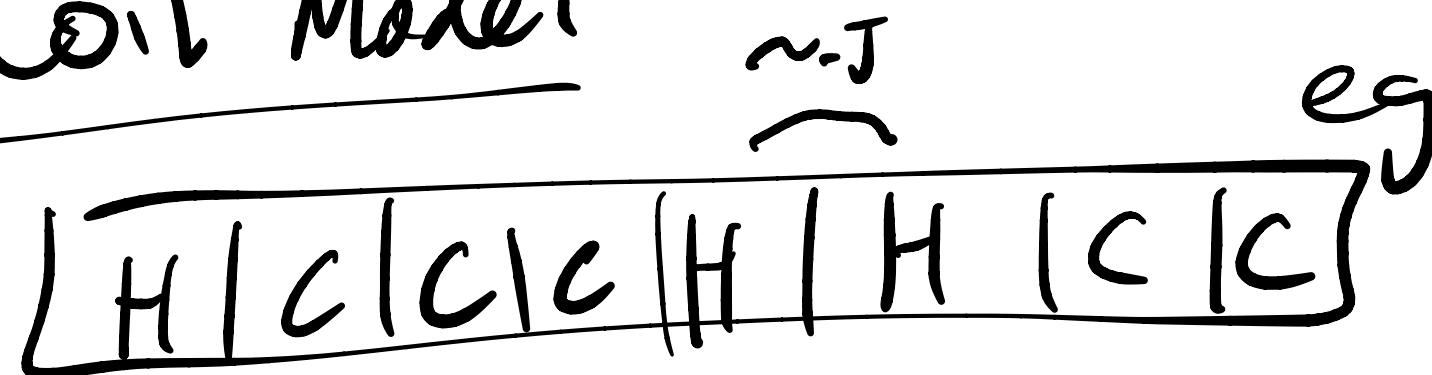


# Helix Coil Model



$C \sim 0$     $H \sim 1$

$\sum$  Energy 0    $\sum$  energy  $-E$

$$N=7$$



$$\sum \text{ weight}$$
$$\varepsilon^2 k^3$$

vs



$$k^3$$

Can "solve"  $\leftarrow$  exact formula for  $Z(N, k, \varepsilon)$

"Zipper" approximation , all H's are next  
 $\gamma \gg 1$  to each other

$$N=7, \quad n=4$$

$$-3J - 4E \quad \text{all } \gamma^3 k^4$$



