*}
\]

(see Appendix A.2 for a proof).

**Signal-response expression.** Using the remaining steady-state equation, \((S14)\), we express \( k_s \) in terms of \( x_8 \) (see Appendix A.3):

\[
k_s = \frac{k_{h1} x_4 + k_{h2} x_8}{x_1}.
\]

Since \( x_4 \) increases in \( x_8 \) and \( x_1 \) decreases in \( x_8 \), it follows that \( k_s \) increases in \( x_8 \). If we express \( x_4, x_1 \) in terms of \( x_8 \) using \((S18)\), we obtain that the exact analytical expression relating \( k_s \) and \( x_8 \) is:

\[
k_s = f(x_8) = \frac{x_8 p_1(x_8) p_2(x_8)}{q_1(x_8) q_2(x_8)}
\]

with \( q_2(x_8) \) given as in \((S19)\) and

\[
\begin{align*}
q_1(x) &= (k_{h2} (k_4 - k_{3r}) + (k_3 k_4 - k_{3r} k_{4r}) T) x - k_3 k_4 R T, \\
p_1(x) &= k_{h2} (k_{h1} (k_4 - k_{4r}) + k_{h2} (k_3 - k_{3r}) + (k_3 k_4 - k_{3r} k_{4r}) T) x \\
&\quad - k_4 k_{h2} (k_{h1} + k_3 T) R - k_{3r} k_{h1} (k_{h2} + k_4 T) C, \\
p_2(x) &= k_{h2} (k_2 - k_{2r}) (k_4 - k_{4r}) x^2 + k_2 k_3 k_4 R C T + \\
&\quad (k_{h2} (k_4 (k_2 - k_{2r}) R + (k_2 k_3 r - k_2 k_3 C) + (k_2 k_3 r - k_2 k_3 C) R C T) x.
\end{align*}
\]

This function is well defined for \( x_8 \) is in \( I = (0, \alpha) \), that is, it is positive and continuous. When \( x_8 \) approaches \( \alpha \) (the upper bound of the interval \( I \)), then \( k_s \) tends to infinity (the denominator of \( f \) tends to zero). Therefore, the image of \( f \) is the interval \((0, +\infty)\). Further the function \( f \) can be differentiably extended at zero such that \( f(0) = 0 \).

Given a rate constant \( k_s \), there is a unique value of \( x_8 \) for which \( f(x_8) = k_s \). This value is the steady-state value of \( x_8 \) corresponding to \( k_s \), and the other steady states are found using \((S18)\).

**Properties of the signal-response curve.** We let \( \varphi \) denote the inverse of \( f \), that is,

\[
\varphi(k_s) = x_8 \quad \text{if} \quad k_s = f(x_8).
\]

Using the Inverse Function Theorem, the signal-response function \( \varphi \) is continuous and differentiable in \([0, +\infty)\). We do not have an analytical expression for \( \varphi \), only of its inverse. However, most of the information required from \( \varphi \) can be retrieved from \( f \):

(i) The function \( \varphi \) is increasing.
Since the second derivative of \( \varphi \) is negative, hyperbolic, that is the derivative of \( \varphi \) at zero is \( \varphi'(k) = 1/f'(x_8) \) for \( x_8 = \varphi(k) \).

(iii) The derivative of \( \varphi \) at a point \( k_s = k \) equals \( \varphi'(k) = -f''(x_8)/f'(x_8)^3 \) for \( x_8 = \varphi(k) \).

For example, the derivative of the signal-response curve \( \varphi \) at zero is:

\[
\frac{k_3k_4\Pi R T}{k_4k_2(h_{12} + k_3)R + k_3y_{12}h_{12} + k_4TRC}.
\]

Since \( f \) is an increasing function in \( I \), we have that \( \varphi'(k) > 0 \) for all \( k \geq 0 \) and the sign of the second derivative of \( \varphi \) at \( k \) is minus the sign of the second derivative of \( f \) at \( \varphi(k) \).

**Practical considerations.**

(v) The signal-response curve is plotted by generating points \((f(x_8), x_8)\).

(vi) The maximal response is easily computed as the first positive root of \( q_2(x) \), which is a degree-2 polynomial.

(vii) Given \( k_s \), the steady-state value of \( x_8 \) is the first positive zero of the polynomial

\[
k_s q_1(x_8)q_2(x_8) - x_8 p_1(x_8)p_2(x_8).
\]

The other steady-state values are obtained from \( x_8 \) and (S18).

### 1.3 Hyperbolic and sigmoidal signal-response curves

**Second derivative at zero.** A function \( g(x) \) that increases at a slower and slower rate is called hyperbolic, that is the derivative \( g'(x) \) of \( g \) is decreasing or, alternatively, the second derivative is negative, \( g''(x) < 0 \). A function \( g(x) \) that initially increases at a faster and faster rate and then slows down is called sigmoidal, that is, \( g'(x) \) is initially increasing then decreasing or, alternatively, \( g''(x) \) is first positive and then becomes negative.

It is difficult in general to establish if a curve is sigmoidal or hyperbolic (or none of these) and we use a simple test to indicate if \( \varphi \) is sigmoidal or hyperbolic. If the second derivative of \( \varphi \) at zero is positive, then the first derivative grows indicating that the curve will likely be sigmoidal. If, on the contrary, the second derivative of \( \varphi \) at zero is negative, then the curve is likely to be hyperbolic. This test is a good indicator of the shape of \( \varphi \), but note that the test only considers the behavior near zero.

We have observed a perfect overlap between the classification obtained using the sign of the second derivative only at zero and the classification obtained by checking whether the second derivative of the entire signal-response curve changes sign (see Methods). This supports that the classification based on the sign of the second derivative at zero is reasonable.

We compute \( \varphi''(0) \) using item (iv) above and \( \varphi(0) = 0 \). Computations are performed in Mathematica. The sign of \( \varphi''(0) \) agrees with the sign of:

\[
S = -k_{h1}(k_{4r}T + k_{h2})(k_{3r}(k_{2k3k4} - k_{2k3k4r} + k_{2r}k_{3r}k_{4r})CT + k_4k_{2r}k_{3r}k_{h2}TR
+ k_{h2}(k_{2k3} - k_{2k3r} + k_{2r}k_{3r})(k_{4r}R + k_{3r}C)) \frac{R}{C}
- k_{2r}k_{3r}(k_{4r}T + k_{h2})(CT + k_{4r}k_{h2}(k_{3r}T + k_{h1})R) \frac{R}{C}
- (k_{3r}k_{h1}(k_{4r}T + k_{h2})C + k_{4r}k_{h2}(k_{3r}T + k_{h1})R)^2.
\]

(S22)
If $S > 0$, then the signal-response curve is classified as sigmoidal. If $S < 0$, then the signal-response curve is classified as hyperbolic. If the blue terms are all positive, then the curve is hyperbolic and hence only if some of the highlighted blue terms are negative can $S > 0$. Note that the negative terms in $S$ are multiplied by $k_{h1}$. Therefore, necessary conditions for $\varphi$ to be sigmoidal are

$$k_{h1} > 0 \quad \text{and} \quad (k_2 k_3 k_4 - k_2 k_3 k_4 + k_2 k_3 k_4 < 0 \quad \text{or} \quad k_2 k_3 - k_2 k_3 + k_2 k_3 < 0).$$

In particular, if $k_{3r} = 0$ then sigmoidality cannot occur. By inspecting in detail the two blue terms, we see that

$$k_2 k_3 - k_2 k_3 + k_2 k_3 = k_2(k_3 - k_3) + k_2 k_3 = k_2 k_3 + (k_2 - k_2) k_{3r}$$
$$k_2 k_3 k_4 - k_2 k_3 k_4 = k_2(k_3 k_4 - k_3 k_4) + k_2 k_3 k_4 = k_2 k_3 k_4 + (k_2 - k_2) k_{3r} k_{4r}.$$

We conclude that necessary conditions for $\varphi$ to be sigmoidal are:

$$k_{h1} > 0 \quad \text{and} \quad k_2 > k_2 \quad \text{and} \quad (k_3 > k_3 \quad \text{or} \quad k_3 k_4 > k_3 k_4). \quad (S23)$$

Further, we conclude the following from an analysis of the expression of $S$:

- If $H$ or $k_2$ are very small, then $S$ is negative and hence $\varphi$ is hyperbolic.
- If $k_{h2} = 0$, then the sign of $S$ agrees with the sign of

$$H(k_2 k_3 k_4 - k_2 k_3 k_4 + k_2 k_3 k_4) + k_3 k_4 k_{h1}.$$

In this case, necessary conditions for $\varphi$ to be sigmoidal are:

$$k_{h1} \neq 0 \quad \text{and} \quad k_2 > k_2 \quad \text{and} \quad k_3 k_4 > k_3 k_4.$$

Hyperbolic curves. We have also shown (see Appendix A.4) that if:

$$k_2 > k_2 \quad \text{and} \quad (k_3 - k_3) k_{h2}(k_4 H + k_3 C) + k_3 k_4(k_3 k_4 - k_3 k_4) C T > 0, \quad (S24)$$

then the second derivative of $\varphi$ strictly decreases over $I$ and hence the curve is hyperbolic (that is, not only the second derivative at zero indicates so). These two inequalities are fulfilled if

$$k_2 > k_2, \quad k_3 > k_3, \quad k_4 > k_4.$$

That is, if the phosphorylation rate constants are larger than their reverse counterparts then the curve is hyperbolic. However, the curve can be hyperbolic without these inequalities being fulfilled.

Tuning by varying total amounts. By expressing the term $S$ in [S22] as a polynomial in one of the total amounts, we can observe that for some parameter values, variation of the total amounts can change the system’s response from sigmoidal ($S$ positive) to hyperbolic ($S$ negative and vice versa). This fact is summarized in the following table:

<table>
<thead>
<tr>
<th>Total amount</th>
<th>Degree</th>
<th>Independent coefficient</th>
<th>Leading coefficient</th>
<th>Sign of $S$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$H$</td>
<td>1</td>
<td>negative</td>
<td>positive for some parameters</td>
<td>$S &lt; 0$ for $H$ small, $S &gt; 0$ for $H$ large</td>
</tr>
<tr>
<td>$C$</td>
<td>2</td>
<td>negative</td>
<td>positive for some parameters</td>
<td>$S &lt; 0$ for $C$ small, $S &gt; 0$ for $C$ large</td>
</tr>
<tr>
<td>$T$</td>
<td>2</td>
<td>positive for some parameters</td>
<td>negative</td>
<td>$S &gt; 0$ for $T$ small, $S &lt; 0$ for $T$ large</td>
</tr>
<tr>
<td>$R$</td>
<td>2</td>
<td>positive for some parameters</td>
<td>negative</td>
<td>$S &gt; 0$ for $R$ small, $S &lt; 0$ for $R$ large</td>
</tr>
</tbody>
</table>
The results of the last column in the table hold for any choice of parameters that make the leading or independent coefficient (depending on the total amount) positive.

