

Energy consumption and cooperation for optimal sensing: Supplementary Information

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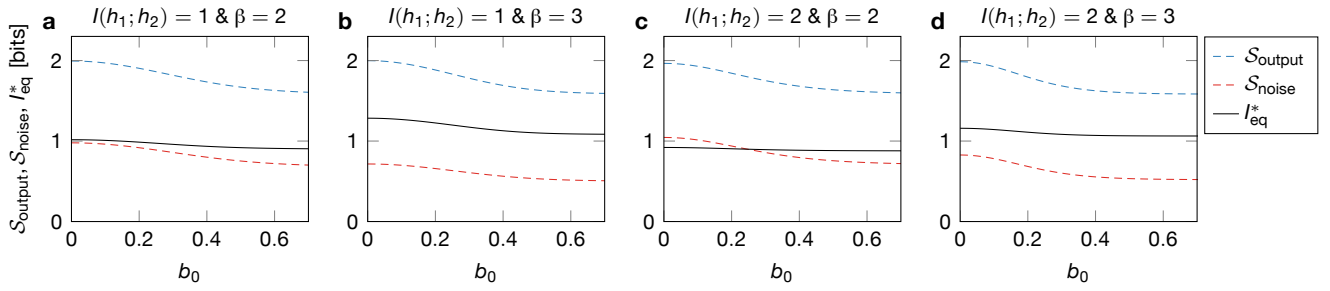
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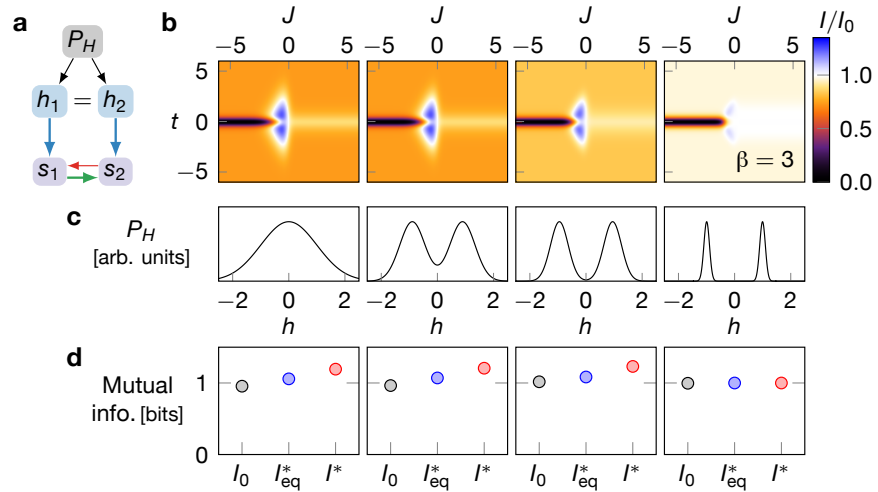
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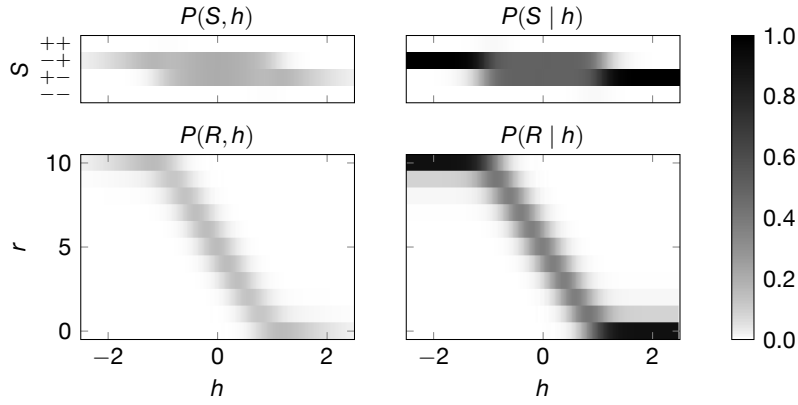
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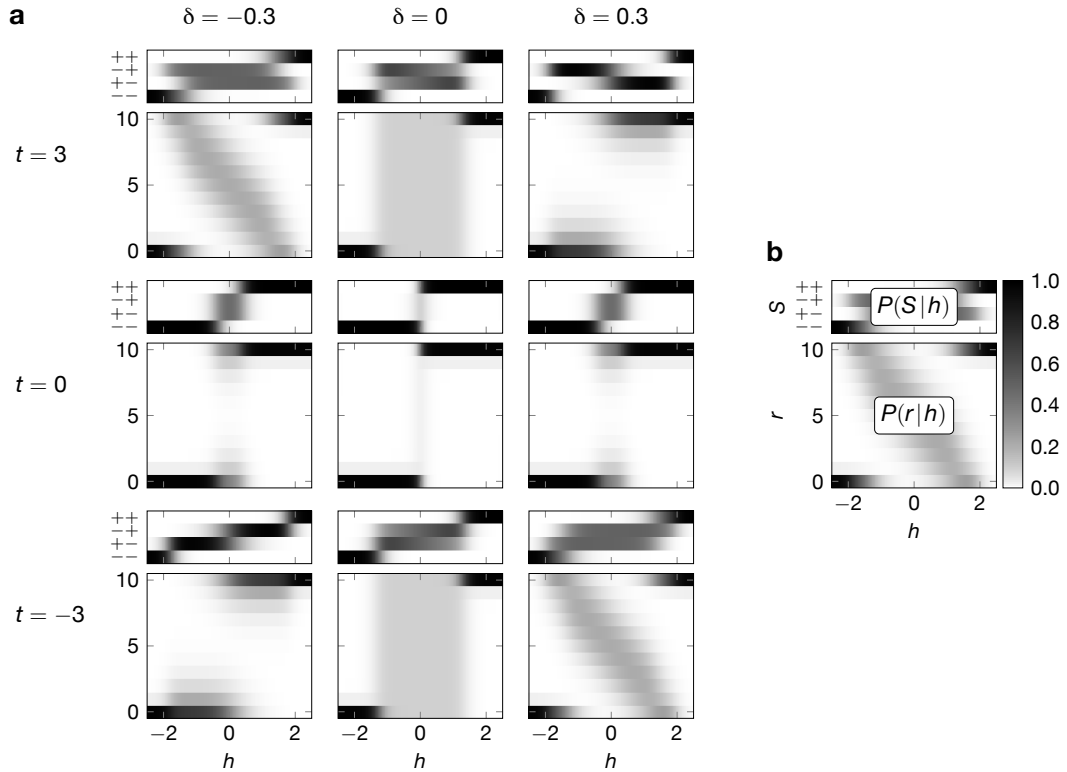
Supplementary Figure 1. Intrinsic (signal-independent) sensor biases *decrease* mutual information in optimal equilibrium sensing. We subject two sensors to bivariate Gaussian signals [Eq. (10)] and a signal-independent bias b_0 such that the net bias becomes $(h_1 + b_0, h_2 - b_0)$. We then optimise the sensor coupling J to obtain maximum mutual information and the corresponding noise and output entropies (a-d). We depict the effects of intrinsic biases for four combinations of signal redundancy $I(h_1; h_2)$ and sensor reliability β (see panel labels), which are representative of the cases where the optimal sensing strategy is equilibrium (a,b) and nonequilibrium (c,d) (see also, Fig. 2b).



