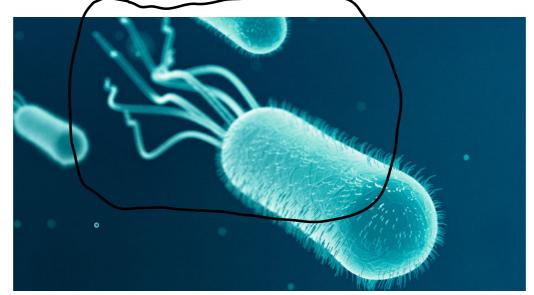
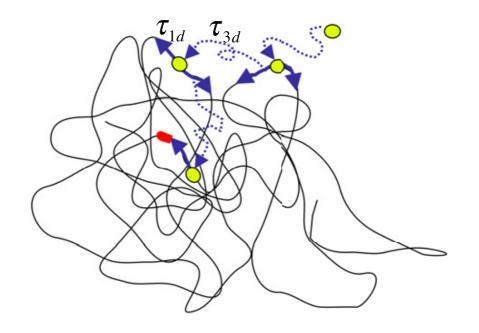
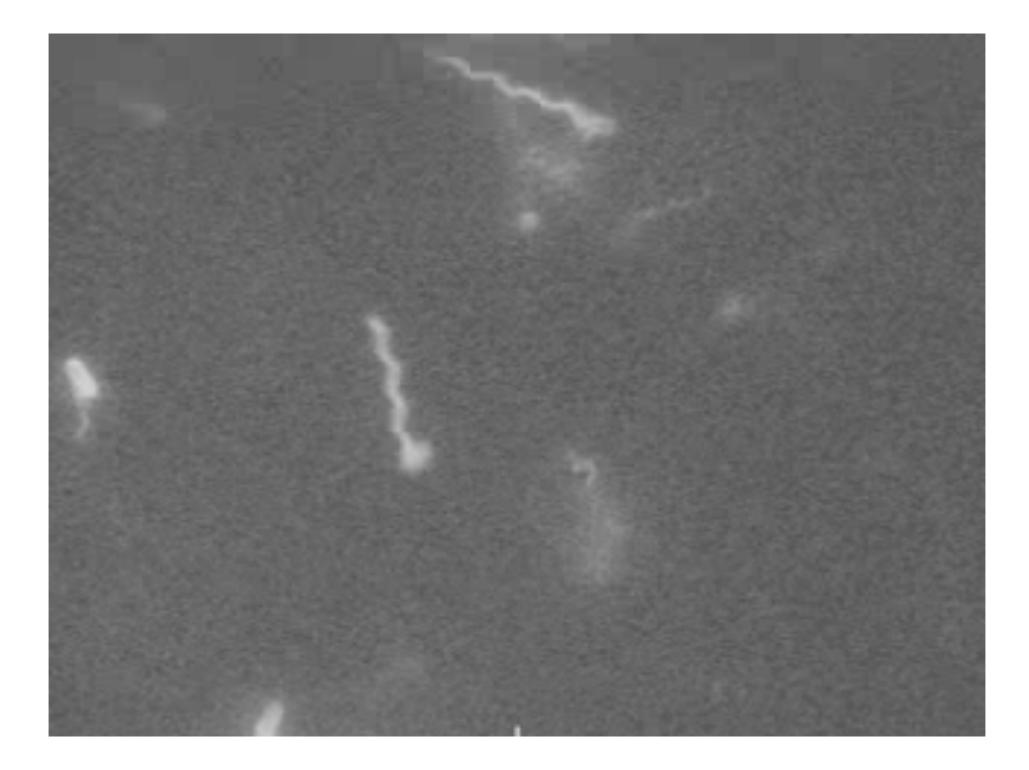
MAE 545: Lecture 18 (4/25) E. coli chemotaxis (continued)



How proteins find target sites on DNA?

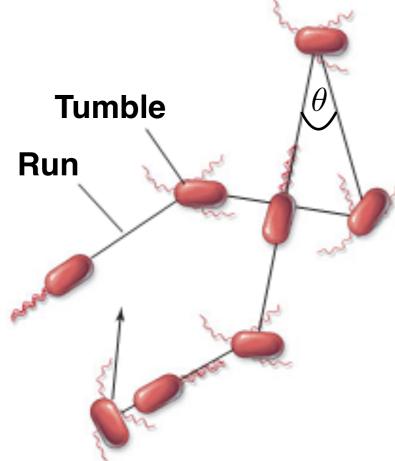


E. coli chemotaxis

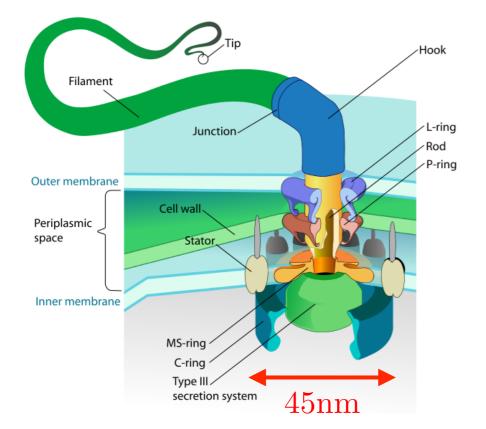


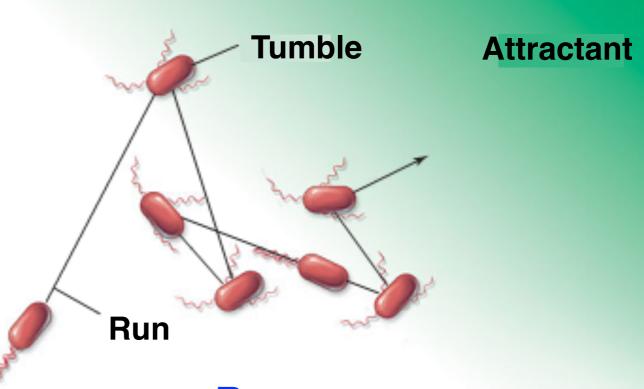
L. Turner, W.S. Ryu, H.C. Berg, <u>J. Bacteriol.</u> **182**, 2793-2801 (2000)

E. coli chemotaxis



Rotary motor





Run

swimming speed: $v_s \sim 20 \mu m/s$

typical duration: $t_r \sim 1 {
m s}$

all motors turning counter clockwise

Increase (Decrease) run durations, when swimming towards good (harmful) environment.