1.4 Examples

We consider a specific example with rate constants and total amounts given by

\[
\begin{align*}
  & k_{h1} = 0.5 & k_{h2} = 1 & k_2 = 0.1 & k_4 = 1 & k_{2r} = 1 & k_{3r} = 1 \\
  & k_{4r} = 0.5 & k_3 = 0.3 & \bar{H} = 1 & \bar{R} = 10 & \bar{T} = 10 & \bar{C} = 100.
\end{align*}
\]

The signal-response function \( f \) given in (S20) is:

\[
k_s = \frac{x_8(505 - 16.25x_8)(2000 + 389x_8 + 0.45x_8^2)}{(-200 + 16x_8)(2000 - 726x_8 + 16.3x_8^2)},
\]

such that

\[
q_2(x_8) = 2000 - 726x_8 + 16.3x_8^2.
\]

The first positive real root of \( q_2 \) is \( \alpha = 2.95024 \). The graph of \( f \) in \([0, \alpha]\) is shown in Figure S4(A). If \( k_s \) is given, then the anti-image of \( k_s \) in this graph is the steady-state value of \( x_8 \) and is found as shown in Figure S4(B). Since the curve is strictly increasing, there is one and only one anti-image. We see from the graph that as \( k_s \) becomes large, \( x_8 \) approaches \( \alpha \), but the value of \( x_8 \) can never exceed \( \alpha \). That is, \( \alpha \) is the upper bound of the response \( x_8 \). The graph of the signal-response curve \( \varphi \) is obtained by reversing the axes (Figure S4(C)). Therefore, when \( k_s \) tends to infinity, \( x_8 \) approaches the maximal response \( \alpha \).

1.5 Model with intermediates

Reactions. We extend the model given in Subsection 1.1 to incorporate the formation of intermediate complexes at the phosphotransfer reactions. That is, the model extended with intermediates consist of the reactions

\[
\begin{align*}
  & \text{HK} \xrightarrow{k_s} \text{HKp} & \text{HKp + REC} \xrightarrow{k_{2a}} \text{Y}_1 \xrightarrow{k_{2b}} \text{HK + RECp} \\
  & \text{RECp} \xrightarrow{k_{h1}} \text{REC} & \text{RECp + Hpt} \xrightarrow{k_{3a}} \text{Y}_2 \xrightarrow{k_{3b}} \text{REC + Hptp} \\
  & \text{RRp} \xrightarrow{k_{h2}} \text{RR} & \text{Hptp + RR} \xrightarrow{k_{4a}} \text{Y}_3 \xrightarrow{k_{4b}} \text{Hpt + RRp}.
\end{align*}
\]

By setting some of the rate constants \( k_{sr} \) or \( k_{hr} \) to zero, we obtain different topologies involving different combinations of reverse phosphorelay and hydrolysis reactions as before.

Ordinary differential equations. We model the protein concentrations in the system using ordinary differential equations (ODEs). To simplify the notation, we define:

\[
\begin{align*}
  x_1 &= [\text{HK}] & x_2 &= [\text{HKp}] & x_3 &= [\text{REC}] & x_4 &= [\text{RECp}] \\
  x_5 &= [\text{Hpt}] & x_6 &= [\text{Hptp}] & x_7 &= [\text{RR}] & x_8 &= [\text{RRp}] \\
  x_9 &= [\text{Y}_1] & x_{10} &= [\text{Y}_2] & x_{11} &= [\text{Y}_3].
\end{align*}
\]
The dynamics of the concentrations in time is modeled with a system of ODEs:

\[
\begin{align*}
\dot{x}_1 &= -k_s x_1 - k_{2br} x_1 x_4 + k_{26} x_9 \\
\dot{x}_2 &= k_s x_1 + k_{2ar} x_9 - k_{2a} x_2 x_3 \\
\dot{x}_3 &= -k_{2a} x_2 x_3 + k_{h1} x_4 + k_{2ar} x_9 + k_{3br} x_{10} - k_{3br} x_3 x_6 \\
\dot{x}_4 &= k_{2b} x_9 - k_{h1} x_4 - k_{2br} x_1 x_4 - k_{3a} x_4 x_5 + k_{3ar} x_{10} \\
\dot{x}_5 &= -k_{3a} x_4 x_5 + k_{3ar} x_{10} + k_{4b} x_{11} - k_{4br} x_5 x_8 \\
\dot{x}_6 &= k_{3b} x_{10} - k_{3ar} x_3 x_6 - k_{4a} x_6 x_7 + k_{4ar} x_{11} \\
\dot{x}_7 &= -k_{4a} x_6 x_7 + k_{h2} x_8 + k_{4ar} x_{11} \\
\dot{x}_8 &= k_{4b} x_{11} - k_{h2} x_8 - k_{4br} x_5 x_8 \\
\dot{x}_9 &= k_{2a} x_2 x_3 - k_{2ar} x_9 + k_{2br} x_1 x_4 - k_{26} x_9 \\
\dot{x}_{10} &= k_{3a} x_4 x_5 - k_{3ar} x_{10} - k_{3b} x_{10} + k_{2ar} x_3 x_6 \\
\dot{x}_{11} &= k_{4a} x_6 x_7 - k_{4ar} x_{11} + k_{4br} x_5 x_8 - k_{4b} x_{11}.
\end{align*}
\]

The system has four conserved amounts:

\[HK_{\text{tot}} = x_1 + x_2 + x_9, \quad REC_{\text{tot}} = x_3 + x_4 + x_9 + x_{10}, \quad Hpt_{\text{tot}} = x_5 + x_6 + x_{10} + x_{11}, \quad RR_{\text{tot}} = x_7 + x_8 + x_{11},\]

where \(HK_{\text{tot}}, REC_{\text{tot}}, Hpt_{\text{tot}},\) and \(RR_{\text{tot}}\) are positive constants given by the initial concentrations of the system. As before, we write

\[
\begin{align*}
\Pi &:= HK_{\text{tot}}, \quad C := REC_{\text{tot}}, \quad T := Hpt_{\text{tot}}, \quad R := RR_{\text{tot}}.
\end{align*}
\]

**Steady-state equations.** We proceed as in the monofunctional case and conclude that the steady states of the system are given as the solutions to the following system of equations:

\[
\begin{align*}
0 &= -k_s x_1 - k_{2br} x_1 x_4 + k_{26} x_9 \\
0 &= -k_{2a} x_2 x_3 + k_{h1} x_4 + k_{2ar} x_9 + k_{3br} x_{10} - k_{3br} x_3 x_6 \\
0 &= k_{2b} x_9 - k_{h1} x_4 - k_{2br} x_1 x_4 - k_{3a} x_4 x_5 + k_{3ar} x_{10} \\
0 &= -k_{3a} x_4 x_5 + k_{3ar} x_{10} + k_{4b} x_{11} - k_{4br} x_5 x_8 \\
0 &= k_{3b} x_{10} - k_{3ar} x_3 x_6 - k_{4a} x_6 x_7 + k_{4ar} x_{11} \\
0 &= -k_{4a} x_6 x_7 + k_{h2} x_8 + k_{4ar} x_{11} \\
0 &= k_{4b} x_{11} - k_{h2} x_8 - k_{4br} x_5 x_8 \\
0 &= k_{2a} x_2 x_3 - k_{2ar} x_9 + k_{2br} x_1 x_4 - k_{26} x_9 \\
0 &= k_{3a} x_4 x_5 - k_{3ar} x_{10} - k_{3b} x_{10} + k_{2ar} x_3 x_6
\end{align*}
\]

**Hyperbolic vs. sigmoidal.** It is not so straightforward in this case to obtain an analytical description of the signal-response curve. Therefore, we adopt a direct route to the computation of the sign of the second derivative at zero.

When \(k_s = 0\), then the steady state of the system equals

\[(x_1, \ldots, x_{11}) = (\Pi, 0, C, 0, T, 0, R, 0, 0, 0, 0).\]

We want to find the derivative of the response \(x_s\) with respect to \(k_s\) at \(k_s = 0\) at steady state. To this end, we do the following steps:
1. We take the derivative with respect to $k_s$ of both sides of the steady-state equations. We obtain new equations, where $p_i = \frac{\partial x_i}{\partial k_s}$:

$$
\begin{align*}
0 &= -x_1 - k_sp_1 - k_{2br}x_{14} - k_{2br}x_{1p4} + k_{2b}p_9, \\
0 &= -k_{2a}x_2p_3 - k_{2a}x_2p_3 + k_{2ar}p_9 - k_{3br}p_3x_6 - k_{3br}x_3p_6 + k_{3b}p_{10} + k_{h1}p_4, \\
0 &= -k_{2br}p_1x_4 - k_{2br}x_{14} + k_{2b}p_9 - k_{3a}x_4p_5 - k_{3ar}p_5 - k_{3br}p_5x_8 - k_{4br}x_5p_8 + k_{4b}p_{11}, \\
0 &= -k_{3a}x_4p_5 - k_{3a}x_4p_5 + k_{3ar}p_{10} - k_{4br}p_5x_8 - k_{4b}p_{11} - k_{h1}p_4, \\
0 &= -k_{3br}p_3x_6 - k_{3br}x_3p_6 + k_{3b}p_{10} - k_{4a}p_6x_7 - k_{4a}x_6p_7 + k_{4ar}p_{11}, \\
0 &= -k_{4a}p_6x_7 - k_{4a}x_6p_7 + k_{4ar}p_{11} + k_{h2}p_8, \\
0 &= -k_{4br}p_5x_8 - k_{4br}x_5p_8 + k_{4b}p_{11} - k_{h2}p_8, \\
0 &= p_1 + p_2 + p_9, \\
0 &= p_3 + p_4 + p_9 + p_{10}, \\
0 &= p_5 + p_6 + p_{10} + p_{11}, \\
0 &= p_7 + p_8 + p_{11}.
\end{align*}
$$

2. We substitute, in the equations above, the steady-state value when $k_s = 0$ and obtain:

$$
\begin{align*}
0 &= -\overline{H} - k_sp_1(0) - k_{2br}p_1(0) - k_{2br}\overline{H}p_4(0) + k_{2b}p_9(0), \\
0 &= -k_{2a}\overline{C}_p_2(0) - k_{2a}p_3(0) + k_{2ar}p_9(0) - k_{3br}p_3(0) - k_{3br}\overline{C}_p_6(0) + k_{3b}p_{10}(0) + k_{h1}p_4(0), \\
0 &= -k_{2br}p_1(0) - k_{2br}\overline{H}p_4(0) + k_{2b}p_9(0) - k_{3a}\overline{T}_p_4(0) - k_{3a}p_5(0) + k_{3ar}p_{10}(0) - k_{h1}p_4(0), \\
0 &= -k_{3a}\overline{T}_p_4(0) - k_{3a}p_5(0) + k_{3ar}p_{10}(0) - k_{4br}p_5(0) - k_{4b}\overline{T}_p_8(0) + k_{4ar}p_{11}(0), \\
0 &= -k_{3br}p_3(0) - k_{3br}\overline{C}_p_6(0) + k_{3b}p_{10}(0) - k_{4a}\overline{R}_p_6(0) - k_{4a}p_7(0) + k_{4ar}p_{11}(0), \\
0 &= -k_{4a}\overline{R}_p_6(0) - k_{4a}p_7(0) + k_{4ar}p_{11}(0) + k_{h2}p_8(0), \\
0 &= -k_{4br}p_5(0) - k_{4br}\overline{T}_p_8(0) + k_{4b}p_{11}(0) - k_{h2}p_8(0), \\
0 &= p_1(0) + p_2(0) + p_9(0), \\
0 &= p_3(0) + p_4(0) + p_9(0) + p_{10}(0), \\
0 &= p_5(0) + p_6(0) + p_{10}(0) + p_{11}(0), \\
0 &= p_7(0) + p_8(0) + p_{11}(0).
\end{align*}
$$

This system is linear in $p_1(0), \ldots, p_{11}(0)$ and hence the derivatives of $x_i$ at $k_s = 0$ can be found by solving the system. We have solved it using Maple. In particular, we have obtained that

$$
p_8(0) = \frac{k_{1a}k_{3b}k_{3a}k_{4b}\overline{R}(k_{h1} + k_{3a}\overline{T}) + k_{h1}k_{3ar}(k_{h2}k_{4ar}(k_{3br}\overline{C} + k_{4a}\overline{R}) + k_{3br}\overline{C}(k_{h2}k_{4br} + k_{4b}k_{4ar}\overline{T}))}{k_{h2}k_{4a}k_{3b}k_{4b}\overline{R}(k_{h1} + k_{3a}\overline{T})}
$$

This is the derivative of the signal-response curve at $k_s = 0$.