Supplementary Figure 2. For highly correlated signals, the nonequilibrium improvement in the low-noise limit does not rely on the specific Gaussian prior. (a) We assume that two sensors are driven by the same bias field $h \equiv h_1 = h_2$. (b) The mutual information at sensor reliability of $\beta = 3$ for four different prior distributions with a zero mean and unit variance, shown in c. For all priors considered, the optimal nonequilibrium drive t^* is clearly finite, suggesting that the nonequilibrium enhancement is robust for most continuous priors. The dependence of the mutual information on J and t remains qualitatively unchanged. This means the mechanism behind the nonequilibrium enhancement is likely to be identical to the one for a Gaussian prior, as described in the main text. (d) The maximum mutual information — I_0 , I_{eq}^* and I^* — for noninteracting, equilibrium and nonequilibrium sensors, respectively. The nonequilibrium enhancement is visible in all cases except for the most binary-like one (far right).



Supplementary Figure 3. The ladder structure in Fig. 5e corresponds to the anticorrelated sensor states $+-$ and $-+$. The joint (left) and conditional (right) probability distributions of the sensor states (top) and the readout population (bottom). Here we use, as in Fig. 5e, $J=-2$, $t=7$, $\delta=-0.6$, $\Delta=1$ and $r_0=10$.



Supplementary Figure 4. (a) Conditional probabilities of sensor state and readout given a perfectly correlated signal (see, (b) for legend) at $J=-2$ and $\beta=4$ for $t=3, 0, -3$ (top-to-bottom) and $\delta=-0.3, 0, 0.3$ (left-to-right). A strong signal $|h| \gg 1$ favours the correlated sensor states ($++$ and $--$) and the extreme readout states ($r=r_0$ and $r=0$), a finite nonequilibrium drive $t \neq 0$ lifts the degeneracy between the sensor states $-+$ and $+-$ (cf. Fig. 4), and $\delta \neq 0$ differentiates the intermediate readout states ($0 < r < r_0$). The ladder structure in Fig. 5e requires both t and δ to be non-zero. Here we use $\Delta=1$ and $r_0=10$.