Optimization of information transmission through a noisy biochemical pathway explains the choice of reaction rates SUPPLEMENTARY

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Estimated Parameters

1.
$$k_1 = 1078.4 \text{ nM}^{-1} \text{ min}^{-1}$$

2.
$$S_{basal} = 86.68 \text{ min}^{-1}$$

3.
$$P_{basal} = 2036.6 \text{ min}^{-1}$$

4.
$$j_x = 0.7653 \text{ nM min}^{-1}$$

5.
$$N_{G_x} = 0.3344$$

6.
$$d_x = 0.0304 \,\mathrm{min}^{-1}$$

7.
$$K_{GFP} = 3.7001 \text{ nM}$$

8.
$$j_{GFP} = 0.0594 \text{ nM min}^{-1}$$

9.
$$d_{GFP} = 0.0363 \text{ min}^{-1}$$

10.
$$N_{GGFP} = 0.2281$$

11.
$$k_2 = 2652.9 \text{ nM}^{-1} \text{ min}^{-1}$$

12.
$$j_P = 0.0174 \text{ nM min}^{-1}$$

13.
$$N_{G_P} = 2.6023$$

14.
$$K_2 = 0.2081 \text{ nM}$$

15.
$$d_P = 0.0174 \text{ min}^{-1}$$

The mean and variance of the one stage phosphorylation cascade

In the schematic diagram of a one-stage phosphoryltion cycle shown below, S corresponds to the signaling molecule. $X_1(X_2)$ represents the dephosphorylated (phosphorylated) form of the signaling protein X.

$$X_1 \stackrel{\alpha_0 + \alpha_1 S}{\longleftarrow} X_2$$

The dynamics of the active protein concentration is given by

$$\frac{dx_2}{dt} = (\alpha_0 + \alpha_1 s)X_T - \{(\alpha_0 + \alpha_1 s) + \beta\}x_2 \tag{1}$$

The steady state value of active $X(x_2)$ is given by

$$x_{2s} = \frac{(\alpha_0 + \alpha_1 s)X_T}{(\alpha_0 + \alpha_1 s) + \beta} \tag{2}$$

The linearized equation around a steady state with adding a Gaussian white noise term

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$$\frac{d\Delta x_2}{dt} = -\left\{ (\alpha_0 + \alpha_1 s) + \beta \right\} \Delta x_2 + \xi_1(t) \tag{3}$$

$$\langle \xi_1 \left(t_1 \right) \rangle = 0 \tag{4}$$

$$\langle \xi_1(t_1)\xi_1(t_2)\rangle = 2\beta x_{2s}\delta(t_1 - t_2) \tag{5}$$

The variance in the active protein concentration

$$\sigma_x^2 = \left\langle \Delta x_2^2 \right\rangle = \frac{(\alpha_0 + \alpha_1 s)\beta X_T}{(\alpha_0 + \alpha_1 s + \beta)^2} \tag{6}$$

Functional Relationship of Fisher Information with amplification and noise

The probability distributions are assumed to be normal at each value of the input

$$p(\mathbf{x} \mid \boldsymbol{\theta}) = \frac{1}{\sqrt{2\pi\sigma(\boldsymbol{\theta})^2}} e^{-\frac{1}{2} \left(\frac{x-\mu(\boldsymbol{\theta})}{\sigma(\boldsymbol{\theta})}\right)^2}$$
(7)

$$\frac{-\partial}{\partial \theta} \log[p(\mathbf{x} \mid \theta)] = \frac{(x - \mu)}{\sigma^2} \frac{\partial \mu}{\partial \theta} + \frac{1}{\sigma} \left(\frac{(x - \mu)^2}{\sigma^2} - 1 \right) \frac{\partial \sigma}{\partial \theta}$$
(8)

$$\left(\frac{\partial}{\partial \theta} \log[p(\mathbf{x} \mid \theta)]\right)^{2} = \frac{(x-\mu)^{2}}{\sigma^{4}} \left(\frac{\partial \mu}{\partial \theta}\right)^{2} + \frac{1}{\sigma^{2}} \left(\frac{(x-\mu)^{4}}{\sigma^{4}} + 1 - 2\frac{(x-\mu)^{2}}{\sigma^{2}}\right) \left(\frac{\partial \sigma}{\partial \theta}\right)^{2} + \frac{1}{\sigma^{3}} \left(\frac{(x-\mu)^{3}}{\sigma^{2}} - (x-\mu)\right) \times \left(\frac{\partial \mu}{\partial \theta}\right) \left(\frac{\partial \sigma}{\partial \theta}\right) \times \left(\frac{\partial \mu}{\partial \theta}\right) \left(\frac{\partial \sigma}{\partial \theta}\right) \tag{9}$$

Taking average over x while the cross term vanishes due to zero moments of the odd exponents for normal distribution and $\langle (x-\mu)^4 \rangle = 3\sigma^4; \langle (x-\mu)^2 \rangle = \sigma^2$, we obtain

$$\left\langle \left(\frac{\partial}{\partial \theta} \log[p(\mathbf{x} \mid \theta)] \right)^2 \right| = \frac{1}{\sigma^2} \left[\left(\frac{\partial \mu}{\partial \theta} \right)^2 + 2 \left(\frac{\partial \sigma}{\partial \theta} \right)^2 \right]$$
 (10)

now,

$$\frac{1}{\sigma^2} \left(\frac{\partial \sigma}{\partial \theta} \right)^2 = \left(\frac{1}{\sigma} \frac{\partial \sigma}{\partial \theta} \right)^2 \tag{11}$$