3. We repeat the steps above one more time: we compute the derivative with respect to $k_s$ of the above equations (step 1). We evaluate the resulting equations at the steady state for $k_s = 0$ and at $p_i = p_i(0)$. We obtain a linear system in the second derivatives of $x_i$ at $k_s = 0$ which can be solved in Maple. As a result, we obtain the second derivative of $x_8$ with respect to $k_s$ at $k_s = 0$ as desired.
Before showing what the second derivative of \( x_8 \) with respect to \( k_s \) at \( k_s = 0 \) is, it is convenient to introduce new parameters. For \( i = 2, 3, 4 \), let
\[
  k_{iy} = \frac{k_{ia}}{k_{iar} + k_{ib}} \quad k_{iyr} = \frac{k_{iar}}{k_{iar} + k_{ib}} \quad k_i = k_{ib} k_{iy} \quad k_{ir} = k_{iar} k_{iyr}.
\]

For an interpretation of these constants see below. In particular, we take \( k_i, k_{ir} \) to be the rates of forward and reverse phosphorylation at each layer.

With these new constants, we obtain that the derivative of \( x_8 \) with respect to \( k_s \) at \( k_s = 0 \), that is \((S25)\), becomes
\[
  p_8(0) = \frac{k_3 k_4 H R T}{k_4 k_{h2} (k_{h1} + k_3 T R) + k_3 r k_{h1} (k_{h2} + k_4 r T) C}.
\]

This expression is identical to the first derivative of \( x_8 \) with respect to \( k_s \) at \( k_s = 0 \) for the model without intermediates, as given in \((S21)\).

Similarly, the sign of the second derivative of \( x_8 \) with respect to \( k_s \) at \( k_s = 0 \) equals the sign of:
\[
  S_y = S - C H k_2 k_3 k_{h1} (k_{i4y} (k_{4r} T + k_{h2}) + T k_4 k_{4yr}) (R z_2 + C T z_1) \\
  - C H k_2 k_3 k_{h1} (z_2 - T z_1) (C z_1 + R k_{h2} k_4) k_3 y + C (k_{4r} T + k_{h2}) k_3 k_{3yr} \\
  - C (k_{h1} z_1 (C - H) + z_3) (H (C z_1 + R k_{h2} k_4) (k_2 k_{2yr} + k_2 r k_{2y}) + (C k_{h1} z_1 + z_3) k_2 y)
\]

where \( S \) is given in \((S22)\) and
\[
  z_1 := k_3 y (k_{4r} T + k_{h2}), \quad z_2 := k_2 h_2 (k_4 R + k_{3r} C), \quad z_3 := R k_{h2} k_4 (k_3 T + k_{h1}).
\]

Recall that the condition for sigmoidality is \( S_y > 0 \). We have marked in blue the only terms that can cause the term \( S_y \) to be positive. Namely, if \( S \) is negative (that is, the model without intermediates is hyperbolic), \( C > H \) and \( z_2 > T z_1 \), then the model with intermediates is hyperbolic as well.

We deduce easily that
- If \( k_{h1} = 0 \), then sigmoidality cannot occur.
- If \( k_{3r} = 0 \) then \( z_1 = 0 \) and \( S < 0 \), and hence sigmoidality cannot occur.

**Interpretation of the new rate constants.** The rate constants \( k_{iy}, k_{iyr} \) are the reciprocal of the Michaelis-Menten constants of each intermediate \( Y_i \) in each direction. These are the coefficients of the expression in \( x_1, \ldots, x_8 \) obtained by imposing \( \dot{x}_9 = \dot{x}_{10} = \dot{x}_{11} = 0 \) and solving for \( x_9, x_{10}, x_{11} \). In particular, at steady state we have:
\[
  x_9 = \frac{k_{2a}}{k_{2ar} + k_{2b}} x_2 x_3 + \frac{k_{2br}}{k_{2ar} + k_{2b}} x_1 x_4 = k_{2y} x_2 x_3 + k_{2yr} x_1 x_4 \\
  x_{10} = \frac{k_{3a}}{k_{3ar} + k_{3b}} x_4 x_5 + \frac{k_{3br}}{k_{3ar} + k_{3b}} x_3 x_6 = k_{3y} x_4 x_5 + k_{3yr} x_3 x_6 \\
  x_{11} = \frac{k_{4a}}{k_{4ar} + k_{4b}} x_6 x_7 + \frac{k_{4br}}{k_{4ar} + k_{4b}} x_5 x_8 = k_{4y} x_6 x_7 + k_{4yr} x_5 x_8.
\]

If we plug these values into the ODEs \( \dot{x}_i, i = 1, \ldots, 8 \), we obtain a mass-action system for the model without intermediates with rate constants \( k_i = k_{ib} k_{iy} \) and \( k_{ir} = k_{iar} k_{iyr} \).
1.6 Model with production and degradation

We investigate if the conditions for sigmoidality are altered by the introduction of production and degradation in the model.

Reactions, equations and steady states. We consider the system with reactions as described in Subsection 1.1 together with degradation reactions for all species:

\[
\begin{align*}
    \text{HK} & \xrightarrow{k_{10}} 0 & \text{HKp} & \xrightarrow{k_{20}} 0 & \text{REC} & \xrightarrow{k_{30}} 0 & \text{RECp} & \xrightarrow{k_{40}} 0 \\
    \text{Hpt} & \xrightarrow{k_{50}} 0 & \text{Hptp} & \xrightarrow{k_{60}} 0 & \text{RR} & \xrightarrow{k_{70}} 0 & \text{RRp} & \xrightarrow{k_{80}} 0
\end{align*}
\]

and production reactions for the unphosphorylated forms:

\[
\begin{align*}
    0 & \xrightarrow{k_{1i}} \text{HK} & 0 & \xrightarrow{k_{3i}} \text{REC} & 0 & \xrightarrow{k_{5i}} \text{Hpt} & 0 & \xrightarrow{k_{7i}} \text{RR}
\end{align*}
\]

We define as usual

\[
\begin{align*}
    x_1 &= [\text{HK}] & x_2 &= [\text{HKp}] & x_3 &= [\text{REC}] & x_4 &= [\text{RECp}] \\
    x_5 &= [\text{Hpt}] & x_6 &= [\text{Hptp}] & x_7 &= [\text{RR}] & x_8 &= [\text{RRp}].
\end{align*}
\]

The dynamics of the concentrations in time is modeled with a system of ODEs:

\[
\begin{align*}
    \dot{x}_1 &= -k_s x_1 - k_2 x_1 x_4 + k_2 x_2 x_3 - k_{1o} x_1 + k_{1i} \\
    \dot{x}_2 &= -k_s x_1 + k_2 x_1 x_4 - k_2 x_2 x_3 - k_{2o} x_2 \\
    \dot{x}_3 &= -k_2 x_2 x_3 + k_{h1} x_4 + k_{2o} x_1 x_4 + k_3 x_4 x_5 - k_{3o} x_3 x_6 - k_{3o} x_3 + k_{3i} \\
    \dot{x}_4 &= k_2 x_2 x_3 - k_{h1} x_4 - k_2 x_1 x_4 - k_3 x_4 x_5 + k_{3o} x_3 x_6 - k_{4o} x_4 \\
    \dot{x}_5 &= -k_2 x_4 x_5 + k_{3r} x_3 x_6 + k_4 x_6 x_7 - k_{4r} x_5 x_8 - k_{5o} x_5 + k_{5i} \\
    \dot{x}_6 &= k_3 x_4 x_5 - k_{3r} x_3 x_6 - k_4 x_6 x_7 + k_{4r} x_5 x_8 - k_{6o} x_6 \\
    \dot{x}_7 &= -k_4 x_6 x_7 + k_{h2} x_8 + k_{4r} x_5 x_8 - k_{7o} x_7 + k_{7i} \\
    \dot{x}_8 &= k_4 x_6 x_7 - k_{h2} x_8 - k_{4r} x_5 x_8 - k_{8o} x_8.
\end{align*}
\]

The system does not have any conservation law. Thus, the steady-state equations are given by setting the derivative of the concentration to zero, that is \( \dot{x}_i = 0 \).

Hyperbolic vs. sigmoidal. The procedure applied to our initial system in Subsection 1.2 to obtain the inverse of the signal-response curve, can be applied here to obtain an analytical expression of the inverse of the signal-response curve. The role of the total amounts \( \mathbb{P}, \mathcal{C}, \mathcal{T} \) and \( \mathcal{R} \) is played by the quotients

\[
\begin{align*}
    K_1 &= \frac{k_{1i}}{k_{1o}}, & K_3 &= \frac{k_{3i}}{k_{3o}}, & K_5 &= \frac{k_{5i}}{k_{5o}}, & K_7 &= \frac{k_{7i}}{k_{7o}}.
\end{align*}
\]

We do not reproduce the analysis here again. The procedure leads to the derivative of the signal-response curve at zero. Alternatively, we can apply the procedure described in the previous subsection to directly obtain the sign of the second derivative of the signal-response curve at zero, without explicitly computing the signal-response curve.

The expression of the second derivative of the signal-response curve at zero is very large, and hence we only provide here the positive monomials with the aim of determining what architectures can exhibit sigmoidality.
We use the definition of $K_1, K_3, K_5, K_7$ above, together with

$$K_2 = \frac{k_{2o}}{k_{1o}}, \quad K_4 = \frac{k_{4o}}{k_{3o}}, \quad K_6 = \frac{k_{6o}}{k_{5o}}, \quad K_8 = \frac{k_{8o}}{k_{7o}}.$$ 

The positive terms that can lead to sigmoidality are then

$$S_{o,pos} = k_2 K_3 K_1 (k_{8o} + k_{4r} K_5 + k_{h2}) (k_{4r} k_{6o} k_3^2 K_5^2 K_6 (k_2 K_3 + k_{2o})$$

$$+ K_4 K_3 k_2 k_{3r} (h_1 + k_{4o}) (k_{4r} k_5 (k_3 + k_{6o}) + (k_{8o} + k_{h2}) (k_4 K_7 + k_{3r} K_3 + k_{6o})))).$$

We observe that if $k_{3r} = k_{4r} = 0$ then $S_{o,pos} = 0$ and sigmoidality cannot occur. Contrary to the system without production and degradation, $k_{3r} = 0$ does not guarantee that sigmoidality cannot occur. This is due to the fact that now there is a degradation of Hptp, which plays the role of the hydrolysis $k_{h1}$ at RECp. Therefore, the reverse phosphorelay between layers 3 and 4 can also account for sigmoidality.

