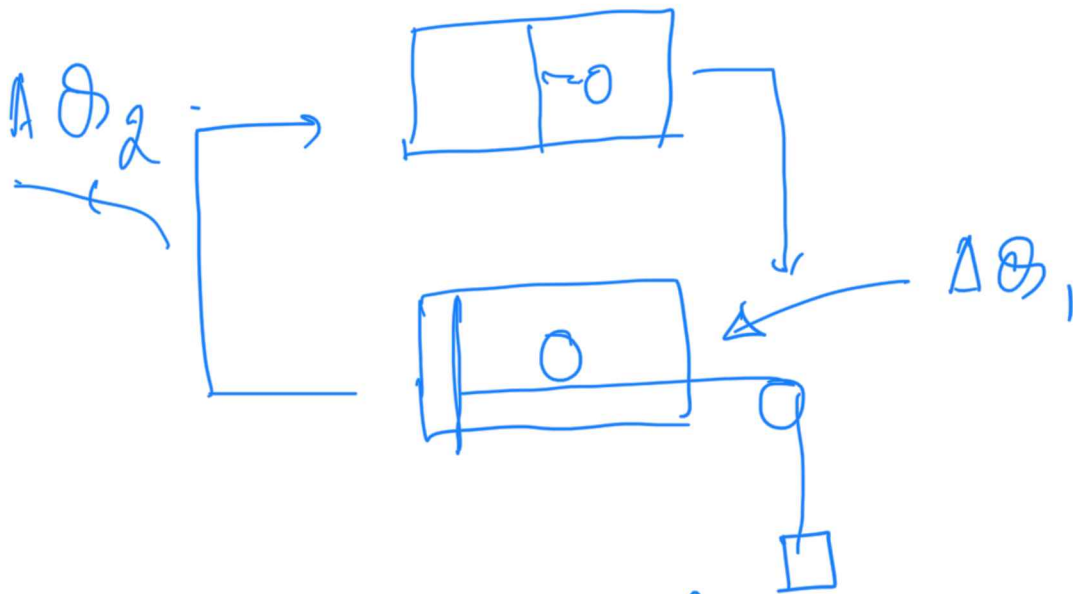


Information and thermodynamics

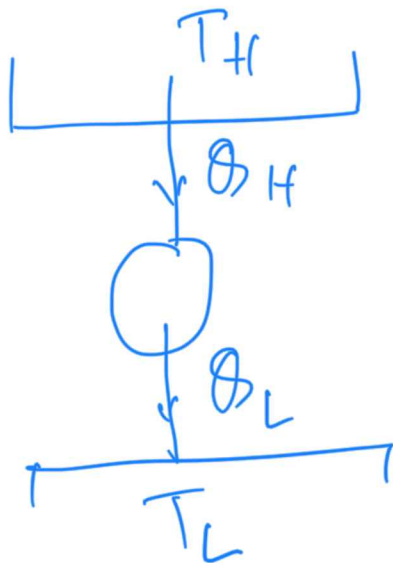
Maxwell's demon

Szilard engine



$$\Delta\Theta_1 = \Delta W = kT \log 2.$$

Carnot engine and Clausius inequality



$$\frac{\Theta_H}{\Theta_L} = \frac{T_H}{T_L}$$

$$\frac{\Theta_H}{T_H} - \frac{\Theta_L}{T_L} = 0$$

$$\oint \frac{dQ}{T} = \frac{Q_H}{T_H} - \frac{Q_C}{T_C} = 0$$

$$\Rightarrow \oint \frac{dQ}{T} = \oint ds \quad \text{S is a state variable.}$$