Thus, the first term in Equation S10 would contain the total number of molecule (X_T) while the second term is independent of X_T . Hence, the Fisher information when X_T is large can be approximated as,

$$F(\theta) = \left\langle \left(\frac{\partial}{\partial \theta} \log[p(\mathbf{x} \mid \theta)] \right)^2 \middle| \approx \frac{1}{\sigma^2} \left(\frac{\partial \mu}{\partial \theta} \right)^2$$
 (12)

Figures

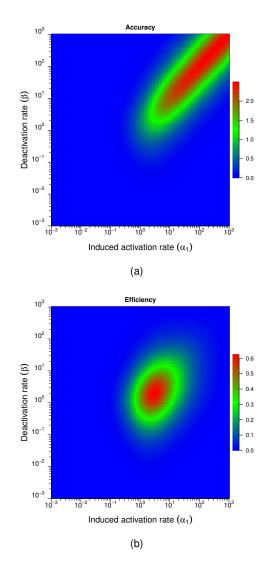


Figure S1. The optimization of information transmission from the analytical calculation in 2-dimension.(a) The heatmap displays the accuracy as function of the induced activation rate (α_1) and the deactivation rate (β) (b) The heatmap displays the efficiency as function of the induced activation rate $(alpha_1)$ and deactivation rate (beta)

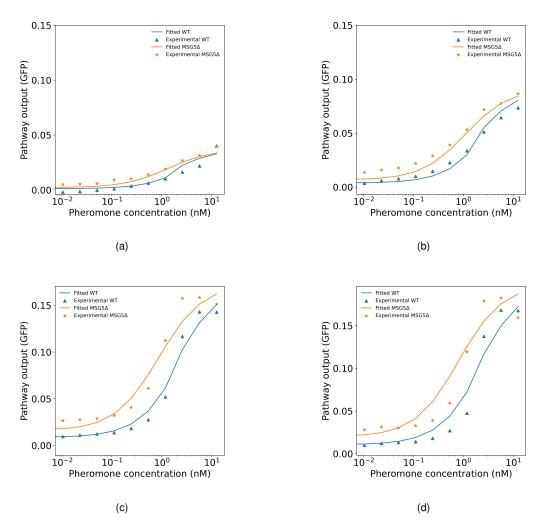
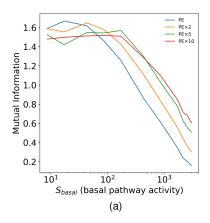
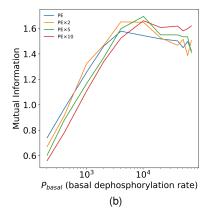


Figure S2. The fitted curves along with the experimentally measured GFP output values at the various time lapse indices shown for the WT and the MSG5 Δ strain as indicated (a) First time-lapse index (b) Second time-lapse index (c) Fourth time-lapse index (d) Fifth time-lapse index





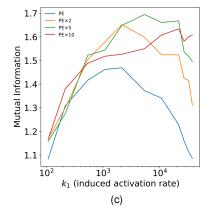


Figure S3. (a) Mutual information vs S_{basal} at four different values of the induced activation rate (k_1) as indicated in the legend while keeping other parameter values fixed at estimated values. PE corresponds to the estimated parameter value of k_1 .(b) Mutual information vs P_{basal} at four different values of the induced activation rate (k_1) as indicated in the legend while keeping other parameter values fixed at estimated values. PE corresponds to the estimated parameter value of k_1 . (c) Mutual information vs k_1 at four different values of the induced activation rate (P_{basal}) as indicated in the legend while keeping other parameter values fixed at estimated values. PE corresponds to the estimated parameter value of P_{basal} .

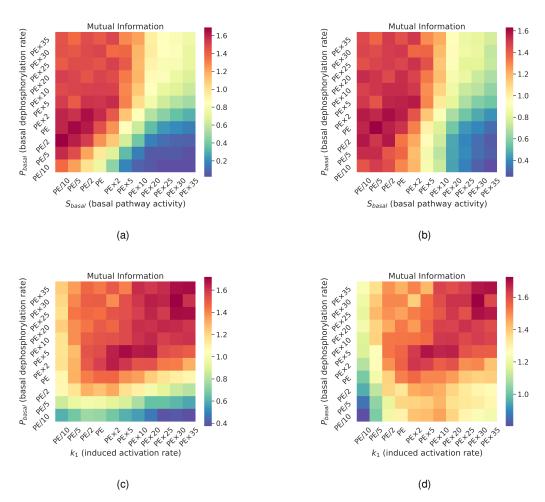


Figure S4. (a) Heatmap shows the mutual information as function of S_{basal} and P_{basal} . The PE on x-axis corresponds to estimated value of the induced activation rate (S_{basal}) while The PE on y-axis corresponds to estimated value of the basal deactivation rate (P_{basal}) . (b) Heatmap shows the accuracy as function of S_{basal} and P_{basal} for WT in presence of the induced feedback loop. The PE on x-axis corresponds to estimated value of the basal pathway activity (S_{basal}) while The PE on y-axis corresponds to estimated value of the basal deactivation rate (P_{basal}) . (c) Heatmap shows the mutual information as function of k_1 and k_2 . The PE on x-axis corresponds to estimated value of the induced activation rate (k_1) while The PE on y-axis corresponds to estimated value of the basal deactivation rate (P_{basal}) . (d) Heatmap shows the accuracy as function of k_1 and k_2 and k_3 for WT in presence of the induced feedback loop. The PE on x-axis corresponds to estimated value of the basal pathway activity (k_1) while The PE on y-axis corresponds to estimated value of the basal deactivation rate (P_{basal}) .