In the system with production/degradation reactions, inclusion of intermediates cannot alter steady-state properties such as the existence of sigmoidality. In recent work, we have shown that in reaction schemes that do not give rise to conservation relations, consideration of complex formation does not alter the system properties at steady state [1].

### 1.7 Model with auto-dephosphorylation at HK

We extend the model given in Subsection 1.1 to incorporate auto-dephosphorylation of HK. That is, we enrich the system with a reaction

$$\text{HKp} \xrightarrow{k_5} \text{HK}.$$ 

The ODE system modeling the protein concentrations is identical to (S1) except for the expressions for $\dot{x}_1, \dot{x}_2$ that become

$$\dot{x}_1 = -k_s x_1 - k_{2r} x_1 x_4 + k_{2x} x_2 x_3 + k_5 x_2$$

$$\dot{x}_2 = k_s x_1 + k_{2r} x_1 x_4 - k_{2x} x_2 x_3 - k_5 x_2.$$ 

The steps followed in Subsection 1.1 can be applied here as well to obtain an analytical expression of the inverse of the signal-response curve. The sign of the second derivative of the signal-response curve at zero agrees with the sign of

$$S_d := (k_5 + k_2 C) S - k_2 k_5 C \left((k_3 T + k_2 T) k_{4} k_{h2} z_3 R + z_1 k_{3} k_2 C R (k_3 z_2 C + k_4 k_{h2} R) + k_{h1} z_3^2\right)$$

where

$$z_1 = k_{3r} T + k_{h2}, \quad z_2 = k_{4} T + k_{h2}, \quad z_3 = k_{3} k_{h2} R + k_{3r} z_1 C,$$

and $S$ is given in (S22). We easily see that $S_d$ can only be positive if $S$ is positive. Therefore, the necessary conditions for sigmoidality established in the main text for the simple model are not altered by explicitly modeling auto-dephosphorylation of HK.
1.8 Model with auto-dephosphorylation at Hpt

We extend the model given in Subsection 1.1 to incorporate auto-dephosphorylation of Hpt. That is, we enrich the system with a reaction

\[ \text{Hpt} \xrightarrow{k_5} \text{Hpt}. \]

We applied the steps described in Subsection 1.5 to obtain an expression of the sign of the second derivative of the signal-response curve at zero in terms of the rate constants and total amounts. The sign is given by

\[ S_3 = S - k_5 z_1 \left( z_1 z_2 (k_{h1} k_{3r} C + k_3 k_5 T) + (k_{2r} H + k_{h1})(k_{3r} C + k_5) + 2 k_4 k_{h2} z_2 (Hk_{2r} + z_2) R \right. \]

\[ + C(H(k_2 k_3^2 (k_4 - k_{4r}) T^2 + (k_2 k_3 k_4 - k_{2r} k_{3r} k_{4r} + k_{2r} k_{3r} k_{4r}) k_{h1} T) \]

\[ \left. + k_{h1} k_{h2} (k_2 k_3 - k_2 k_{3r} + k_{2r} k_{3r}) \right) \]

where

\[ z_1 = k_{4r} T + k_{h2}, \quad z_2 = k_3 T + k_{h1} \]

and \( S \) is given in (S22). The terms highlighted in blue are the terms that can possibly contribute to \( S_3 \) being positive. The last two highlighted terms are also highlighted in \( S \) in (S22) and further, they are multiplied by \( k_{h1} \).

When \( k_5 \) is set to zero, the sign of the second derivative of the signal-response curve at zero agrees with the corresponding sign for the model without auto-dephosphorylation at Hpt. However, when \( k_5 \neq 0 \) then sigmoidality can arise even if \( k_{h1} = k_{3r} = 0 \) but \( k_{4r} > k_4 \). For this model, necessary conditions for sigmoidality of the signal-response curve are either that

\[ k_{h1} > 0 \quad \text{and} \quad k_2 > k_{2r} \quad \text{and} \quad (k_{3r} > k_3 \quad \text{or} \quad k_{3r} k_{4r} > k_3 k_4). \]

or that

\[ k_5 > 0 \quad \text{and} \quad k_{4r} > k_4. \]

In other words, necessary conditions for sigmoidality of the signal-response curve are

\[ k_{h1} (k_2 - k_{2r})(k_{3r} - k_3) \neq 0 \quad \text{or} \quad k_{h1} (k_2 - k_{2r})(k_{3r} k_{4r} - k_3 k_4) \neq 0 \quad \text{or} \quad k_5 (k_{4r} - k_4) \neq 0. \]

2 The kinase HK is bifunctional

We consider the case in which the kinase HK is bifunctional, that is, HK acts as a phosphatase for REC.

2.1 Reactions, equations and steady states

Reactions. The minimal set of reactions that the system has, consists of the autophosphorylation reaction on HK and the forward phosphotransfer reactions

\[ \text{HK} \xrightarrow{k_s} \text{HKp} \quad \text{HKp} + \text{REC} \xrightarrow{k_2} \text{HK} + \text{RECp} \]

\[ \text{RECp} + \text{Hpt} \xrightarrow{k_3} \text{REC} + \text{Hptp} \quad \text{Hptp} + \text{RR} \xrightarrow{k_4} \text{Hpt} + \text{RRp}, \]
together with the reaction for phosphatase activity of the histidine kinase HK:

$$HK + RECp \xrightarrow{k_5} HKRECp \xrightarrow{k_6} HK + REC.$$  

We study the extensions of this minimal reaction mechanism obtained by adding reverse phospho-
transfer reactions involving the HK, REC, Hpt and the RR, as well as hydrolysis reactions at RECp and RRp:

$$HK + RECp \xrightarrow{k_2} HKp + REC \quad RECp \xrightarrow{k_{h1}} REC \quad \text{REC + Hptp} \xrightarrow{k_{3r}} \text{RECp + Hpt} \quad \text{RRp \xrightarrow{k_{h2}} RR} \quad \text{Hpt + RRp \xrightarrow{k_{4r}} Hptp + RR.}$$

By setting some of the rate constants $k_{s}$ or $kh_{s}$ to zero, we obtain a total of 32 different topologies involving different combinations of reverse phosphorelay and hydrolysis reactions (Table 1 and Table S1), similarly to the situation where the HK kinase is monofunctional. The minimal set of reactions is always part of the system, meaning that $k_{2}, k_{3}, k_{4}, k_{s} \neq 0$ for all topologies.

**Ordinary differential equations.** We model the protein concentrations in the system using ordinary differential equations (ODEs). To simplify the notation, we define:

$$x_1 = [HK] \quad x_2 = [HKp] \quad x_3 = [REC] \quad x_4 = [RECp] \quad x_5 = [Hpt] \quad x_6 = [Hptp] \quad x_7 = [RR] \quad x_8 = [RRp] \quad x_9 = [HKRECp].$$

The dynamics of the concentrations in time is modeled with a system of ODEs:

\[
\begin{align*}
\dot{x}_1 &= -k_s x_1 - k_{2r} x_1 x_3 + k_2 x_2 x_3 - k_5 x_1 x_4 + k_{5r} x_9 + k_6 x_9 \\
\dot{x}_2 &= k_s x_1 + k_{2r} x_1 x_4 - k_2 x_2 x_3 \\
\dot{x}_3 &= -k_2 x_2 x_3 + kh_1 x_4 + k_{2r} x_1 x_4 + k_3 x_4 x_5 - k_{3r} x_3 x_6 + k_6 x_9 \\
\dot{x}_4 &= k_2 x_2 x_3 - k_{h1} x_4 - k_{2r} x_1 x_4 - k_3 x_4 x_5 + k_{3} x_3 x_6 - k_5 x_1 x_4 + k_{5r} x_9 \\
\dot{x}_5 &= -k_3 x_4 x_5 + k_{3r} x_3 x_6 + k_{4r} x_6 x_7 - k_{4r} x_5 x_8 \\
\dot{x}_6 &= k_3 x_4 x_5 - k_{3r} x_3 x_6 - k_{4r} x_6 x_7 + k_{4r} x_5 x_8 \\
\dot{x}_7 &= -k_4 x_6 x_7 + k_{h2} x_8 + k_{4r} x_5 x_8 \\
\dot{x}_8 &= k_4 x_6 x_7 - k_{h2} x_8 - k_{4r} x_5 x_8 \\
\dot{x}_9 &= k_5 x_1 x_4 - k_{5r} x_9 - k_6 x_9.
\end{align*}
\]

This system has four conserved amounts:

$$\Pi = x_1 + x_2 + x_9, \quad C = x_3 + x_4 + x_9, \quad T = x_5 + x_6, \quad R = x_7 + x_8.$$  

It is assumed that total amounts are positive.

**Steady-state equations.** We proceed as in the monofunctional case and conclude that the steady states of the system are given as the solutions to the following system of equations:
and equals \( x \) Zero concentrations. We assume that \( k \) as a phosphatase for the dephosphorylation of the receiver protein. Additionally, we are assuming

\[ R \text{ and leave (S35) and (S36) as they are. This results in the following equivalent system of equations:} \]

Rearrangement of the steady-state equations. We change equations \((S32)-(S36)\) by linear combinations of them. Specifically, we replace:

- \((S34)\) by \((S34)+(S35)\),
- \((S33)\) by \((S33)+(S34)+(S35)+(S36)\),
- \((S32)\) by \((S32)+(S33)+(S34)+(S35)+(S36)\),

and leave \((S35)\) and \((S36)\) as they are. This results in the following equivalent system of equations:

\[ \bar{P} = x_1 + x_2 + x_9 \quad (S37) \]
\[ \bar{C} = x_3 + x_4 + x_9 \quad (S38) \]
\[ \bar{T} = x_5 + x_6 \quad (S39) \]
\[ \bar{R} = x_7 + x_8 \quad (S40) \]

Zero concentrations. We assume that \( k_5, k_{5r}, k_6 \neq 0 \), that is, the kinase is bifunctional and acts as a phosphatase for the dephosphorylation of the receiver protein. Additionally, we are assuming that \( k_s, k_2, k_3, k_4 \neq 0 \) and that all total amounts are positive.

In this scenario, for the topologies topologies 9-13, 15, the signal-response curve is constant and equals \( x_8 = \bar{R} \). All the other topologies have non-constant signal-response curves and all concentrations at steady state are non-zero (see Appendix B.1).

The topologies 17-24 \((k_{h1} = k_{h2} = 0)\) exhibit signal-response curves defined piecewise and are treated differently. Zero steady-state values occur (see Appendix B.3).