only 4 cfgs vs  $\binom{7}{4}$

$$Z = 1 + \sum_{n_H=1}^N (N-n_H+1) K^{n_H} \gamma^{(n_H-1)}$$

$$Z = 1 + \sum_{n_H=1}^N (N-n_H+1) k^{n_H} z^{(n_H-1)}$$

$$= 1 + \frac{1}{z} \sum_{n_H=1}^N (N-n_H+1) [kz]^{n_H}$$

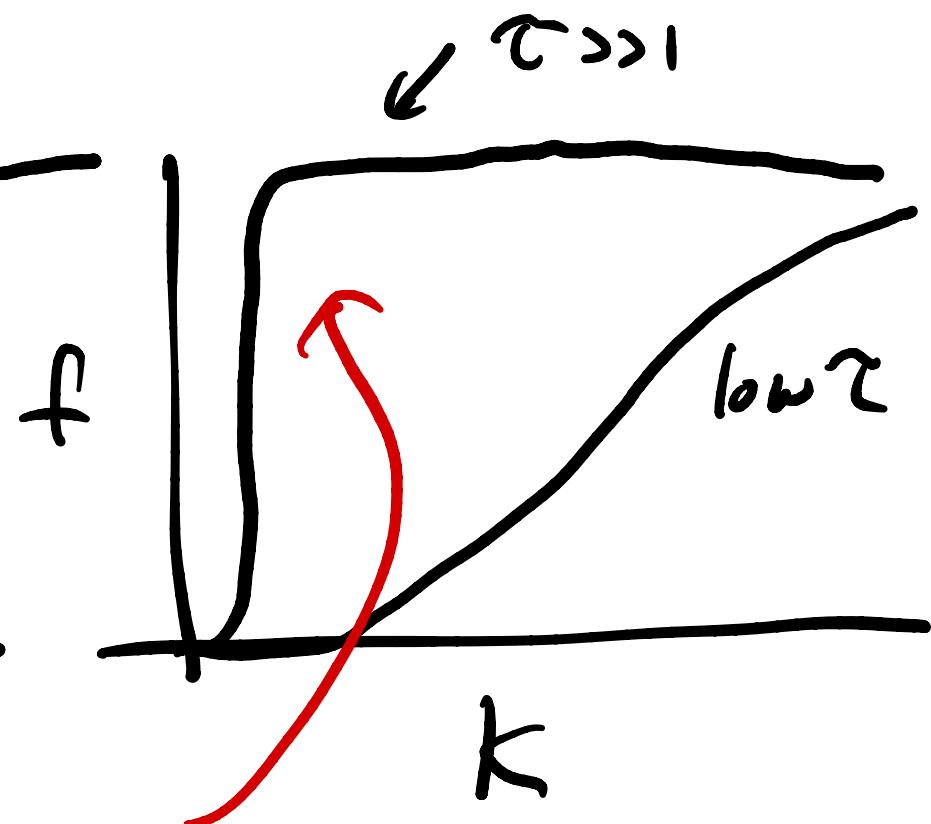
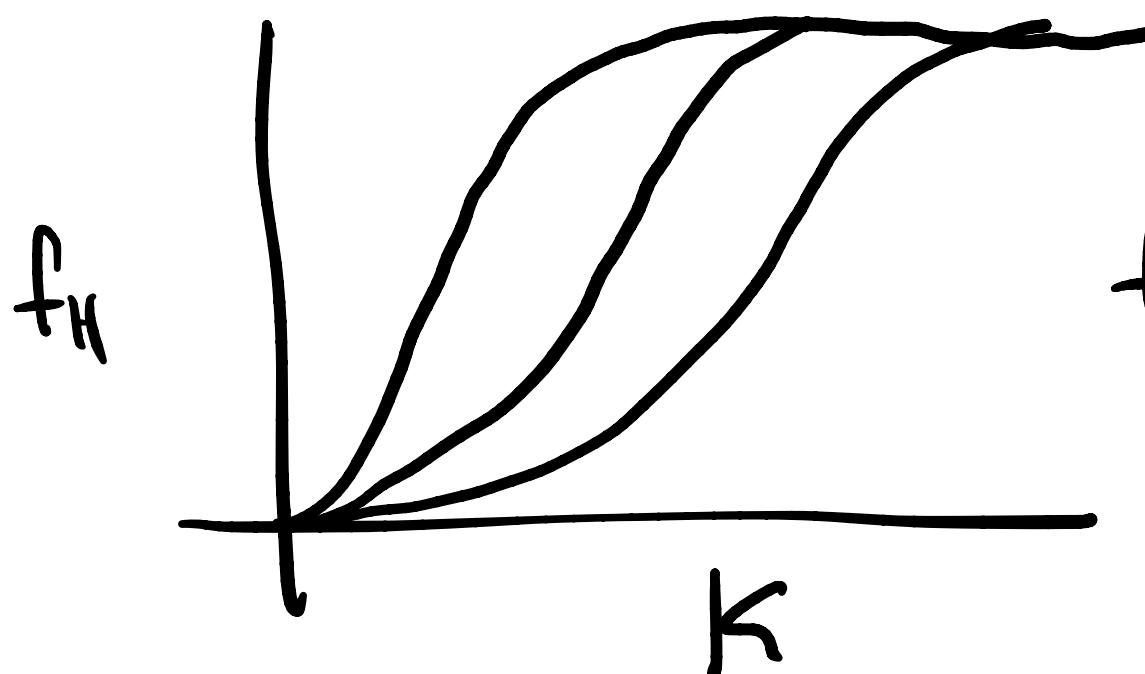
$$= 1 + \frac{1}{z} \left[ \underbrace{\sum_{n_H=1}^N (N+1)}_{\text{const}} [kz]^{n_H} + \sum_{n_H=1}^N n_H [kz]^{n_H} \right]$$