Tumble

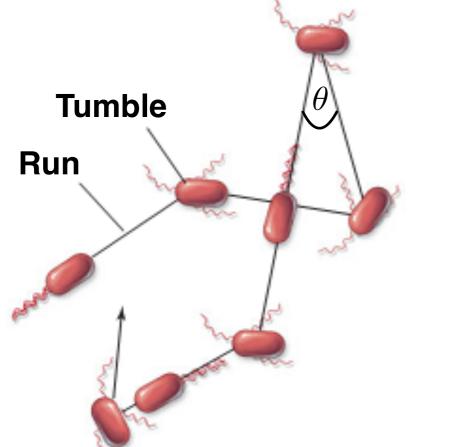
random change in orientation $\langle \theta \rangle = 68^{\circ}$

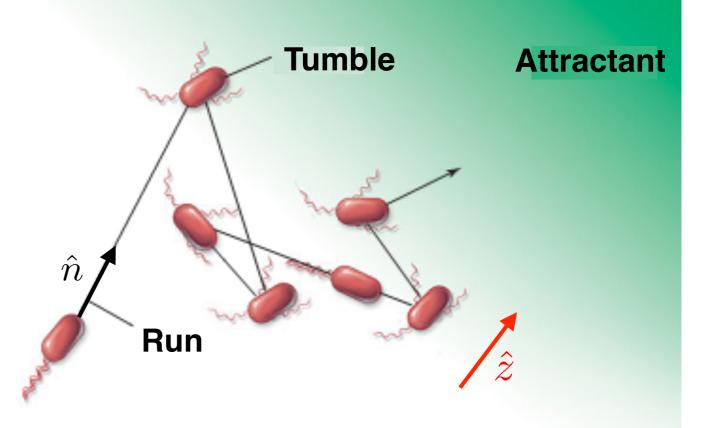
typical duration: $t_t \sim 0.1 \mathrm{s}$

one or more motors turning clockwise

E. coli chemotaxis

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Homogeneous environment

run duration: $t_r \sim 1 \mathrm{s}$ tumble duration: $t_t \sim 0.1 \mathrm{s}$ swimming speed: $v_s \sim 20 \mu \mathrm{m/s}$

drift velocity

 $v_d =$

effective diffusion

$$D_{\text{eff}} = \frac{\left\langle \Delta \ell^2 \right\rangle}{6 \left\langle \Delta t \right\rangle}$$
$$D_{\text{eff}} \approx \frac{v_s^2 t_r^2}{6(t_r + t_t)} \sim 60 \mu \text{m}^2/\text{s}$$

Gradient in "food" concentration

run duration increases (decreases) when swimming towards (away) from "food"

$$t_r(\hat{n}) = \bar{t}_r + \alpha(\hat{n} \cdot \hat{z})(\partial c/\partial z)$$

drift velocity

$$v_d = \frac{\langle \Delta z \rangle}{\langle \Delta t \rangle} \approx \frac{v_s \alpha (\partial c / \partial z)}{3(\bar{t}_r + t_t)}$$

$$\langle \Delta z \rangle = \langle v_z(\hat{n}) t_r(\hat{n}) \rangle = \langle v_s(\hat{n} \cdot \hat{z}) t_r(\hat{n}) \rangle$$

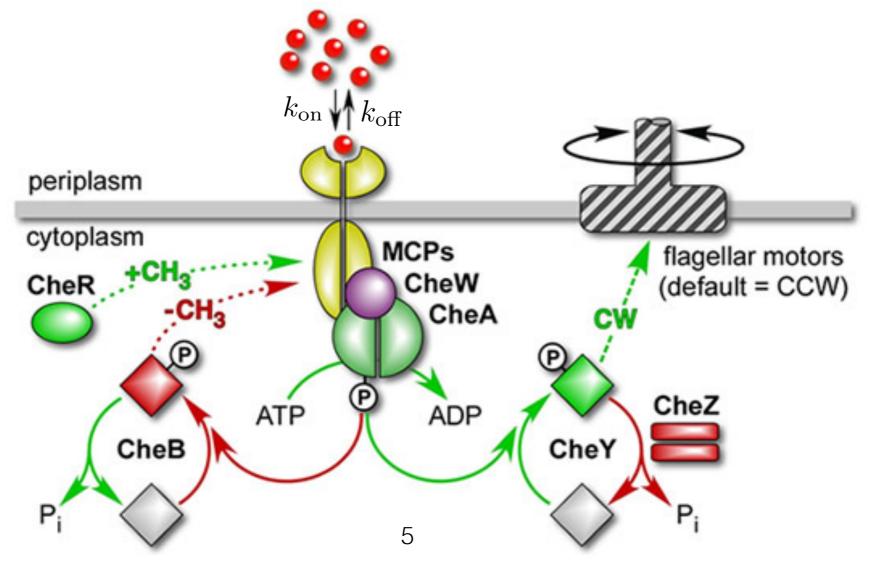
Sensing of environment

E. coli surface is covered with receptors, which can bind specific molecules.

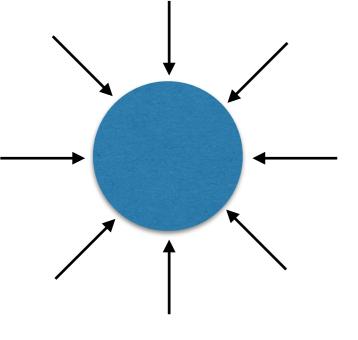
Average fraction of bound receptors p_B is related to concentration *c* of molecules.



Chemical signaling network inside E. coli analyzes state of receptors and gives direction to rotary motor.



Diffusion limited flux of molecules to E. coli



1917 Smoluchowski theory

Fick's law

$$\frac{\partial c}{\partial t} = D\nabla^2 c = D\frac{1}{r^2}\frac{\partial}{\partial r}\left(r^2\frac{\partial c}{\partial r}\right)$$

boundary conditions

$$c(r \to \infty) = c_{\infty}$$
$$c(R) = 0$$

steady state $c(r) = c_{\infty} \left[1 - \frac{R}{r} \right]$

flux density of molecules

 $J(r) = -D\frac{\partial c(r)}{\partial r} = -\frac{Dc_{\infty}R}{r^2}$

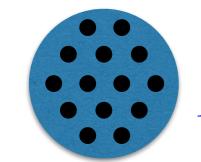
absorbing sphere

rate of absorbing molecules

 $I(r) = J(r) \times 4\pi r^2 = -4\pi DRc_{\infty} = I_0 = -k_{\rm on}c_{\infty}$

diffusion constant for small molecules





N equally spaced absorbing disks of radius s

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E. coli can use many types of receptors specific for different molecules, without significantly affecting the diffusive flux

Accuracy of concentration measurement

How many molecules do we expect inside a volume occupied by E. coli?

 $\overline{N} \sim R^3 c$

Probability *p(N)* that cell measures *N* molecules follows Poisson distribution

 $p(N) = \frac{\overline{N}^N E^{-\overline{N}}}{N!} \qquad \text{mean } \overline{N} \qquad \begin{array}{l} \text{standard} \\ \text{deviation} \quad \sigma_N = \sqrt{\overline{N}} \end{array}$

Error in measurement

$$\operatorname{Err} \sim \frac{\sigma_N}{\overline{N}} \sim (R^3 c)^{-1/2} \qquad \text{for } c = 1\mu \mathrm{M} = 6 \times 10^{20} \mathrm{m}^{-3} \Rightarrow \mathrm{Err} \sim 4\%$$

E.coli can be more precise by counting molecules for longer time t. However, they need to wait some time t_0 in order for the original molecules to diffuse away to prevent double counting of the same molecules!