2.2 Steady-state relations and signal-response curve

We proceed to find an expression for the signal-response curve. As in the previous case, the signal is taken to be the value of the rate constant \( k_s \), and the response is the steady-state value of phosphorylated response regulator \((x_8)\) corresponding to \( k_s \) (with all the other rate constants and total amounts fixed).

Contrary to the previous system, this case does not allow for an explicit analytical relation. Instead, we infer the existence of an analytical function relating \( k_s \) and \( x_8 \) and derive properties of this function. We start by establishing relations between each concentration \( x_i \) and \( x_8 \) at steady state.

Steady-state relations. We study here the steady-state solutions that do not have vanishing concentrations. We let

\[ k_y = \frac{k_5}{k_{5r} + k_6} \quad (S46) \]
be the reverse of the Michaelis-Menten constant of HK. We express the concentrations \( x_1, \ldots, x_7 \) at steady state in terms of \( x_8 \) and \( x_9 \), independently of \( k_S \). In addition, we find a relation between \( x_8 \) and \( x_9 \) at steady state.

<table>
<thead>
<tr>
<th>Expression</th>
<th>Behavior as function of ( x_8 ) and ( x_9 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( x_7 = \bar{R} - x_8 )</td>
<td>( x_7 ) decreases in ( x_8 )</td>
</tr>
<tr>
<td>( x_6 = x_8(k_4\bar{T} + k_h^2) )</td>
<td>( x_6 ) increases in ( x_8 )</td>
</tr>
<tr>
<td>( x_5 = \frac{k_4T_{x_7} - k_h^2x_8}{k_4T_{x_7} + k_2x_8} )</td>
<td>( x_5 ) decreases in ( x_8 )</td>
</tr>
<tr>
<td>( x_4 = \frac{k_{12}x_8 + k_{3r}x_8(C - x_8)}{k_3x_5 + k_{3r}x_6} )</td>
<td>( x_4 ) increases in ( x_8 ) and decreases in ( x_9 )</td>
</tr>
<tr>
<td>( x_3 = \frac{k_{12}x_5(C - x_8) - k_h^2x_8}{k_3x_5 + k_{3r}x_6} )</td>
<td>( x_3 ) decreases in ( x_8 ) and in ( x_9 )</td>
</tr>
<tr>
<td>( x_2 = \frac{k_{23}x_4(\bar{T} - x_9) + k_6x_9 + k_{3r}x_9 + k_{12}x_4 + k_{3r}x_9}{k_{23}x_4 + k_{2r}x_4} )</td>
<td>( x_2 ) increases in ( x_8 )</td>
</tr>
<tr>
<td>( x_1 = \frac{k_{23}x_4(\bar{T} - x_9) - k_6x_9 - k_{3r}x_9 - k_{12}x_4 - k_{3r}x_9}{k_{23}x_4 + k_{2r}x_4} )</td>
<td>( x_1 ) decreases in ( x_8 ) and in ( x_9 )</td>
</tr>
<tr>
<td>( x_9 = g(x_8) )</td>
<td></td>
</tr>
</tbody>
</table>

The first seven rows of (S47) give an iterative way to find the steady-state values of concentrations \( x_1, \ldots, x_7 \) once the values of \( x_8 \) and \( x_9 \) are known. The last entry gives the relation between \( x_9 \) and \( x_8 \). See Appendix B.2 for a proof.

The steady-state values are all positive if and only if \( x_8 \) is in the interval \((0, \alpha)\), where \( \alpha \) is the first positive root of the degree-2 polynomial \( q_2 \) given in (S19) (that is, the same as for the monofunctional case).

In (S47), the variables \( x_1, \ldots, x_7 \) are expressed as functions of \( x_8, x_9 \). The variable \( x_9 \) cannot explicitly be written as a function of \( x_8 \). The function \( g \) is known to exist, but we do not have an analytical expression of it. However, there is a procedure to obtain the steady-state value of \( x_9 \) corresponding to a given value of \( x_8 \). For each fixed \( x_8 \) strictly between 0 and \( \alpha \), \( x_9 \) is the first positive root of the following polynomial \( G(x_8, x_9) \):

\[
G(x_8, x_9) = c_0(x_8) + c_1(x_8)x_9 + c_2(x_8)x_9^2 + c_3(x_8)x_9^3,
\]

where if we denote

\[
 z_1 := k_{3r}(k_4\bar{T} + k_h^2), \quad z_2 := k_2k_3(k_4\bar{T} + k_h^2), \quad z_3 := k_4r - k_4, \quad z_4 := \frac{-z_1}{k_3},
\]

\[
 c_0(x_8) = z_1, \quad c_1(x_8) = z_2, \quad c_2(x_8) = z_3, \quad c_3(x_8) = z_4
\]
then the coefficients \( c_i(x_8) \) are:

\[
c_0(x_8) = k_y x_8 (k_{h2}(k_4R + z_3 x_8) + z_1 C) (k_{h2}(z_3(k_{h1} + k_2 \Pi) + z_4) x_8^2 \\
+ (C(k_{h1} z_1 + z_2 \Pi) + k_{h2} k_4 R(k_2 \Pi + k_3 T + k_{h1} )) x_8 - k_{h2} k_4 C H T R)
\]

\[
c_1(x_8) = (k_{2y} z_1 - z_2) C x_8 - k_{h2}(k_2 - k_{2y})(k_4 R + z_3 x_8) x_8 + k_{h2} k_4 C T R (k_{3y} T R + z_4 x_8)
\]

\[
c_2(x_8) = (k_{h2} z_3(z_2 + k_{2y} z_1)x_8^2 + (k_{h1} z_1^2 + (2z_2 C + k_{h2} k_4 R - k_6 z_4 + z_2 \Pi) z_1
\]

\[
- k_{3y} k_2 \Pi (k_{2y} k_4 z_3 T - z_2) x_8 - k_{3y} k_4 T R (k_{3y} k_4 R + z_1 (k_6 + 2C k_2 + k_4 \Pi))) x_8 k_y
\]

\[
c_3(x_8) = z_1 (k_{3y} k_4 \Pi - z_2 x_8) k_y x_8
\]

**Signal-response expression.** Using the remaining steady-state equation, (S41), we express \( k_s \) in terms of \( x_8 \):

\[
k_s = f_b(x_8) = \frac{k_{h1} x_4 + k_{h2} x_8 + k_6 x_9}{x_1}, \tag{S48}
\]

where \( x_1, x_4, x_9 \) are given in terms of \( x_8 \) as well (see (S47) above) and \( x_1 \neq 0 \) at steady state. If \( x_8 \) approaches \( \alpha \), then \( x_1 \) tends to zero and \( k_s \) tends to infinity. It follows that \( \alpha \) is precisely the maximal response of \( x_8 \).

The function \( f_b \) is continuous and differentiable in \([0, \alpha]\) and is strictly increasing (see Appendix B.3). It admits an inverse

\[
\varphi_b = f_b^{-1},
\]

which is the signal-response curve. The signal-response curve is increasing, continuous and differentiable in \([0, +\infty)\). When \( k_s \) tends to infinity then the response \( x_8 \) tends to \( \alpha \).

If \( k_{h1} = k_{h2} = 0 \) then \( \varphi_b \) is defined by \( f_b^{-1} \) if \( k_s \in [0, k_6 k_y C] \) and \( \varphi_b = \Pi \) for \( k_s > k_6 k_y C \).

**Practical considerations.** In order to plot the signal-response curve we use the following procedure:

(i) Compute \( \alpha \) (the first positive root of \( q_2(x_8) \) in \( x_8 \)) and choose a grid of values for \( x_8 \), strictly between 0 and \( \alpha \).

(ii) For each value of \( x_8 \), find the first positive root of \( G(x_8, x_9) \) as a function of \( x_9 \), that is, for each value of \( x_8 \) we find a value of \( x_9 \).

(iii) Compute \( x_1, x_4 \) using (S47) and the pair of values \( (x_8, x_9) \).

(iv) Compute \( k_s \) using (S48) in terms of \( x_1, x_4, x_8, x_9 \).
In this way, points on the signal-response curve \((f_b(x_8), x_8)\) are generated. Because of the relationship between \(f_b\) and \(\varphi_b\), the points give a plot of the function \(\varphi_b\).

**Comparison of the monofunctional and bifunctional cases.** We have shown that the maximal response of phosphorylated response regulator \(x_8 = RRP\) is independent of whether the kinase is bifunctional or not. In particular, the value is independent of the rate constants \(k_5, k_{5r}, k_6\).

However, the signal-response curve in the bifunctional case is always below the signal-response curve in the monofunctional case. Indeed, if the kinase is monofunctional, then the expressions in \([S18]\) can be obtained from those in \([S47]\) by setting \(x_9 = 0\). It follows that if the common reactions have the same rate constants in the two cases, then for every fixed \(x_8\), the value of \(x_4\) (resp. \(x_1\)) in the bifunctional case is smaller (resp. larger) than in the monofunctional case. Consequently, for any rate constants \(k_5, k_{5r}, k_6\), we have \(f_b(x_8) < f(x_8)\). In other words, the signal \(k_s\) required to achieve a certain response \(x_8\) is smaller in the monofunctional case than in the bifunctional case. Nevertheless, as \(k_s\) increases, the steady-state value of \(x_8\) tends to the maximal response (which is the same value in both cases). Therefore, the difference between the steady-state value of \(x_8\) in the two cases becomes negligible for large values of \(k_s\).

This is due to sequestration of substrate in \(x_9\). Therefore, the signal-response curve in the monofunctional case is always above the corresponding curve in the bifunctional case (for any choice of additional rate parameters). Since the maximal response is independent of the role of the kinase, a smaller signal is required to get close to the maximal response, when the kinase is monofunctional. Furthermore, if the inverse of the Michaelis-Menten constant \(k_y = k_5/(k_{5r} + k_6)\) (equation \([S46]\)) increases and \(k_6\) is fixed, then \(f_b(x_8)\) increases for a fixed \(x_8\) (see Appendix \(B.3\)).

### 2.3 Hyperbolic and sigmoidal signal-response curves

We apply the same indicator to classify a curve as sigmoidal or hyperbolic as in the previous case. That is, we calculate the sign of the second derivative of the signal-response curve at zero and classify the curve accordingly. We have computed \(\varphi_b''(0)\) using the method introduced in Subsection \(1.5\).

If \(\varphi_b''(0) > 0\) then we classify the signal-response curve as sigmoidal, and if \(\varphi_b''(0) < 0\) then we classify the signal-response curve as hyperbolic.