$$\sum_{n=1}^N x^n = \frac{(x^{N-1})x}{x-1}$$

↑  
derivative

$$z = 1 + \kappa \left[ \frac{(k^2)^{N+1} - N(k^2) - (k^2 + N)}{(k^2 - 1)^2} \right]$$

$$f_H = \frac{\kappa}{N} \frac{\partial h z}{\partial k}$$



"cooperative"

flow get  $Z$  exactly:

"transfer matrices"

Weight of stick in column for  
residue  $i$  following row  $i-1$

$$\omega = \begin{pmatrix} k^2 & 1 \\ k & 1 \end{pmatrix}$$

$$\omega^2 = \begin{pmatrix} k^2c^2 + k & k\bar{c} + 1 \\ k^2\bar{c} + k & k + 1 \end{pmatrix}$$



$$Z = \sum \text{bottom row of } \omega^N$$

$$Z = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \omega^N \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

1 pick  $\downarrow$  add columns  
bottom row

$\rightarrow$  what if  $N \rightarrow \infty$

$$\omega = U D U^T$$

$$Z = \begin{pmatrix} 0 \\ 1 \end{pmatrix}^T U D^N U^T \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

eigenvalues  
 $\downarrow$

$$D = \begin{pmatrix} a & 0 \\ 0 & b \end{pmatrix}$$

$$D^N = \begin{pmatrix} a^N & 0 \\ 0 & b^N \end{pmatrix}$$

Ising model

$$\begin{aligned}Z &= \text{Tr} [ \omega^N ] \\&= \lambda_1^N + \lambda_2^N \\&= \lambda_1^N \left( 1 + \left( \frac{\lambda_2}{\lambda_1} \right)^N \right) \\&\approx \lambda_1^N\end{aligned}$$