For reversible process η_R and
irreversible process η_{IR}

$$\eta_R > \eta_{IR}$$

$$\Rightarrow 1 - \frac{Q_H}{Q_L} > 1 - \frac{Q_H^{IR}}{Q_L^{IR}}$$

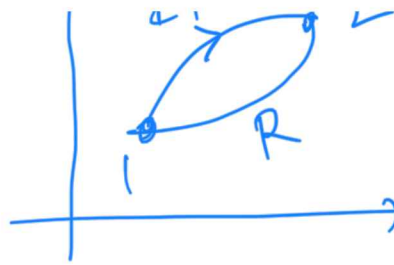
$$\frac{Q_H}{Q_L} > \frac{Q_H^{IR}}{Q_L^{IR}}$$

$$\frac{T_H}{T_L} > \frac{Q_H^{IR}}{Q_L^{IR}} \quad \left(\frac{Q_H}{T_H} - \frac{Q_L}{T_L} \right) < 0$$

$$\oint \frac{dQ}{T} < 0$$

For both cases $\oint \frac{dQ}{T} \leq 0$

Let us consider an cycle process
a, PR a



$$\oint \frac{dQ}{T} \leq 0 \quad \int_1^2 \frac{dQ}{T} + \int_2^1 \frac{dQ}{T} \leq 0$$

$$\int_1^2 \frac{dQ}{T} + S_1 - S_2 \leq 0$$

$$S_2 - S_1 \geq \int_1^2 \frac{dQ}{T}$$

$$\Delta S \geq \int_1^2 \frac{dQ}{T}$$

$$\Delta S_{\text{total}} = \int_1^2 \frac{dQ}{T} + \Delta S_g$$

$$\Delta S = \Delta S_I + \Delta S_g$$

For standard case

$$\Delta S = \left(\frac{Q_1 - Q_2}{T} \right) + \Delta S_g$$

For thermal production :

Entropy production

$$\Delta S = \Delta S_I + \Delta S_g$$

$$\frac{d}{dt} \Delta S : \frac{d}{dt} \Delta S_I + \frac{d}{dt} \Delta S_g.$$

At steady state

$$\frac{d}{dt} \Delta S_I + \frac{d}{dt} \Delta S_g = 0$$

For a stochastic process

$$\frac{d p(n, t)}{dt} = \sum_{n'} \left(W_{nn'} p(n', t) - W_{n'n} p(n, t) \right)$$

$$S = - \sum_n p(n, t) \log p(n, t)$$

$$\frac{dS}{dt} = - \sum \frac{dp}{dt} \log p(n, t)$$

$$= - \sum_n \sum_{n'} \left(w_{nn'} p(n', t) - w_{n'n} p(n, t) \right) \log p(n, t)$$

$$= - \sum_n \sum_{n'} \left(w_{n'n} p(n, t) - w_{nn'} p(n', t) \right) \log p(n', t)$$

$$\frac{ds}{dt} = \sum \frac{w_{nn'} p(n', t) - w_{n'n} p(n, t)}{\log p(n', t)}$$

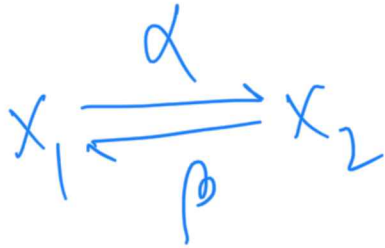
$$\frac{ds}{dt} = \frac{1}{2} \sum_n \sum_{n'} \left(w_{nn'} p(n', t) - w_{n'n} p(n, t) \right) \log \frac{p_{n'}}{p_n}$$

$$= \frac{1}{2} \sum_{nn'} \left(w_{nn'} p(n', t) - w_{n'n} p(n, t) \right) \log \frac{w_{nn'} p(n', t)}{w_{n'n} p(n, t)}$$

$$- \frac{1}{2} \sum_{nn'} \left(w_{nn'} p(n', t) - w_{n'n} p(n, t) \right) \log \frac{w_{nn'}}{w_{n'n}}$$

At steady state flux.

$$\frac{dS_g}{dt} = \frac{1}{2} \left(\omega_{nn'} p(n') - \omega_{nn'} p(n) \right) \log \frac{\omega_{nn'}}{\omega_{n'n}}$$