 $t_0 \sim R^2/D \sim 10^{-3}s$ $\overline{N} \sim R^3 ct/t_0 \sim DRct$ for *t*=1s, precision Err $\sim (DRct)^{-1/2}$ improves to Err~0.1%

When E. coli is swimming, it wants to swim faster than the diffusion of small molecules

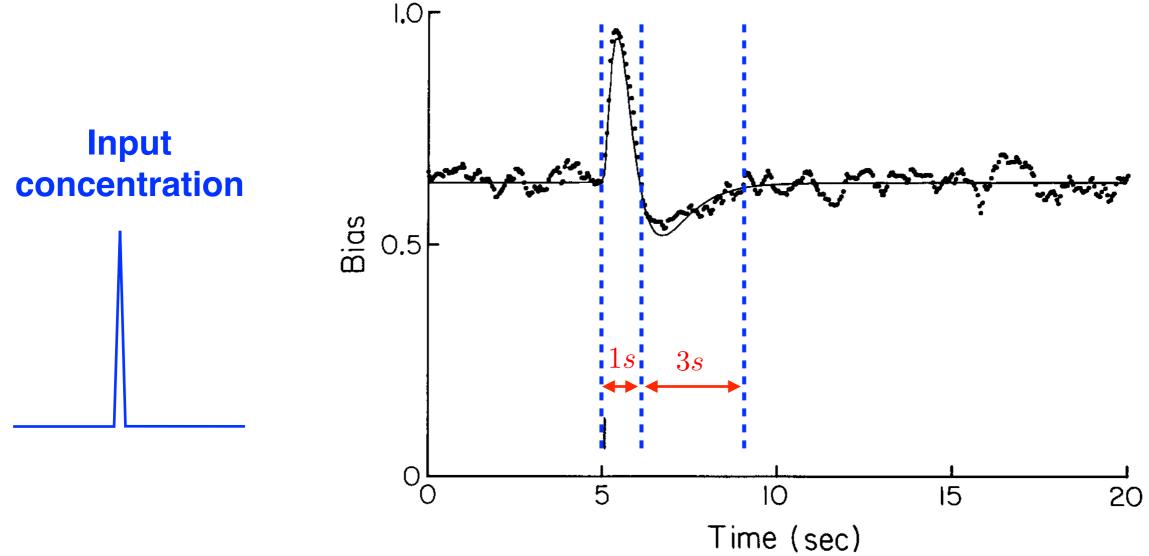
$$v_s t \gtrsim (Dt)^{1/2} \Rightarrow t \gtrsim D/v_s^2 \sim 1s$$

Molar concentration

 $1M = 6 \times 10^{26} \mathrm{m}^{-3}$

How E. coli actually measures concentration?

Probability for motor to rotate in CCW direction (runs) as a function of time in response to short pulse in external molecular concentration



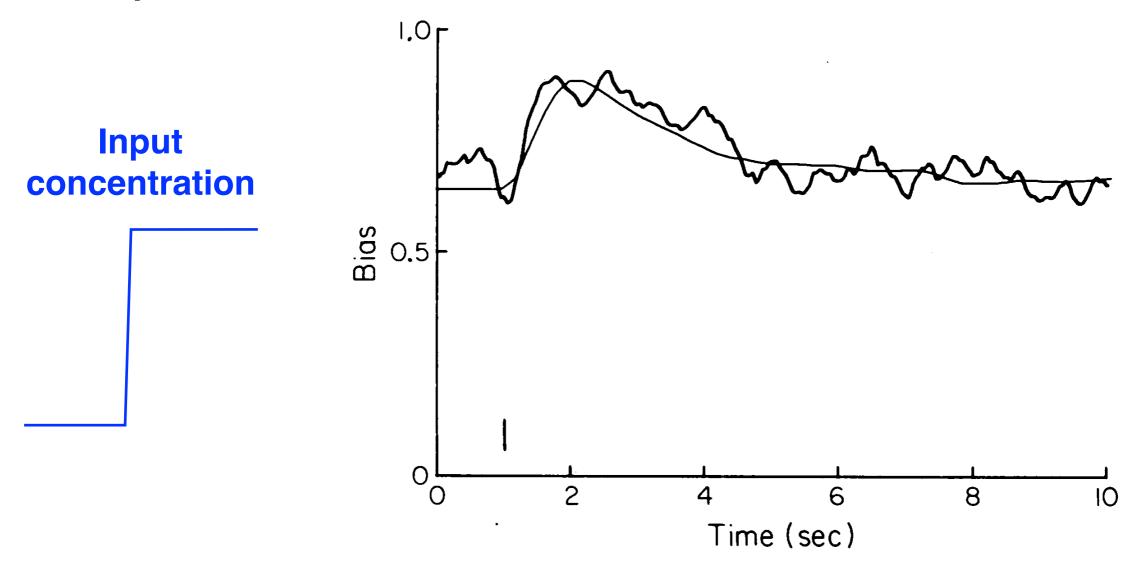
E. coli integrates measured concentration observed during the last second and compare this with measured concentration during the previous 3 seconds. If difference is positive then increase the probability of runs, otherwise increase the probability of tumbles.

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J. E. Segall, S. M. Block, and H. C. Berg, <u>PNAS</u> 83, 8987–8991 (1986)

Adaptation

Probability for motor to rotate in CCW direction (runs) as a function of time in response to a sudden increase in external molecular concentration

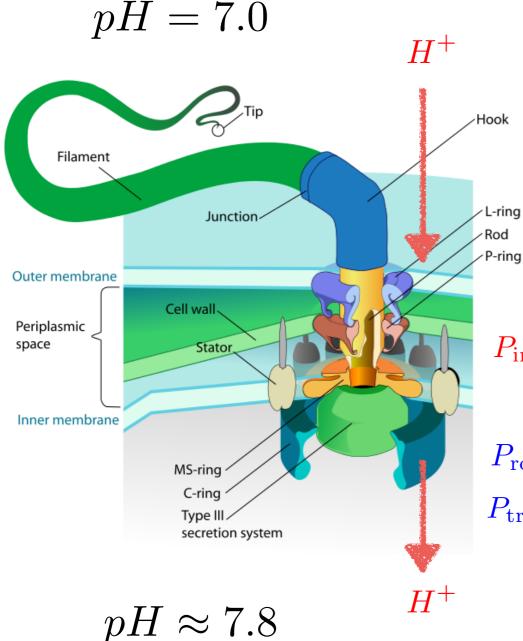


E. coli adapts to the new level of concentration in about 4 seconds. This enables E. coli to be very sensitive to changes in concentration over a very broad range of concentrations!

J. E. Segall, S. M. Block, and H. C. Berg, <u>PNAS</u> 83, 8987–8991 (1986)

How efficient is motor of E. coli?