Ligand Binding



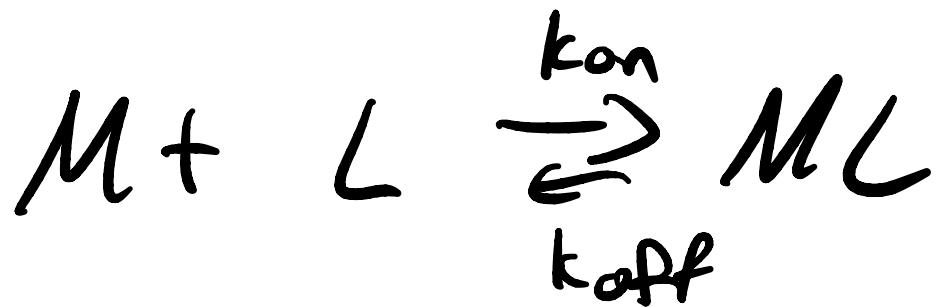
$$K_b \downarrow \quad K_d$$

low  $K_d$   
high affinity

2 other considerations

↳  $k_{off}$

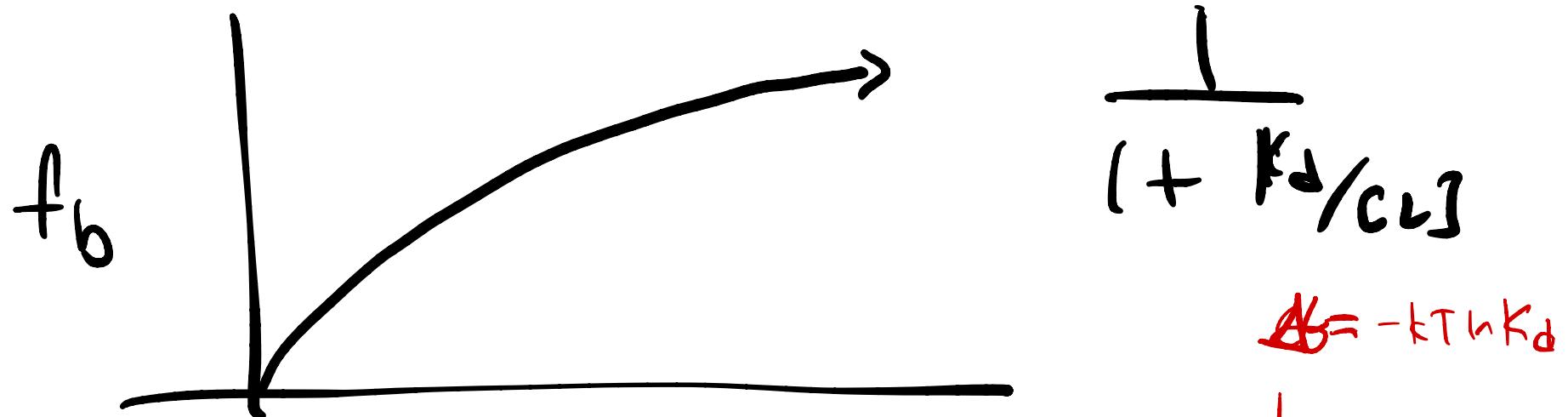
↳ bio availability