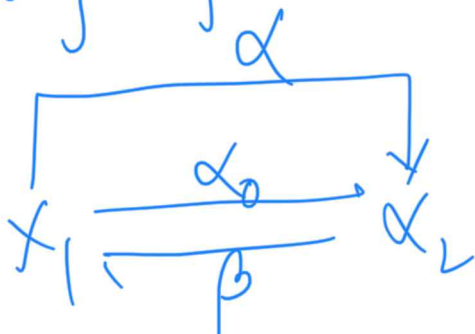
$$\frac{dS_g}{dt} = \frac{1}{2} (\alpha p_1 - \beta p_2) \log \left(\frac{\alpha}{\beta} \right)$$

$$p_1 = \frac{\beta}{\alpha + \beta} \quad p_2 = \frac{\alpha}{\alpha + \beta}$$

$$\langle n_1 \rangle = \frac{\beta}{\alpha + \beta} N \quad p_1 = \frac{\langle n_1 \rangle}{N}$$

$$\frac{dS_g}{dt} = \frac{1}{2} (\alpha \langle x_1 \rangle - \beta \langle x_2 \rangle) \log \frac{\alpha}{\beta}$$

Breaking of detailed balance



$$\frac{dS_g}{dt} = (\alpha x_1 - \beta x_2) \log \frac{\alpha}{\beta} + (\alpha_0 x_1 - \beta_0 x_2) \log \frac{\alpha_0}{\beta_0}$$

$$\frac{dx_2}{dt} = (\alpha + \alpha_0) x_1 - (\beta + \beta_0) x_2$$

$$\langle x_2 \rangle = \frac{\alpha + \alpha_0}{\alpha + \alpha_0 + \beta + \beta_0} x_T$$

$$\langle x_1 \rangle = \frac{\beta + \beta_0}{\alpha + \alpha_0 + \beta + \beta_0} x_T$$

$$\alpha x_1 - \beta x_2 = \left(\frac{\alpha(\beta + \beta_0)}{\alpha + \alpha_0 + \beta + \beta_0} - \frac{\beta(\alpha + \alpha_0)}{\alpha + \alpha_0 + \beta + \beta_0} \right) x_T$$

$$= \frac{\alpha\beta_0 - \alpha_0\beta}{\alpha + \alpha_0 + \beta + \beta_0}$$

$$\frac{dS_g}{dt} = x_T \frac{\alpha\beta_0 - \alpha_0\beta}{\alpha + \alpha_0 + \beta + \beta_0} \log \left(\frac{\alpha\beta_0}{\alpha_0\beta} \right)$$

$$\alpha + \alpha_0 + \beta + \beta_0$$

$$\approx \frac{x_T \beta \alpha}{\alpha + \alpha_0} \log\left(\frac{\alpha}{\alpha_0}\right)$$

The Fisher information

$$p(x_2|s) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_2 - \langle x_2 \rangle)^2}{2\sigma^2}}$$

$$\sigma^2 = \frac{\alpha' \beta' s}{(\alpha' s + \beta')^2} x_T$$

$$\alpha' = \alpha + \alpha_0$$

$$\beta' = \beta + \beta_0$$

$$\langle x_2 \rangle = \frac{\alpha' s}{\alpha' s + \beta'} x_T \frac{1}{\Omega}$$

For large Ω

$$F(s) = \frac{1}{\sigma^2} \left(\frac{2x_2}{2s} \right)^2$$

$$\frac{2x_2}{2s} = \frac{\alpha'}{\alpha' s + \beta'} - \frac{\alpha' s}{(\alpha' s + \beta')^2} \cdot \alpha' x_T$$

$$= \frac{\alpha'}{\alpha' s + \beta'} \left[\alpha' s + \beta' - \alpha' s \right]$$