Energy source for rotary motor are charged protons



Each proton gains energy due to

Transmembrane electric potential difference

 $\delta\psi\approx-120\mathrm{mV}$

Change in pH $\delta U = (-2.3k_BT/e)\Delta pH \approx -50mV$

Total protonmotive force

 $\Delta p = \delta \psi + \delta U \approx -170 \mathrm{mV}$

Need 1200 protons per one body revolution

Input power

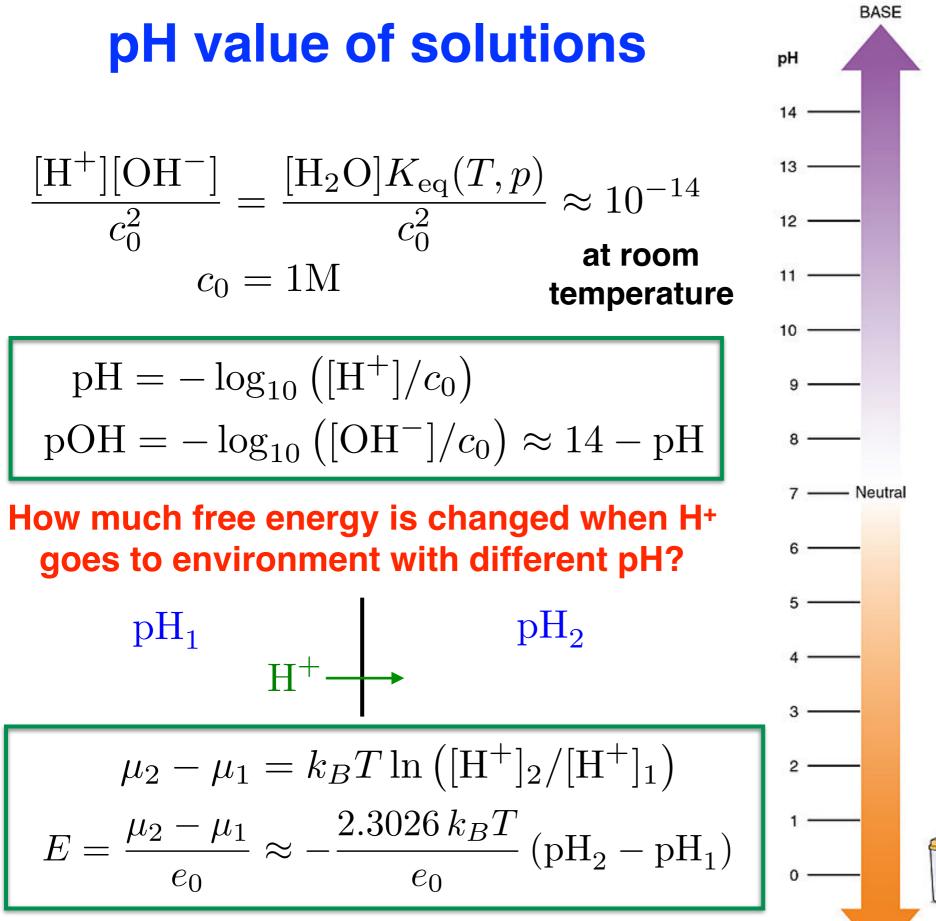
 $P_{\rm in} = n \times e\Delta p \times f = 1200 \times 0.17 \text{eV} \times 10 \text{Hz} \approx 3.2 \times 10^5 \text{pN nm/s}$

Power loss due to Stokes drag

 $P_{\rm rot} = N \times (2\pi f) \approx 4600 \text{pN nm} \times (20\pi \text{Hz}) \approx 2.9 \times 10^5 \text{pN nm/s}$ $P_{\rm trans} = F \times v \approx 0.4 \text{pN} \times 20000 \text{nm/s} \approx 8 \times 10^3 \text{pN nm/s}$

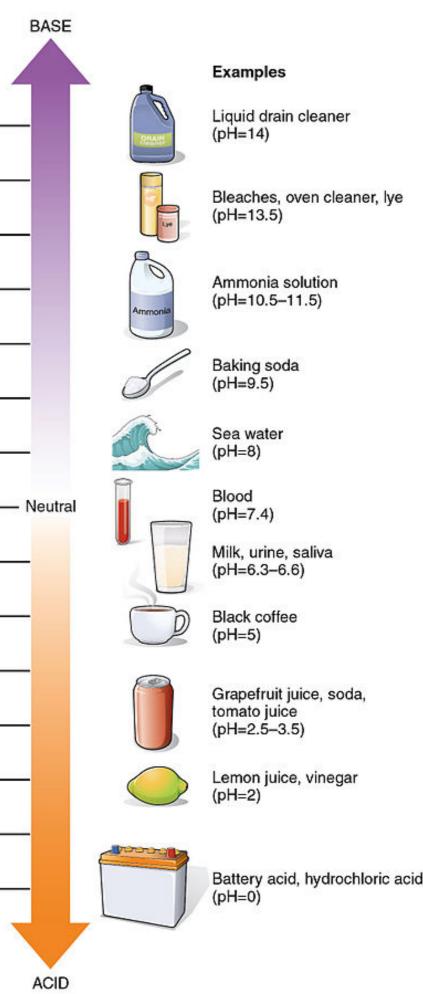
Motor efficiency

 $\frac{P_{\rm trans} + P_{\rm rot}}{P_{\rm in}} \approx 90\%$

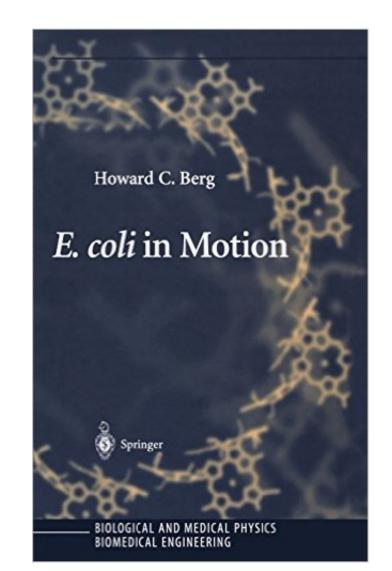


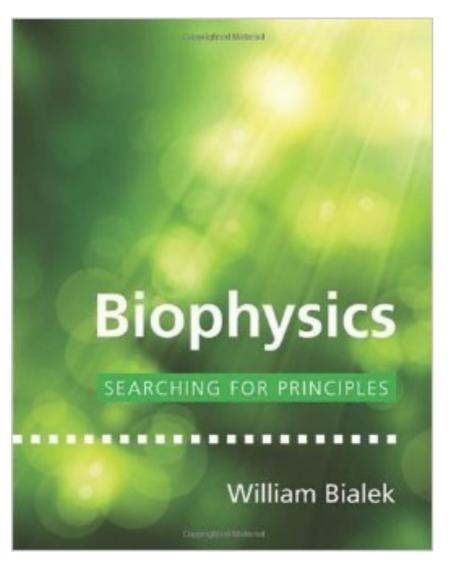
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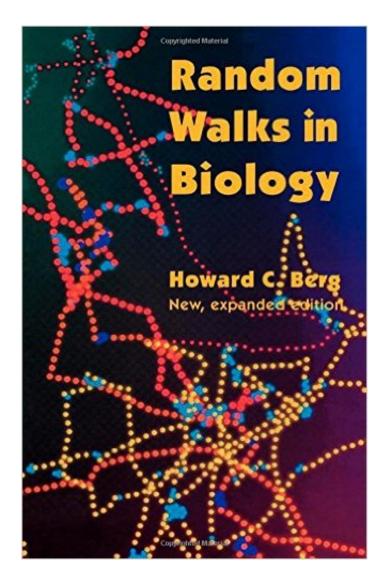
Nernst electric potential *E*