$$K_b = \frac{k_{on}}{k_{off}} = \frac{[ML]}{[M][L]}$$

$$f_{\text{bound}} = \frac{[ML]}{[M] + [ML]} = \frac{1}{1 + \frac{k_d}{[L]}}$$

$$f([L] = K_d) = \frac{1}{2}$$

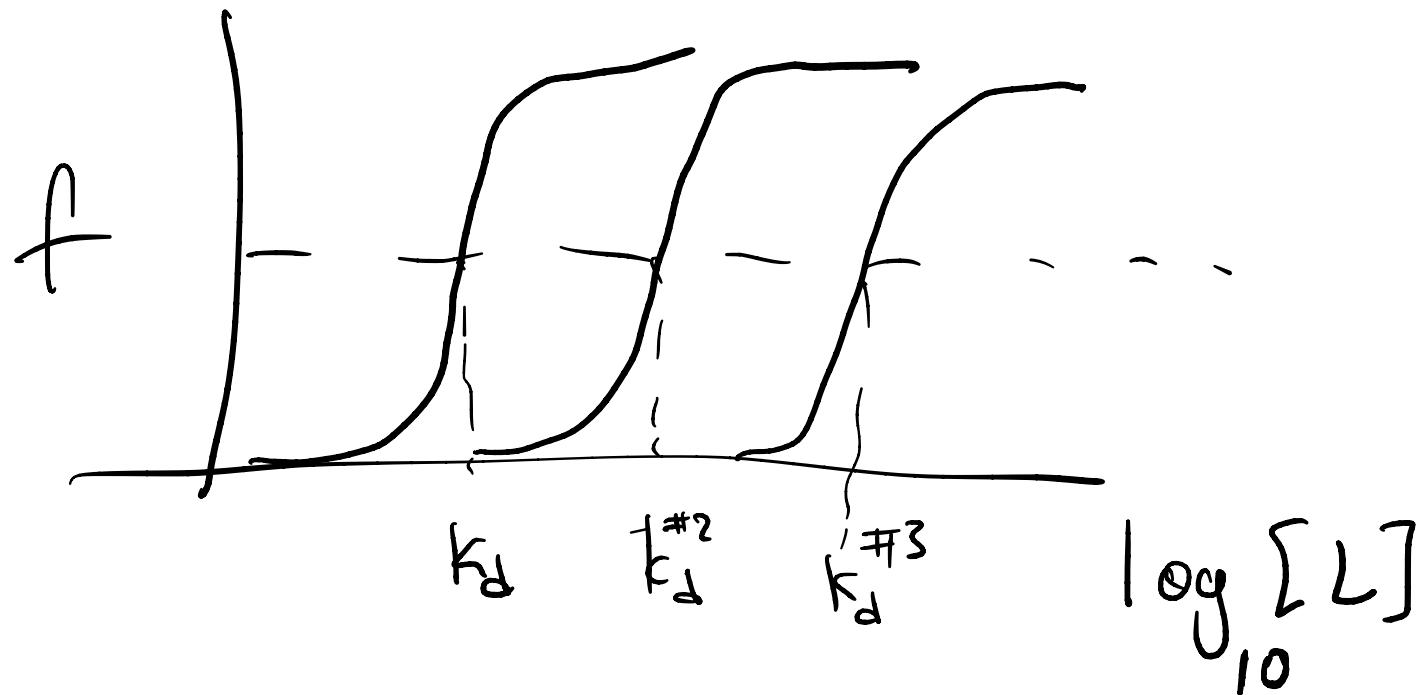


$$\frac{1}{1 + K_d/C_L}$$

$$\Delta G = -kT \ln K_d$$