The sign of the second derivative of \(\varphi_b\) at 0 agrees with the sign of

\[
S_b = S + H(\alpha_2 H + \alpha_3)k_y + \alpha_1 H k_y^2
\]  
(849)

where \(k_y = \frac{k_5}{k_{5r} + k_6}\) and

\[
\alpha_1 = C(k_{4r} T + k_{h2})H^2k_2k_6k_{3r}(k_{3r}C(k_{4r}T + k_{h2}) + k_4k_{h2}R)
\]

\[
\alpha_2 = -C(k_{4r} T + k_{h2})k_2(Ck_{3r}k_6k_{3r}(k_{4r} T + k_{h2}) - k_{3r}^2(k_{h1} C + k_6)(k_{4r} T + k_{h2})
+ k_4k_{h2}R(k_6(k_{3} - k_{3r}) - k_{h1}k_{3r}))
\]

\[
\alpha_3 = -C(k_{3r}C(k_{4r} T + k_{h2}) + k_4k_{h2}R)(k_{3r}k_{h1}(k_{4r} CT + k_{h2}) + k_4k_{h2}R(k_{3}T + k_{h1}))(k_{3}C + k_6)
\]

The term \(\alpha_1\) is always positive and the term \(\alpha_3\) is always negative. The independent term (obtained by setting \(k_y = 0\)) is identical to the term given in the monofunctional case \([S22]\). We have that if \(k_{3r} = 0\), then \(S_b < 0\) and the function \(\varphi_b\) is hyperbolic. However, when \(k_{h1} = 0\) the system can show sigmoidality (because \(\alpha_1 \neq 0\)).

Observe that the leading coefficient of the term \(S_b\) in \([849]\) seen as a polynomial of degree 2 in \(k_y\) is positive. Therefore, by increasing \(k_y\) enough, \(S_b\) becomes positive and the curve sigmoidal. Recall that \(k_y\) is the inverse of the Michaelis-Menten constant of the enzyme HK for its dephosphorylation
activity. Therefore, increasing \( k_y \) corresponds to making the enzyme mediated dephosphorylation of RECp more efficient.

A Proof of the claims: monofunctional case

This appendix provides the proofs of the claims in the Section 1.

A.1 Constant signal-response curves and zero solutions

- \( k_{h1} = k_{h2} = 0 \): From (S14) we have that \( x_1 = 0 \) (since \( k_s \neq 0 \)) and thus \( x_2 = \overline{H} \) at steady state. From (S15) it follows that \( x_3 = 0 \) (since \( k_2 \neq 0 \)) and thus \( x_4 = \overline{C} \). From (S16) we have that \( x_5 = 0 \) (since \( k_3 \neq 0 \)) and so \( x_6 = \overline{T} \) and finally from (S17) it follows that \( x_7 = 0 \) (since \( k_4 \neq 0 \)) and \( x_8 = \overline{R} \). This fact is independent of the value of all other constants.

- \( k_{h2} = k_{3r} = 0 \): From (S16) either \( x_4 = 0 \) or \( x_5 = 0 \). If \( x_4 = 0 \) we have that \( x_3 \neq 0 \). From (S14), \( x_1 = 0 \) and from (S15) we have \( x_2 = 0 \) which is a contradiction. Therefore \( x_5 = 0 \) and consequently, \( x_6 = \overline{T} \neq 0 \) and \( x_7 = 0 \). Thus \( x_8 = \overline{R} \) and \( x_1, \ldots, x_4 \) fulfill (S14), (S15) and the equations for the total amounts. Additionally, the signal-response curve for Hpt is also constant (that is, at steady state \( x_6 = \overline{T} \)), but the curves corresponding to HK and REC are not constant.

- \( k_{h2} = k_{4r} = 0 \): From (S17) either \( x_6 = 0 \) or \( x_7 = 0 \). If \( x_6 = 0 \) we have that \( x_5 \neq 0 \). From (S16) \( x_4 = 0 \) and from (S14) we have \( x_1 = 0 \). By (S15) we have \( x_2x_3 = 0 \) which is a contradiction. Therefore \( x_7 = 0 \), and then \( x_8 = \overline{R} \neq 0 \). \( x_1, \ldots, x_6 \) fulfill (S14)-(S16) and the equations for the total amounts. In this case the signal-response curves for HK, REC and Hpt are not constant.

Assume that none of the cases above hold, that is, that either \( k_{h2} \neq 0 \) or \( k_{h2} = 0 \) and \( k_{h1}k_{3r}k_{4r} \neq 0 \). Assume that all total amounts are positive. We show that in this case any non-negative solution to the steady-state equations is positive, that is, all concentrations are non-zero. As a consequence, the signal-response curve cannot be constant (equal to \( \overline{T} \); because this would imply \( x_7 = 0 \)).

- If \( x_1 = 0 \), then by (S14) it must be that \( k_{h1}x_4 = k_{h2}x_8 = 0 \) and \( x_2 \neq 0 \). From (S15) it follows that \( x_3 = 0 \) and hence \( x_1 \neq 0 \) (because \( \overline{C} > 0 \)) and thus \( k_{h1} = 0 \). In this case \( k_{h2} \neq 0 \) and hence \( x_8 = 0 \). From (S16) we have \( x_5 = 0 \) and hence \( x_6 \neq 0 \). From (S17) we see that \( x_7 = 0 \) contradicting \( \overline{R} > 0 \).

- If \( x_2 = 0 \) or \( x_3 = 0 \) then from (S15) we have \( k_{h1}x_4 = k_{h2}x_8 = k_{2r}x_1x_4 = 0 \). It follows from (S14) that \( x_1 = 0 \) and we reach a contradiction with the item above.

- If \( x_4 = 0 \) or \( x_5 = 0 \) then from (S16) we have \( k_{3r}x_3x_6 = k_{h2}x_8 = 0 \). If \( x_4 = 0 \) using (S14) we have that \( k_{3r}x_1 = 0 \) and hence \( x_1 = 0 \), which is a contradiction. If \( x_5 = 0 \) then from (S17) we have \( k_{2r}x_2x_7 = 0 \). Since \( x_6 \neq 0 \) (because \( \overline{T} > 0 \)) and \( k_4 \neq 0 \) by hypothesis, we have \( x_7 = 0 \). As a consequence \( x_8 \neq 0 \). Hence \( k_{h2} = 0 \) and by hypothesis \( k_{h1}k_{3r}k_{4r} \neq 0 \). If \( k_{3r} \neq 0 \) we have \( x_3 = 0 \) contradicting the item above.

- If \( x_6 = 0 \) or \( x_7 = 0 \) then \( k_{h2}x_8 = k_{4r}x_5x_8 = 0 \). Since we showed that \( x_5 \neq 0 \), and \( k_{h2}k_{4r} \neq 0 \) by hypothesis, it follows that \( x_8 = 0 \). If \( x_7 = 0 \) we reach a contradiction. If \( x_6 = 0 \) then using (S16) we have \( x_4x_5 = 0 \) which contradicts the item above.

- If \( x_8 = 0 \) then \( x_7 \neq 0 \) and hence by (S17) \( x_6 = 0 \) which contradicts the item above.
A.2 Steady-state relations

Here we derive the relations shown in (S18).

(1) From the total amount $\overline{R}$, we have

$$x_7 = \overline{R} - x_8.$$  \hfill (S50)

This expression shows that $x_7$ decreases in $x_8$. For $x_7, x_8 > 0$ we require $0 < x_8 < \overline{R}$. Hence, if

we let $\alpha_1 = \overline{R}$ and $I_1 = (0, \alpha_1)$, any positive steady state must satisfy $x_8 \in I_1$.

(2) Solving (S17) for $x_6$ and subsequently solving $x_5$ using the total amount $\overline{R}$, we obtain:

$$x_6 = \frac{(k_{h2} + k_{r5})x_8}{k_{4x7} + k_{4r}x_8}, \quad x_5 = \frac{k_{4}17x7 - k_{h2}x_8}{k_{4x7} + k_{4r}x_8}.$$  \hfill (S51)

The expression of $x_5$ decreases in $x_8$ and increases in $x_7$. Since $x_7$ decreases in $x_8$, we conclude that after substituting $x_7$ with (S50), $x_5$ decreases in $x_8$. Similarly, $x_6$ increases in $x_8$.

Assume that $k_{h2} \neq 0$. For $x_5 > 0$ we require $k_{4}17x7 > k_{h2}x_8$. When $x_8 = 0$ this inequality holds. The right-hand side of the inequality increases in $x_8$ and goes to infinity. The left-hand side decreases in $x_8$ and is zero when $x_8 = \alpha_1$. Therefore, there exists a unique value of $x_8$, $\alpha_2 < \alpha_1$ at which $k_{4}17x7 = k_{h2}x_8$. Then, the inequality holds if and only if $x_8 < \alpha_2$. If $k_{h2} = 0$, then $x_5 > 0$ for all $x_8 \in I_1$ and we define $\alpha_2 = \alpha_1$.

Let $I_2 = (0, \alpha_2)$. Since $\alpha_2 \leq \alpha_1$, if $x_8 \in I_2$ then $x_5, \ldots, x_8 > 0$. Observe that if $x_8 = \alpha_2$, then $x_5 = 0$. If $x_8 = 0$, then $x_5 = \overline{R}$.

(3) Using (S16) to express $x_4$ in terms of $x_5, x_6$ and $x_8$ and using the total amount $\overline{C}$ we obtain

$$x_4 = \frac{k_{3x}C x_6 + k_{h2}x_8}{k_{3x}x_5 + k_{3r}x_6}, \quad x_3 = \frac{k_{3x}C x_5 - k_{h2}x_8}{k_{3x}x_5 + k_{3r}x_6}.$$  \hfill (S52)

The expression of $x_3$ decreases in $x_6, x_8$ and increases in $x_5$. Since $x_5$ decreases in $x_8$, and $x_6$ increases in $x_8$, we conclude that after substituting $x_5, x_6$ with (S51), $x_3$ decreases in $x_8$. Similarly, $x_4$ increases in $x_8$.

We proceed to discuss positivity following the reasoning above. Assume that $k_{h2} \neq 0$. Then for $x_3 > 0$ we require

$$k_{3x}C x_5 > k_{h2}x_8.$$  \hfill (S53)

The inequality holds at $x_8 = 0$. The right-hand side of the inequality increases in $x_8$ and goes to infinity. The left-hand side decreases in $x_8$ and is zero when $x_8 = \alpha_2$. Therefore, there exists a unique value of $x_8$, $\alpha_3 < \alpha_2$ at which $k_{3x}C x_5 = k_{h2}x_8$ and the inequality holds if and only if $x_8 < \alpha_3$. If $k_{h2} = 0$, then $x_5 > 0$ for all $x_8 \in I_2$ and $\alpha_3 = \alpha_2$.

Let $I_3 = (0, \alpha_3)$. Since $\alpha_3 \leq \alpha_2$, we have $x_3, \ldots, x_8 > 0$ if and only if $x_8 \in I_3$. Note that if $x_8 = \alpha_3$, then $x_3 = 0$. If $x_8 = 0$ then $x_3 = \overline{C}$.

(4) Using (S15) and the total amount $\overline{R}$ and obtain:

$$x_2 = \frac{k_{2x}R x_4 + k_{h1}x_4 + k_{h2}x_8}{k_{2x}x_3 + k_{2r}x_4}, \quad x_1 = \frac{k_{2x}R x_3 - k_{h1}x_4 - k_{h2}x_8}{k_{2x}x_3 + k_{2r}x_4}.$$  \hfill (S53)

The expression of $x_1$ decreases in $x_4, x_8$ and increases in $x_3$. Since $x_3$ decreases in $x_8$, and $x_4$ increases in $x_8$, we conclude that after substituting $x_3, x_4$ with (S52), $x_1$ decreases in $x_8$. Similarly, $x_2$ increases in $x_8$.  