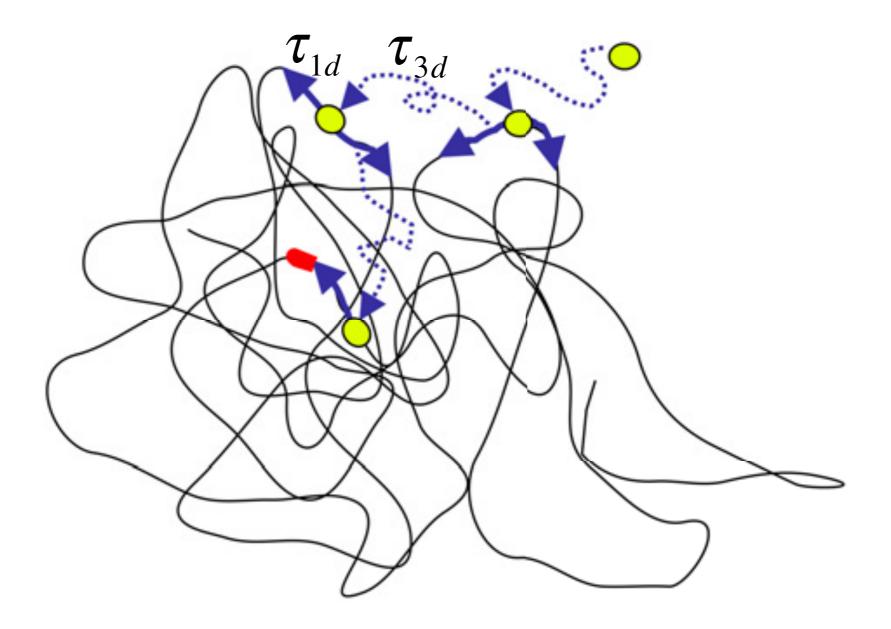
Further reading



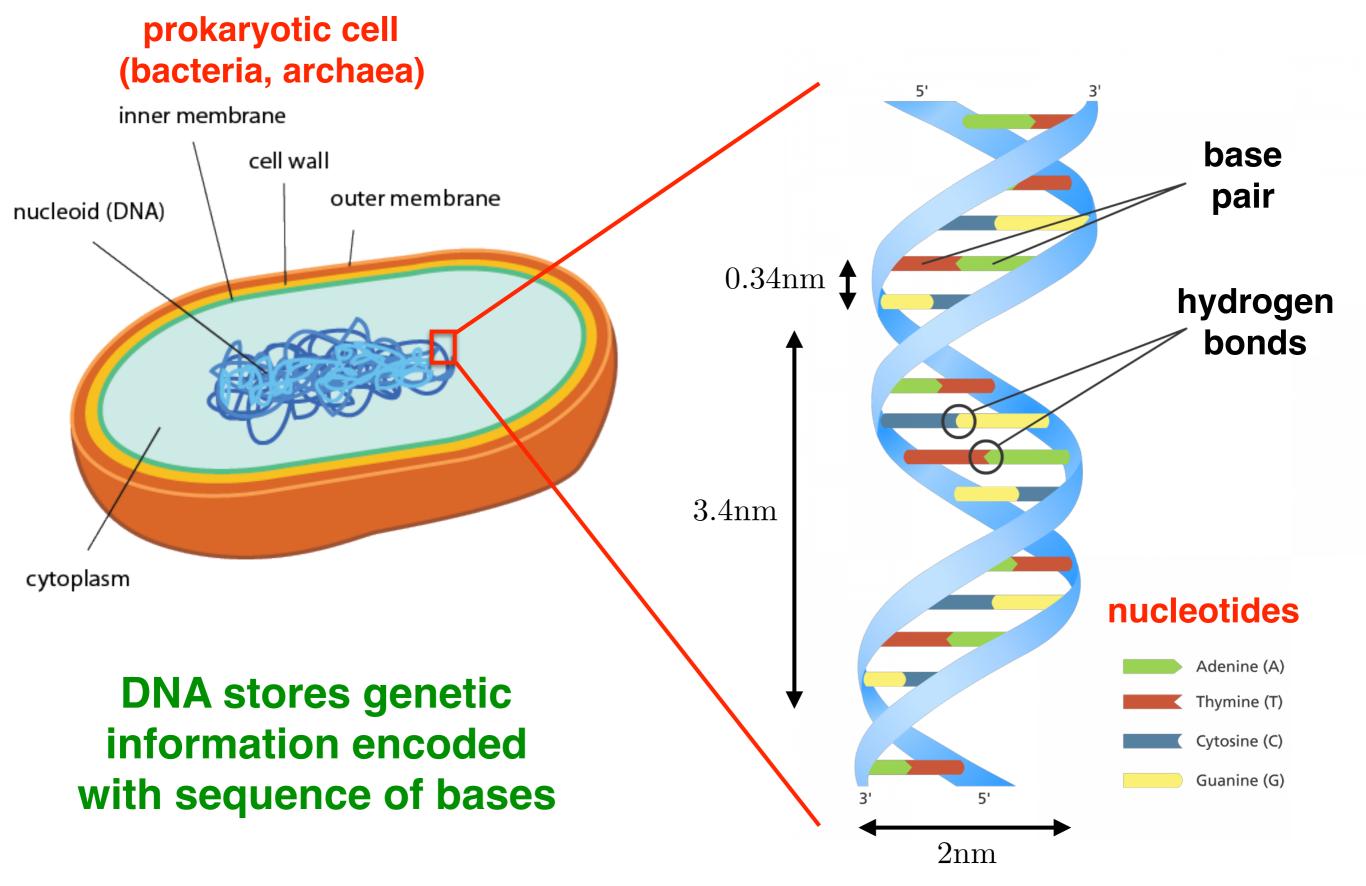


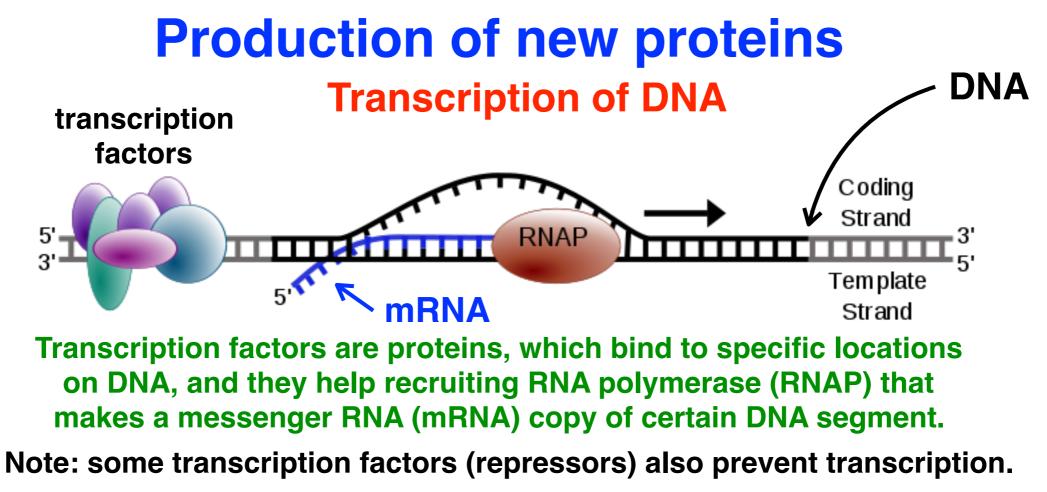


How proteins find target sites on DNA?