$$f_b = \frac{1}{1 + \frac{1}{K_b[L]}} = \frac{1/(1+K_d)}{1/(1+K_d) + 1/(K_b[L]+1)} = \frac{K_b[L]}{K_b[L] + 1}$$

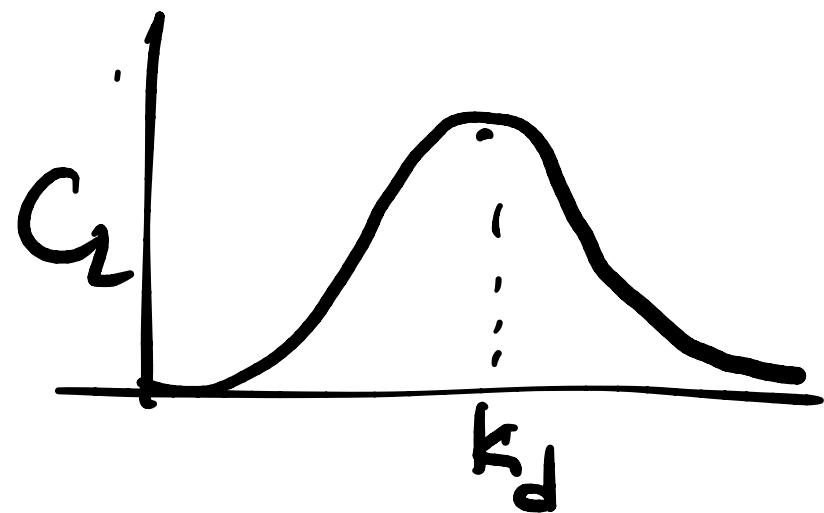
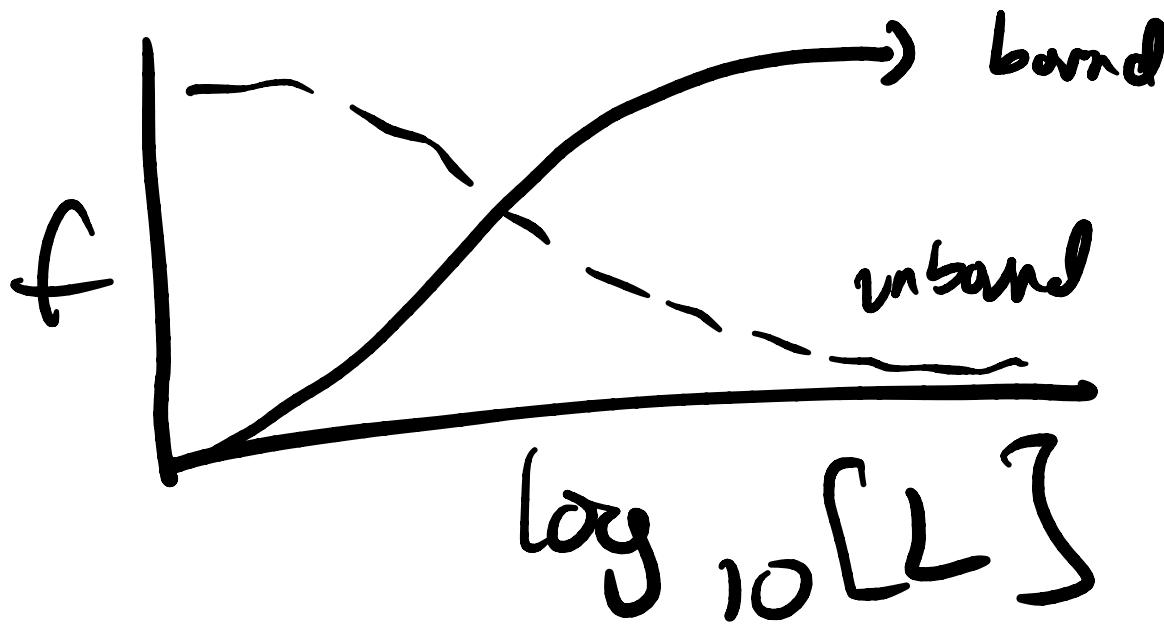
$$K = K_b[L] \sim \frac{1}{1+K}$$