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For $x_1 > 0$ we require
\[ k_2 \overline{P} x_3 > k_{h1} x_4 + k_{h2} x_8. \]

Recall that by hypothesis we have either $k_{h1} \neq 0$ or $k_{h2} \neq 0$, that is, the two hydrolysis rate constants cannot vanish simultaneously. If $x_8 = 0$ then the inequality is satisfied (because $x_4 = 0$). The right-hand side of the inequality increases in $x_8$ and goes to infinity. The left-hand side increases in $x_8$ and is zero when $x_8 = 0$. Therefore, there exists a unique value of $x_8$ in $I_3$, $\alpha < \alpha_3$, at which $k_2 \overline{P} x_3 = k_{h1} x_4 + k_{h2} x_8$ and the inequality holds if and only if $x_8 < \alpha$.

Let $I = (0, \alpha)$. Since $\alpha \leq \alpha_3$, we have $x_1, \ldots, x_8 > 0$ if and only if $x_8 \in I$. Note that if we let $x_8 = \alpha$, then $x_1 = 0$ and if $x_8 = 0$ then $x_1 = \overline{P}$.

The value $\alpha$ is the first positive value of $x_8$ for which
\[ k_2 \overline{P} x_3 - k_{h1} x_4 - k_{h2} x_8 = 0. \] (S54)

After substituting $x_3, x_4$ by (S52), and subsequently by (S51) and (S50), the left-hand side of this equality is a quotient of polynomials. The first positive root (in $x_8$) of the numerator is $\alpha$. The numerator is precisely the polynomial $q_2(x_8)$ given in (S19).

A.3 Signal-response curve

The entries of (S18) are derived using all steady-state equations except for (S14). From (S14) we obtain that
\[ k_s = \frac{k_{h1} x_4 + k_{h2} x_8}{x_1}. \] (S55)

The expression of $k_s$ is positive provided that $x_8 \in I$. Since $x_4$ increases in $x_8$ and $x_1$ decreases in $x_8$, we see that $k_s$ is expressed as an increasing positive function for $x_8 \in I$. When $x_8$ approaches the upper bound of the interval $I$, $\alpha$, then $x_1$ tends to zero and $x_4$ to some finite number. Hence $k_s$ grows to infinity.

The explicit form of $f$ given in (S20) is obtained from (S55) by plugging in the values of $x_1, \ldots, x_7$ obtained in (S50)-(S53) (computations are done in Mathematica).

A.4 Hyperbolic shape when phosphorelay rates are large

We prove here that if (S24) holds, then the second derivative of $\varphi$ at any value of $k_s$ is negative.

Using Mathematica, we compute the second derivative of $x_4$ with respect to $x_8$ and find that its sign equals the sign of
\[ (k_3 - k_3 r)(k_4 k_{h2} \overline{R} + k_{3 r} k_{h2} \overline{C}) + k_3 r (k_3 k_4 - k_3 r k_4 r) \overline{C} \overline{T}. \]

Consider now
\[ x_2 = \frac{k_2 \overline{P} x_3 + k_{h1} x_4 + k_{h2} x_8}{k_2 x_3 + k_{h2} x_8} = \frac{(k_2 \overline{P} + k_{h1}) x_4 + k_{h2} x_8}{(k_2 r - k_2) x_4 + k_{h2} \overline{C}}. \]

Let $\beta = (k_2 \overline{P} + k_{h1}) x_4 + k_{h2} x_8$ be the numerator of $x_2$ and $\gamma = (k_2 r - k_2) x_4 + k_{h2} \overline{C}$ be the denominator of $x_2$. Both terms are positive. We take the second derivative of $x_2$ with respect to $x_8$ and obtain:
\[ x_2'' = \frac{(\beta'' \gamma - \beta' \gamma') \gamma - 2 \gamma' (\beta' \gamma - \beta \gamma') \gamma^3}{\gamma^3} = \frac{((k_2 r \overline{P} + k_{h1}) \gamma - (k_2 r - k_2) \beta) \gamma x_4'' - 2 (k_2 r - k_2) x_4' (\beta' \gamma - \beta \gamma')}{\gamma^3}. \]
Let

\[ A = ((k_2, \overline{P} + k_1) \gamma - (k_2 - k_2) \beta) \gamma x''_4 = (k_2, (\overline{P} \gamma - \beta) + k_1 \gamma + k_2 \beta) \gamma x''_4; \]

\[ B = -2(k_2 - k_2) x'_4 (\beta' \gamma - \beta';') \]

such that \( x''_2 = (A + B)/\gamma^3 \). The denominator is positive. Therefore, the sign of \( x''_2 \) is determined by the sign of \( A + B \). Since \( x_2, x_4 \) increase in \( x_8 \) we have that \( \beta' \gamma - \beta';' > 0 \) and \( x'_4 > 0 \). Therefore, the sign of \( B \) equals the sign of \( k_2 - k_2 \).

The term \( \overline{P} \gamma - \beta = -k_2 x_4 \overline{P} + k_2 C \overline{H} - k_1 x_4 - k_{h2} x_8 \) is positive in \( I \) because it agrees with the numerator of \( x_1 \). It follows that the sign of \( A \) equals the sign of \( x''_4 \). If the signs of \( A \) and \( B \) agree, then \( x''_4 \) has a constant sign over \( I \).

Consider now the inverse of the signal-response curve:

\[ k_s = f(x_8) := \frac{k_{h1} x_4 + k_{h2} x_8}{x_1} \]

and let \( \delta = k_{h1} x_4 + k_{h2} x_8 \). The second derivative of \( f \) with respect to \( x_8 \) is

\[ f'' = \frac{x_1 (k_{h1} x'_4 x_1 - \delta x''_4) - 2 x'_4 \delta' x_1 - \delta x'_1)}{x_1^3} \]

The term \(-2 x'_4 (\delta' x_1 - \delta x'_1)\) is positive because \( x'_1 < 0 \) and \( (\delta' x_1 - \delta x'_1) > 0 \) (it is the numerator of the derivative of \( f \)). If \( (k_{h1} x''_4 x_1 - \delta x''_4) > 0 \), then the signal-response curve is hyperbolic (because the sign of the second derivative of \( \varphi \) is minus the sign of the second derivative of \( f \)). In particular, this is the case if \( x''_2 > 0 \) and \( x''_4 < 0 \). For \( x'_1 < 0 \) we require \( x''_2 > 0 \).

Therefore, if \( x''_2, x''_4 > 0 \), then the signal-response curve is hyperbolic. Using the computations above, we conclude that if

\[ k_2 - k_2, \quad \text{and} \quad (k_3 - k_{3r})(k_{h1} k_{h2} \overline{R} + k_{3r} k_{h2} C) + k_{3r} (k_{3} k_{4} - k_{3r} k_{4r}) \overline{C} \overline{T} > 0, \]

then the curve is hyperbolic. These two inequalities are in particular fulfilled if

\[ k_2 > k_2, \quad k_3 > k_{3r}, \quad k_1 > k_{4r}, \]

that is, if the forward phosphorelay rate constants are larger than their reverse counterparts.

B Proof of the claims: bifunctional case

This appendix provides a sketch of the proofs of the claims in Section 2.

B.1 Zero concentrations

We start by checking that the combinations \( k_{h2} = 0 \) and either \( k_{3r} = 0 \) or \( k_{4r} = 0 \) provide constant signal-response curves. Assume that \( k_{h2} = 0 \) and \( x_4 = 0 \) at steady state. Then by (S45), \( x_9 = 0 \). Consequently from (S41), we have \( x_1 = 0 \) and hence \( x_2 \neq 0 \) (S37). Similarly from (S38), we have that \( x_3 \neq 0 \). But then (S42) cannot hold. Therefore, if \( k_{h2} = 0, x_4 \neq 0 \) at steady state.

- \( k_{h2} = k_{3r} = 0 \): From (S43), \( x_5 = 0 \) because \( x_4 \neq 0 \) at steady state. Consequently, \( x_6 = \overline{T} \neq 0 \) and from (S44) it follows that \( x_7 = 0 \). Thus \( x_8 = \overline{R} \neq 0 \) at steady state.
• $k_{h_2} = k_{4r} = 0$: From (S44), either $x_6 = 0$ or $x_7 = 0$. If $x_6 = 0$ we have that $x_5 \neq 0$. From (S43) if follows $x_4 = 0$, which is a contradiction. Therefore $x_7 = 0$, and hence $x_8 = R \neq 0$ at steady state.

Assume now that none of the two scenarios above occur, and further that $k_{h_1} = k_{h_2} = 0$ does not occur. If $x_9 = 0$ is a solution at steady state, then by (S45) we must either have $x_1 = 0$ or $x_4 = 0$. If $x_1 = 0$ then $x_2 \neq 0$. Further from (S41) we have $k_{h_1}x_1 = k_{h_2}x_8 = 0$. Hence from (S42) we have $x_3 = 0$ and as a consequence $x_4 \neq 0$. If $k_{h_1} \neq 0$ then $k_{h_1}x_4 \neq 0$, which is a contradiction. Hence, assume that $k_{h_1} = 0$. Then using (S43) we deduce that $x_5 = 0$ and hence $x_6 \neq 0$ from the conservation law. From (S44) we obtain $x_7 = 0$ which contradicts (S40) for $R > 0$ only if $k_{h_2} \neq 0$.

Assume now that $x_4 = 0$. Then by (S38) we have $x_3 \neq 0$. Further, from (S43) we have $k_{h_2}x_8 = 0$. From (S42) we have $x_2 = 0$ and from (S37) $x_1 \neq 0$, contradicting (S41).

Therefore, if $k_{h_1} \neq 0$ or $k_{h_2} \neq 0$ then $x_9 = 0$ is not a solution at steady state. If one of the concentrations $x_1, \ldots, x_4$ is zero at steady state, then the positive term in one of the equations (S42), (S43), (S45) vanishes, implying that all the other monomials must vanish as well. For any of the equations, it would follow imply that $k_6x_9 = 0$ contradicting $x_9 \neq 0$. That is, $x_1, \ldots, x_4 \neq 0$ at steady state. If $x_5 = 0$ and $k_{3r} \neq 0$ then using (S43) and (S39), $x_3 = 0$ which is a contradiction. If $k_{3r} = 0$ then $k_{h_2} \neq 0$ (by assumption) and hence $x_8 = 0$. By (S44) we have $x_6 = 0$ or $x_7 = 0$. The latter contradicts (S40) because $x_8 = 0$. Hence $x_6 = 0$. However this contradicts (S39), because $x_5 = 0$.

Therefore, if $k_{h_2} \neq 0$ or if $k_{h_2} = 0$ but $k_{3r}k_{4r}k_{h_1} \neq 0$, then there are no zero concentrations at steady state.

### B.2 Steady-state relations

Here we derive the expressions shown in (S47). We study the concentrations at steady state that are non-zero. We assume either (1) $k_{h_2} \neq 0$ or (2) $k_{h_2} = 0$ and $k_{3r}k_{4r}k_{h_1} \neq 0$.

1. Using the total amount equation for $R$, we have
   \[
   x_7 = R - x_8, \tag{S56}
   \]
   such that $x_7$ is expressed as a decreasing function of $x_8$. We have $x_7, x_8 > 0$ if and only if $0 < x_8 < \alpha_1 := R$.