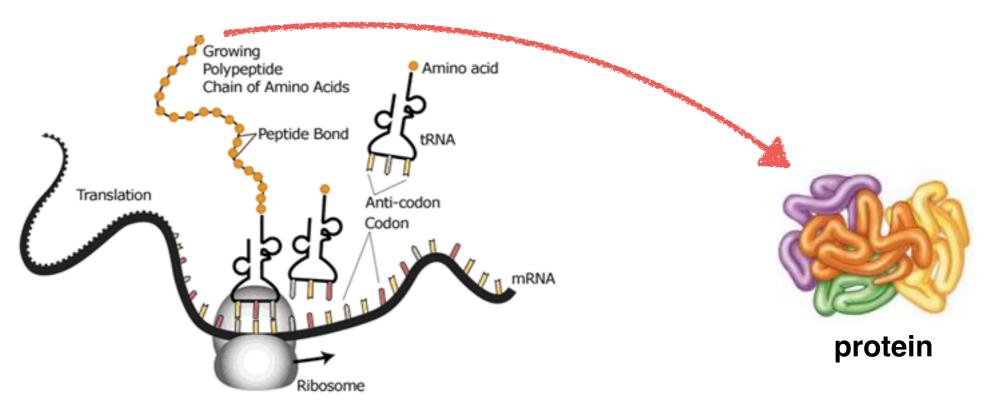


DNA

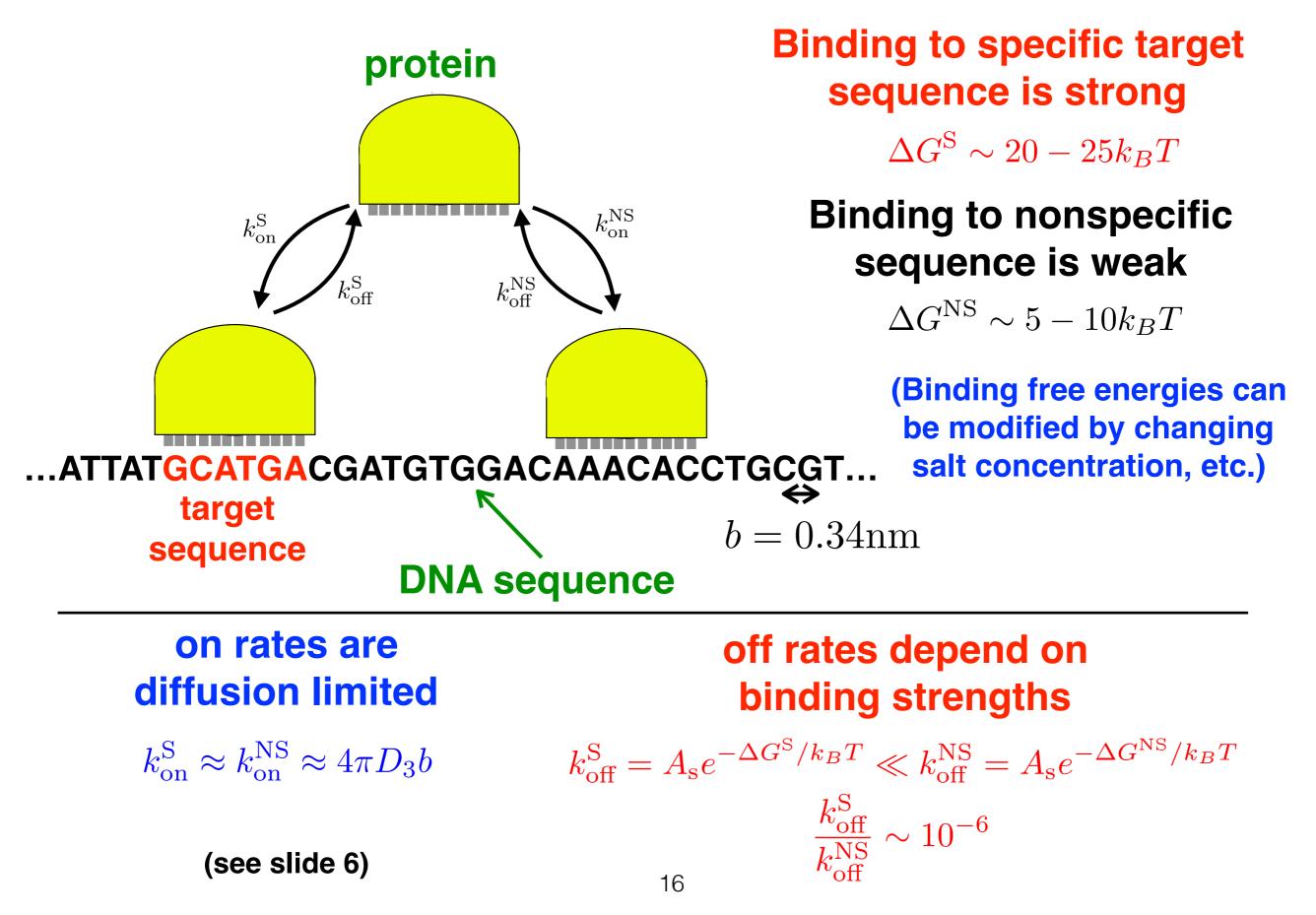




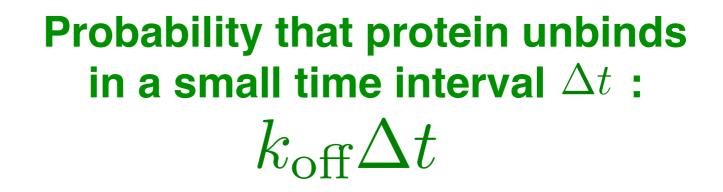
Translation of mRNA



Protein-DNA interactions



How long proteins remain bound on DNA?