$$(\alpha s + \beta)$$

$$= \frac{\alpha' \beta'}{(\alpha' s + \beta')^2} x_T$$

$$F(s) = \frac{(\alpha' s + \beta')^2 \Omega}{\alpha' \beta' s x_T} \frac{(\alpha' \beta')^2 x_T^2}{(\alpha' s + \beta')^4}$$

$$= \frac{\alpha' \beta'}{s (\alpha' s + \beta')^2} x_T \Omega$$

$$pr(s) = \frac{\alpha' \beta' s}{(\alpha' s + \beta')^2} x_T \Omega$$

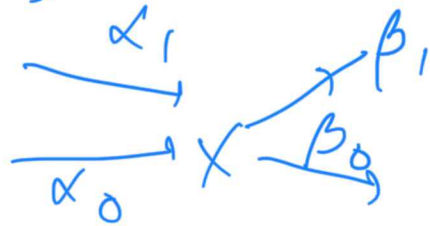
$$= \frac{\beta'}{\alpha'} \frac{s}{(s + \beta'/\alpha')^2} x_T \Omega$$

For the poisson process

$$\frac{\alpha_0}{\alpha_0} \times \frac{\beta_0}{\beta_0}$$

.....

Breaking detail balance



$$\begin{aligned} \frac{d p(n, t)}{d t} = & \left(\alpha_1 p(n-1) - \beta p(n) \right) \\ & - \left(\alpha_1 p(n) - \beta p(n+1) \right) \\ & + \left(\alpha_0 p(n-1) - \beta p(n) \right) \\ & - \left(\alpha_0 p(n) - \beta p(n) \right) \end{aligned}$$

$$\begin{aligned} \frac{d s}{d t} = & \sum_n \left(\alpha_1 p(n) - \beta(n) p(n+1) \right) \log \frac{\alpha_1}{\beta(n+1)} \\ & - \sum_n \alpha_0 p(n) - \beta_0(n+1) p(n+1) \log \frac{\alpha_1}{\beta(n+1)} \end{aligned}$$

$$p(n) = \frac{\mu^n}{n!} e^{-\mu}$$

$$\mu = \frac{\alpha_0 + \alpha_1}{\beta_0 + \beta_1}$$

$$\frac{ds}{dt} = \sum_n J_n \log \frac{\alpha_1}{\beta(n+1)}$$

$$J_n = (\alpha_1 p(n) - \beta_1(n+1) p(n+1))$$

$$= \left(\alpha_1 \frac{\mu^n}{n!} - \beta_1(n+1) \frac{\mu^{n+1}}{(n+1)!} \right) e^{-\mu}$$

$$= \frac{\mu^n}{n!} (\alpha_1 - \beta_1 \mu) e^{-\mu}$$

$$\frac{ds}{dt} = \sum \frac{\mu^n}{n!} e^{-\mu} (\alpha_1 - \beta_1 \mu) e^{-\mu} \log \frac{\alpha_1}{\beta(n+1)}$$

$$\hookrightarrow \mu^n - \mu^{n+1} \dots - \mu^{n+1} \dots \alpha_0$$

$$-\sum_{n=1}^{\infty} \frac{\mu^n}{n!} e^{-\mu} (\alpha_1 - \beta_1 \mu) e^{\log \frac{\alpha_1 \beta_0}{\alpha_0 \beta_1}}$$

$$\alpha_1 - \beta_1 \mu = \frac{\alpha_1 + \alpha_0}{\beta_1 + \beta_0}$$

$$= \sum_n \frac{\mu^n}{n!} e^{-\mu} (\alpha_1 - \beta_1 \mu) \log \frac{\alpha_1 \beta_0}{\alpha_0 \beta_1}$$

$$= (\alpha_1 \beta_0 - \beta_1 \alpha_0) \log \frac{\alpha_1 \beta_0}{\alpha_0 \beta_1}$$

$$\approx \beta \alpha_1 \log \left(\frac{\alpha_1}{\alpha_0} \right) \quad \beta \approx \beta_0 \quad \alpha_1 \gg \alpha_0$$

Fisher information