real  $K_{ds}$  range from  $\mu M - mM$

Binding capacity

$$C_L = \frac{df}{d\log_{10}[L]} = 2 \cdot 3 \frac{df}{dnL}$$



$$\begin{aligned}
 C_L &= 2.3 \frac{df}{d \ln(L)} \leftrightarrow 2.3 \frac{k[L]}{(1 + k[L])^2} \\
 &= 2.3 f_{\text{bound}} f_{\text{unbound}} \\
 &= 2.3 f_{\text{bound}} (1 - f_{\text{bound}})
 \end{aligned}$$

$[L]$  is the free ligand conc.

$$[L]_{\text{tot}} = [L] + [ML]$$

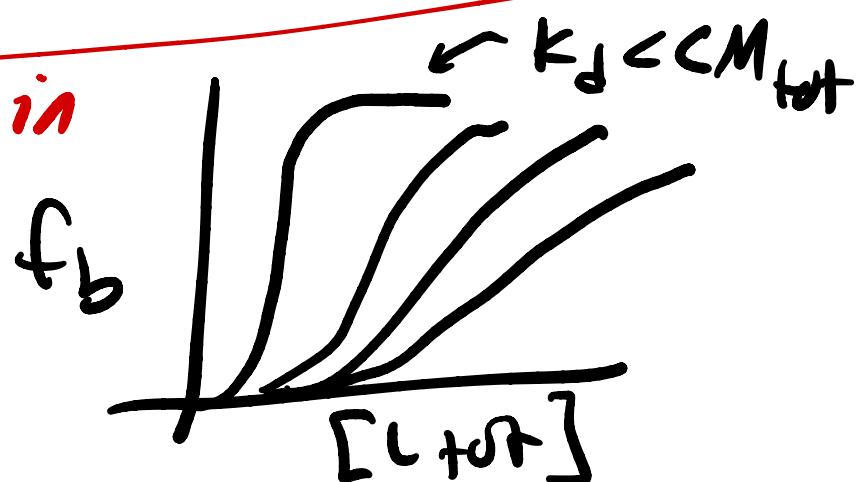
$$[M]_{\text{tot}} = [M] + [M2]$$

$$K_b = \frac{[ML]}{([M]_{\text{tot}} - [ML])([L]_{\text{tot}} - [ML])}$$

← Solve for  $M_L$

$$f = \frac{[ML]}{[M]_{\text{tot}}}$$

← plug in



# Hill plot

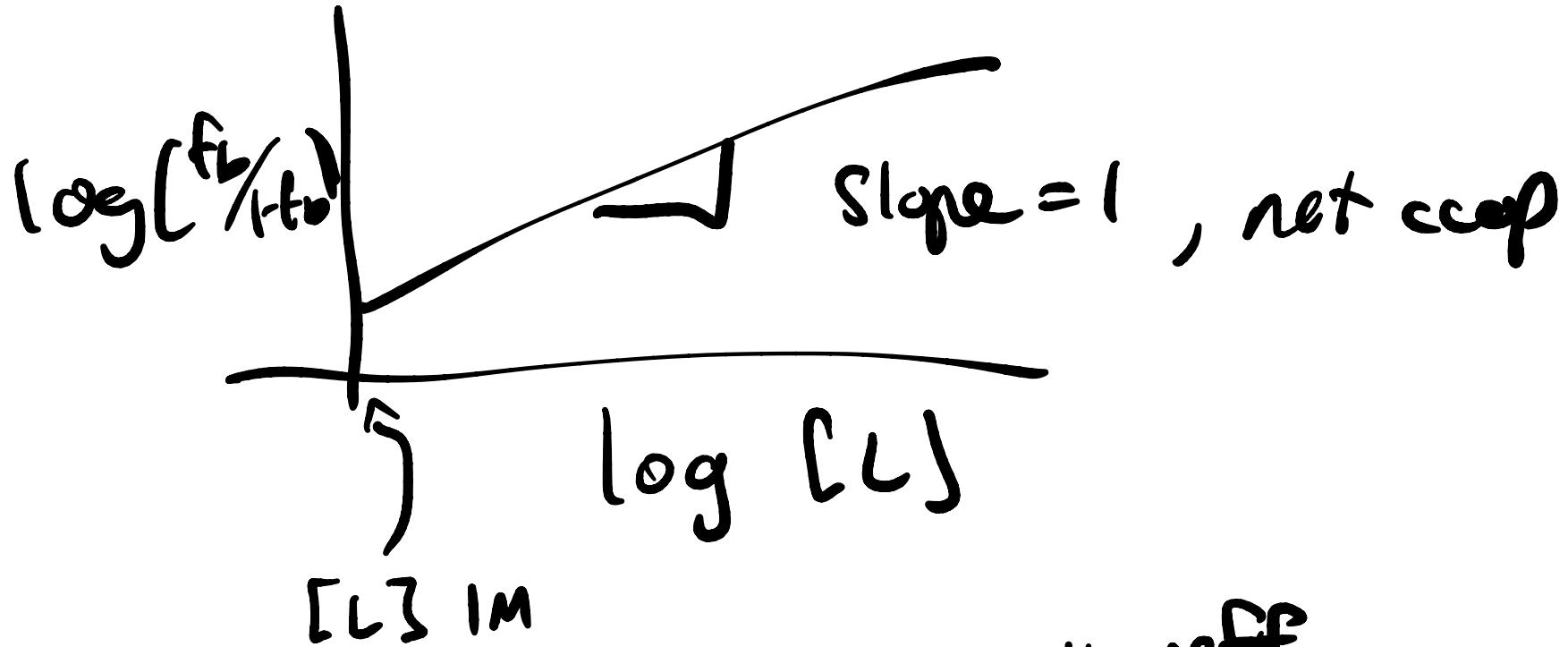


$$f_b = \frac{[L]k_b}{1 + [L]K_b}$$

$$f_u = 1 - f_b \\ = \frac{1}{1 + K_b [L]}$$

$$\log\left(\frac{f_b}{f_u}\right) = \log([L]k_b) = \log [L] + \log k_b$$

$$\log\left(\frac{f_b}{f_a}\right) = \log([L]k_b) = \log[L] + \log k_b$$



$$\frac{d \log\left[\frac{f_b}{(1-f_b)}\right]}{d \log[L]} = n \quad , \quad \text{hill coeff}$$