Probability that protein remains bound for time t and then it unbinds between time t and $t+\Delta t$:

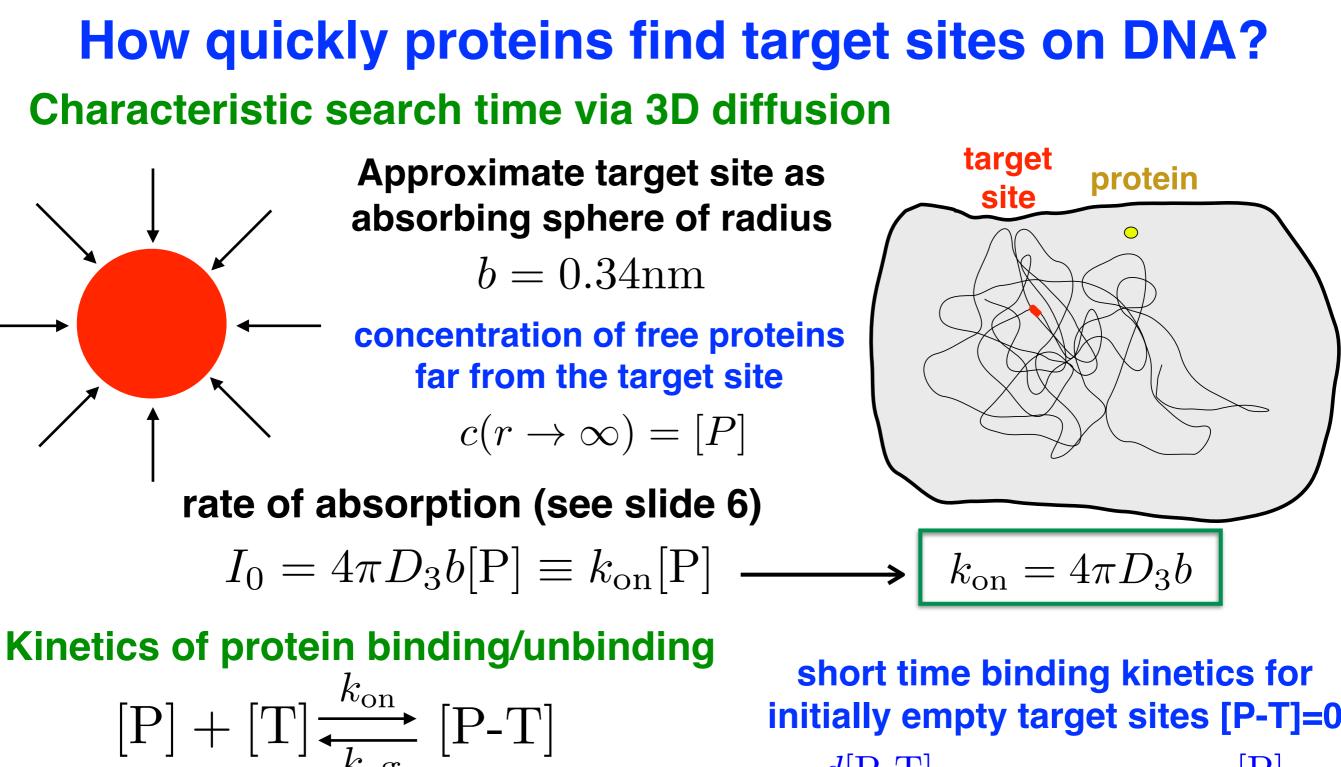
$$k_{\text{off}}\Delta t \times (1 - k_{\text{off}}\Delta t)^{t/\Delta t}$$

$$\lim_{t \to 0} \int b_{\text{off}} e^{-k_{\text{off}}t}$$

$$p(t) = k_{\text{off}} e^{-k_{\text{off}}t}$$

Average binding time $\langle t \rangle = \int_0^1 t \, p(t) dt = \frac{1}{k_{\text{off}}}$

Proteins remain bound to specific target sites for minutes to hours, while they unbind from nonspecific sites after milliseconds to seconds.



$$\frac{d[P-T]}{dt} = k_{\rm on}[P][T] - k_{\rm off}[P-T]$$

[P-T] concentration of proteins bound to target sites [P] concentration of free proteins [T] concentration of empty target sites 18

initially empty target sites [P-T]=0

$$\frac{d[\text{P-T}]}{dt} = (k_{\text{on}}[\text{T}])[\text{P}] \equiv \frac{[\text{P}]}{t_s}$$

characteristic search time

$$t_s = (k_{\rm on}[T])^{-1}$$

How quickly proteins find target sites on DNA?

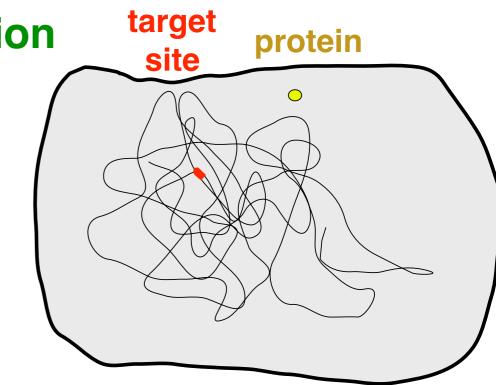
Characteristic search time via 3D diffusion

$$k_{\rm on} = 4\pi D_3 b$$
 $t_s = (k_{\rm on}[{\rm T}])^{-1}$

1917 Smoluchowski theory

Example: characteristic search time for lac repressor protein in E. coli

 $b \approx 0.34 \text{nm}$ $D_3 \approx 30 \mu \text{m}^2/\text{s}$ [T] ~ 1 per cell ~ $10^{-9}M$ $k_{\text{on}} \sim 10^8 M^{-1} s^{-1}$ $t_s \sim 10s$



Molar concentration $1M = 6 \times 10^{26} \text{m}^{-3}$

in vitro experiments (1970) $k_{\rm on}^{\rm exp} \sim 10^{10} M^{-1} s^{-1}$ $t_s \sim 0.1s$ <u>J. Mol</u>

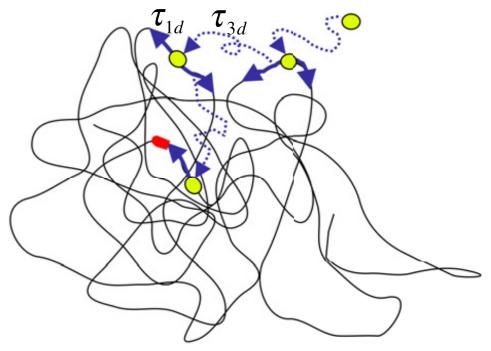
A.D.Riggs *et al.*, <u>J. Mol. Biol.</u> **53**, 401-417 (1970)

Why is experimentally observed rate 100 times larger?

Berg - von Hippel theory (1980s)

(facilitated diffusion)

- 1. Proteins diffuse in space and nonspecifically bind to a random location on DNA.
- 2. Proteins slide (diffuse) along the DNA.
- 3. Proteins jump (diffuse) to another random location on DNA and continue this sliding/ jumping process until the target site is found.



 $b = 0.34 \mathrm{nm}$ L - DNA length D_3 - diffusion constant in space D_1 - diffusion constant along the DNA

How long that is it take to find a target site in this process?