2. Using (S44) and the total amount equation for $T$ we obtain
   \[
   x_6 = \frac{x_8(k_{4r}T + k_{h_2})}{k_4x_7 + k_{4r}x_8}, \quad x_5 = \frac{k_4T x_7 - k_{h_2}x_8}{k_4x_7 + k_{4r}x_8}. \tag{S57}
   \]
   The expression for $x_5$ decreases in $x_8$ and increases in $x_7$. Since $x_7$ decreases in $x_8$, we conclude that after substituting $x_7$ with (S56), $x_5$ decreases in $x_8$. Similarly $x_6$ increases in $x_8$.

   For $x_5, x_6, x_7 > 0$, we require $k_4T(R - x_8) - k_{h_2}x_8 > 0$, that is,
   \[
   0 < x_8 < \alpha_2 := \frac{k_4TR}{k_4T + k_{h_2}} \leq \alpha_1.
   \]
   Hence, $0 < x_8 < \alpha_2$ if and only if $x_5, x_6, x_7, x_8 > 0$. 

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(3) Using (S43) and the total amount equation for \( \mathcal{C} \), we obtain:

\[
x_4 = \frac{k_{32}x_8 + k_{33}x_6(C - x_9)}{k_{32}x_5 + k_{33}x_6}, \quad x_3 = \frac{k_{33}x_5(C - x_9) - k_{32}x_8}{k_{32}x_5 + k_{33}x_6}.
\]

(S58)

\( x_4 \) is positive provided \( 0 < x_8 < \alpha_2 \) and \( 0 < x_9 < C \). \( x_3 \) is positive provided \( x_8, x_9 \) satisfy \( k_{33}x_5(C - x_9) > k_{32}x_8 \), that is,

\[
x_9 < \frac{k_{33}x_5C - k_{32}x_8}{k_{33}x_5} \leq C.
\]

The right-hand side decreases in \( x_8 \). It is zero when \( k_{33}x_5C = k_{32}x_8 \). If \( x_8 > 0 \) then \( x_5 > 0 \) and hence the value that makes the right-hand side zero satisfies \( x_9 < \alpha_2 \). Therefore, \( x_3, \ldots, x_9 \) are positive for \( x_8, x_9 \) in the set

\[
\Omega_1 := \left\{(x_8, x_9) \in \mathbb{R}_+^2 \mid x_9 < \frac{k_{33}x_5C - k_{32}x_8}{k_{33}x_5}\right\},
\]

with \( x_5 \) given by (S56) and (S57). For each value of \( 0 < x_9 < C \), let \( \beta(x_9) \) be the value of \( x_8 \) for which \( x_9 = \frac{k_{33}x_5C - k_{32}x_8}{k_{33}x_5} \), that is, the upper-bound of allowed values for \( x_8 \). Note that \( \beta(x_9) \) decreases in \( x_9 \).

The expression for \( x_4 \) in (S58) increases in \( x_8 \) and decreases in \( x_5, x_9 \). The derivative of \( x_4 \) with respect to \( x_6 \) equals

\[
\frac{k_{33}x_3}{(k_{32}x_5 + k_{33}x_6)},
\]

and hence it is positive provided \( x_3, x_5, x_6 > 0 \). Therefore, for \( (x_8, x_9) \in \Omega_1 \), \( x_4 \) increases in \( x_8 \) and decreases in \( x_9 \). Similarly, \( x_3 \) decreases in \( x_8 \) and in \( x_9 \).

(4) Using (S42) and the total amount equation for \( \mathcal{H} \) we obtain

\[
x_1 = \frac{k_{23}(\mathcal{H} - x_9) - k_6x_9 - k_{14}x_4 - k_{25}x_8}{k_{23}x_4 + k_{24}x_4}, \quad x_2 = \frac{k_{24}x_4(\mathcal{H} - x_9) + k_6x_9 + k_{14}x_4 + k_{25}x_8}{k_{23}x_4 + k_{24}x_4}.
\]

For \( (x_8, x_9) \in \Omega_1 \), \( x_1 \) is positive provided that

\[
k_{23}(\mathcal{H} - x_9) > k_{14}x_4 + k_{25}x_8 + k_6x_9.
\]

Fix a value of \( 0 < x_9 < \min(\mathcal{H}, \mathcal{C}) \). Then the left-hand side of the inequality is a decreasing function of \( x_9 \) and the right-hand side of the inequality is increasing in \( x_9 \). It follows that there exists a value \( \gamma(x_9) \) such that the inequality is fulfilled if and only if \( x_9 < \gamma(x_9) \).

If \( x_8 = \beta(x_9) \), then \( x_3 = 0 \) while the right-hand side of the inequality is positive. It follows that \( \beta(x_9) > \gamma(x_9) \). \( x_2 \) is positive if \( x_4, x_3 \) are positive and \( x_9 < H \). Therefore, \( x_1, \ldots, x_9 \) are positive provided \( x_8, x_9 \) belong to

\[
\Omega_2 := \left\{(x_8, x_9) \in \mathbb{R}_+^2 \mid x_9 < \min(\mathcal{H}, \mathcal{C}), \ k_{23}(\mathcal{H} - x_9) > k_{14}x_4 + k_{25}x_8 + k_6x_9\right\}.
\]

It can be seen that in \( \Omega_2 \), \( x_1 \) decreases in \( x_9 \) and in \( x_8 \). Similarly, \( x_2 \) increases in \( x_8 \). Further, the numerator of \( x_1 \) also decreases in \( x_9 \). It follows that the supremum of \( x_8 \) in \( \Omega_2 \) is obtained by setting \( x_9 = 0 \):

\[
k_{23}\mathcal{H} - k_{14}x_4 - k_{25}x_8 = 0,
\]

where \( x_9 = 0 \) is inserted into the expression of \( x_3, x_4 \). The solution of this equation is precisely the value \( \alpha \) obtained in the monofunctional case (see equation (S54)). Furthermore, the possible values of \( x_8 \) in \( \Omega_2 \) are in \( I = (0, \alpha) \).
(5) Using (S45) we obtain another expression for $x_1$ at steady state:

$$x_1 = \frac{(k_{5r} + k_6)x_9}{k_5x_4} = \frac{x_9}{k_yx_4},$$

where $k_y = k_5/(k_{5r} + k_6)$. This expression decreases in $x_8$. We equate the two expressions for $x_1$:

$$\frac{x_9}{k_yx_4} = \frac{k_2x_3\Pi - k_2x_3x_9 - k_6x_9 - k_h1x_4 - k_h2x_8}{k_2x_3 + k_2x_4}$$

(S59)

in order to relate $x_8$ and $x_9$. This equality does not provide a linear equation in $x_8$ nor in $x_9$, when substituting the expressions for $x_3, x_4$ in terms of $x_8, x_9$. Thus, we have to proceed in a different way from what we have done so far. For a fixed value of $x_8$ in $I$, the left-hand side of the equation increases in $x_9$ and the right-hand side decreases in $x_9$. Therefore, for a fixed value of $x_8$ in $I = (0, \alpha)$, the two sides of the equality intersect in exactly one point: $x_9 = g(x_8)$. Since the intersection point ensures that the right-hand side is positive, the intersection point satisfies by construction that $(x_8, g(x_8)) \in \Omega_2$.

We do not have an analytical description of $g$ but we have a procedure to determine $g(x_8)$ from a given $x_8$. The function $g$ is given by the Implicit Function Theorem. Let

$$G(x_8, x_9) = (k_2x_3(\Pi - x_9) - k_6x_9 - k_h1x_4 - k_h2x_8)k_yx_4 - x_9(k_2x_3 + k_2x_4) = 0.$$

Then, for every value of $x_8$, $g(x_8)$ is the first positive root of $G(x_8, x_9)$. It follows that $g$ is continuous in $I$ and differentiable. The derivative of $g$ with respect to $x_8$ is given by

$$g'(x_8) = -\frac{(\partial G/\partial x_8)(x_8, g(x_8))}{(\partial G/\partial x_9)(x_8, g(x_8))}$$

The function $g$ can be extended to $x_8 = 0$ with $g(0) = 0$.

B.3 Signal-response curve

The entries of (S47) are derived using all steady-state equations except for (S41). From (S41) we obtain that

$$k_s = f_b(x_8) = \frac{k_h1x_4 + k_h2x_8 + k_6x_9}{x_1}. \quad (S60)$$

If $x_8 \in I$, then $f_b(x_8)$ is positive. Therefore, all concentrations at steady state are positive. This function is continuous and differentiable. When $x_8$ approaches the upper bound of the interval $I$, $\alpha$, then $x_1$ tends to zero, $x_4$ to some finite number and $x_9$ to zero. Hence $k_s$ grows to infinity (provided $k_{h1}$ or $k_{h2}$ are non-zero, see below for the case $k_{h1} = k_{h2} = 0$). It follows that the image of $f_b$ is $(0, +\infty)$ which guarantees the existence of at least one steady state. The function $f_b$ can be differentiably extended at zero such that $f_b(0) = 0$.

Using the Chemical Reaction Network toolbox [2], we know that the system does not admit multiple positive steady states. By continuity, it follows that $f_b$ must be monotone, that is, an increasing function. By the Inverse Function Theorem, there exists a continuous and differentiable function in $(0, +\infty)$,

$$x_8 = \varphi_b(k_s)$$

defined by $\varphi_b(k_s) = x_8$ if and only if $k_s = f_b(x_8)$.

If we increase $k_y$ while keeping $k_6$ fixed, the right-hand side of (S59) increases. It follows that the value $g(x_8)$ increases. Consequently, $x_4$ increases and $x_1$ decreases, which implies that $k_s$ must increase as well. We conclude that as $k_y$ increases the graphs of $f_b$ pile on top of each other and hence the graphs of $\varphi_b$ lie below each other.
$k_{h_1} = k_{h_2} = 0$: First of all, an easy check shows that $(0, \bar{H}, 0, \bar{C}, 0, \bar{T}, 0, \bar{R}, 0)$ is a steady state for all values of $k_s$. But for $k_s$ small enough, a second positive steady state exists as well.

In this case we have $\alpha = \bar{R}$ and

$$k_s = f_b(x_8) = \frac{k_6 x_9}{x_1}.$$  

Hence both the numerator and denominator of $f_b$ tend to zero as $x_8$ tend to $\alpha = \bar{R}$. By plugging the expression of $x_1$ into $f_b$, we have:

$$f_b(x_8) = \frac{k_6 k_{3r} k_{4r} k_y (\bar{C} - x_9) x_8}{k_3 k_4 (\bar{R} - x_8) - k_{3r} k_{4r} x_8}.$$  

We deduce that when $x_8 = \bar{R}$, then $x_9 = 0$ and $f_b(x_8) = k_6 k_y \bar{C}$. It follows that $k_s = f_b(x_8)$ does not tend to infinity as $x_8$ approaches the upper bound of $I, \bar{R}$. In this case, the signal-response curve is defined by $f_b^{-1}$ for $k_s \in [0, k_6 k_y \bar{C}]$ and is constant at $\bar{R}$ for $k_s > k_6 k_y \bar{C}$.

References