Non coop

$$n=1$$

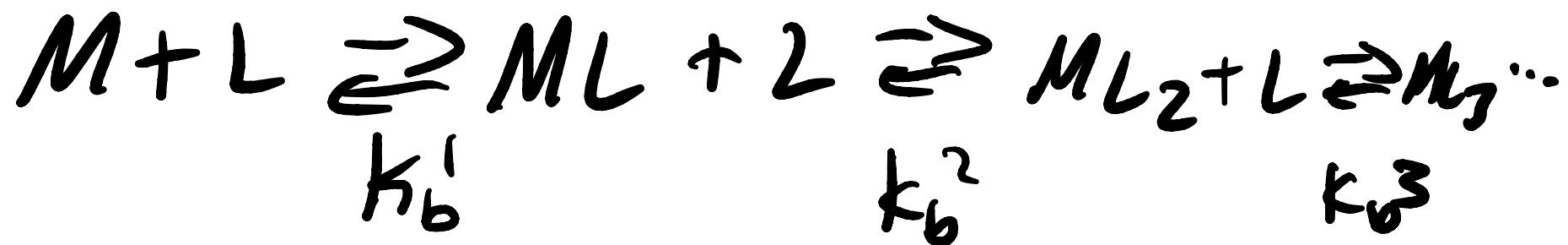
coop

$$n > 1$$

$$\frac{1}{1 + \left(\frac{k_d}{k_L}\right)^n} = f_{\text{bund}}$$

$$\frac{\partial \log f}{\partial \log x} = \left( \frac{\partial x}{\partial \log x} \right) \left( \frac{\partial \log f}{\partial x} \right) = x \frac{\partial \log(f)}{\partial x}$$

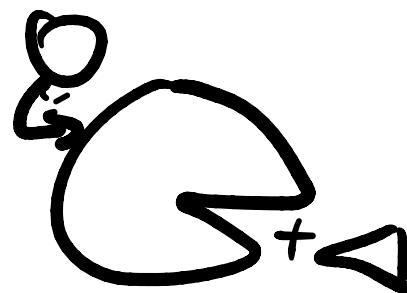
# Binding of multiple ligands



positive coop,  $k$ 's increase

neg coop,  $k$ 's decrease

allostery:

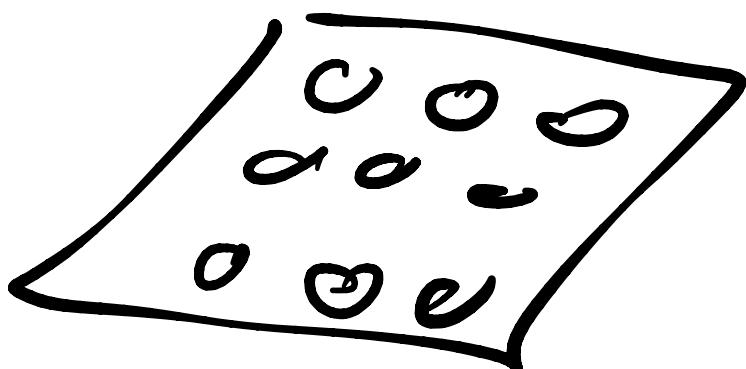


$$\beta_N = k_1 k_2 \cdots k_N$$

$$= \frac{[MC_N]}{[M][L]^N}$$

S binding sites

$$f_b = \frac{1}{s} \frac{\sum_i \beta_i [L]^i}{\sum \beta_i [L]^i} \leftarrow \text{grand canonical}$$



$$P = \sum \beta_i [L]^i$$