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O.G.Berg et al., <u>Biochemistry</u> **20**, 6929-48 (1981)

Berg - von Hippel theory (1980s)

First assume fixed sliding time τ_{1d}

Number of distinct sites visited during each sliding event

 $n = \sqrt{16D_1 \tau_{1d} / (\pi b^2)}$

(valid for n >> 1)

Probability that target site is found during a sliding event

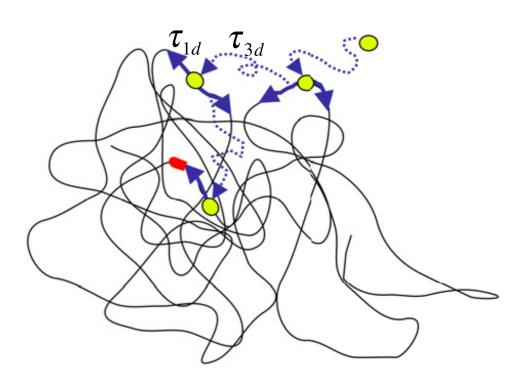
$$q = nb/L$$

Probability that target site is found exactly after *N_R* rounds

 $p(N_R) = q(1-q)^{N_R-1}$

Average number of rounds needed to find the target $\underset{\infty}{\infty}$

$$\overline{N_R} = \sum_{N_R=1} N_R p\left(N_R\right) = 1/q$$



 $b = 0.34 \mathrm{nm}$ L - DNA length D_3 - diffusion constant in space D_1 - diffusion constant along the DNA

Average search time

$$\overline{t_s} = \overline{N_R} \left(\tau_{1d} + \tau_{3d} \right)$$

O.G.Berg et al., <u>Biochemistry</u> **20**, 6929-48 (1981)

Facilitated diffusion

In reality sliding times are exponentially distributed

$$p(\tau_{1d}) = k_{\text{off}}^{\text{NS}} e^{-k_{\text{off}}^{\text{NS}}\tau_{1d}}$$
$$\langle \tau_{1d} \rangle = \int_0^\infty d\tau_{1d} \tau_{1d} p(\tau_{1d}) = 1/k_{\text{off}}^{\text{NS}}$$

Average number of distinct sites visited during each sliding

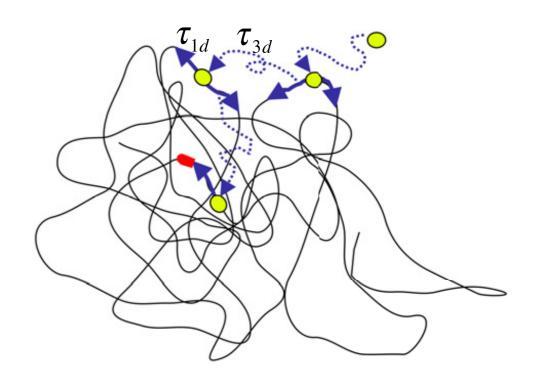
$$\langle n \rangle = \int_0^\infty d\tau_{1d} \, p(\tau_{1d}) \sqrt{16D_1 \tau_{1d} / (\pi b^2)}$$
$$\langle n \rangle = 2\sqrt{D_1 \langle \tau_{1d} \rangle / (b^2)}$$

Average probability that target site is found during a sliding event

 $\langle q \rangle = \langle n \rangle \, b/L$

Average number of rounds *N_R* needed to find the target site

$$\overline{\langle N_R \rangle} = 1/\left\langle q \right\rangle$$

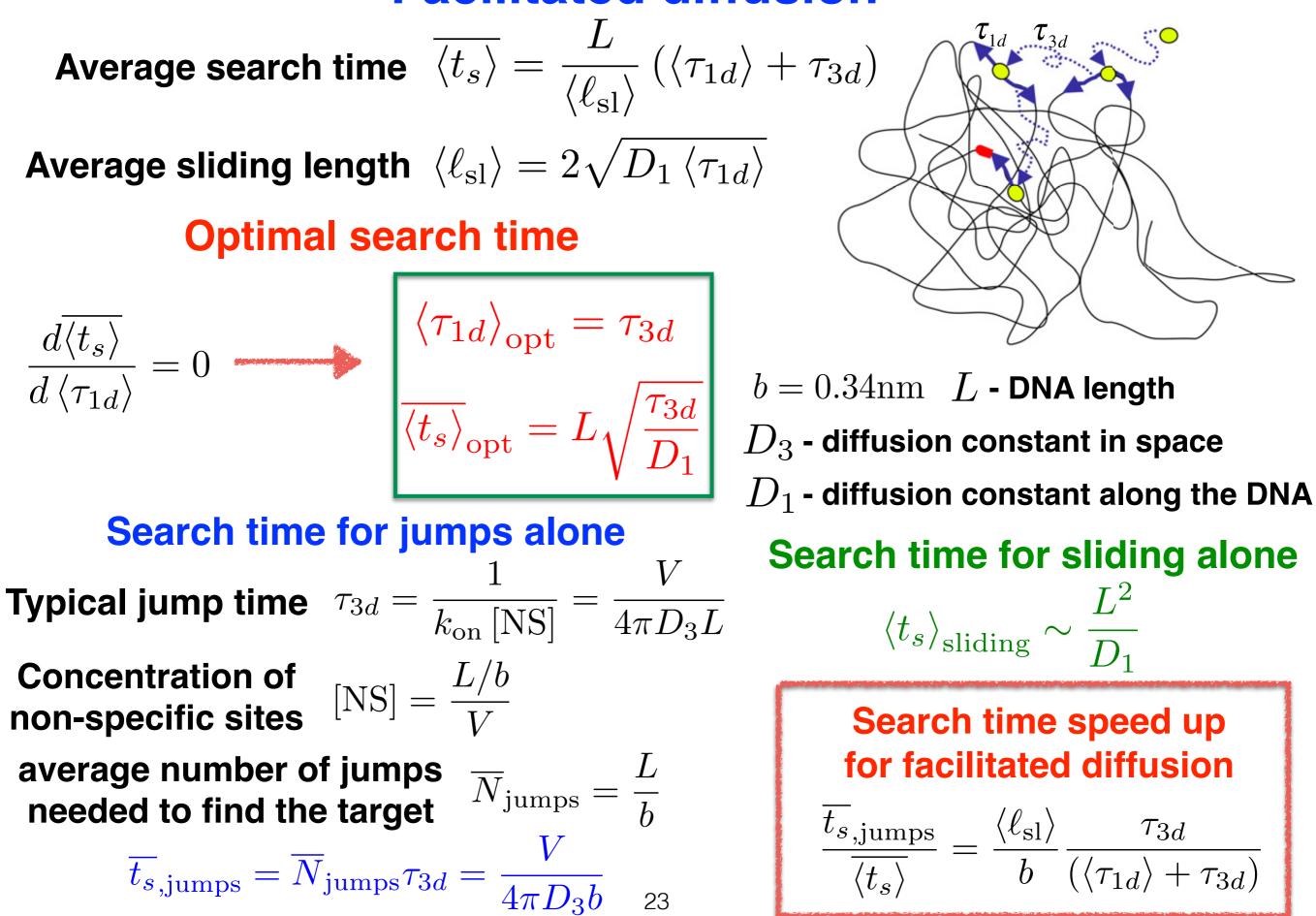


b = 0.34nm L - DNA length D_3 - diffusion constant in space D_1 - diffusion constant along the DNA

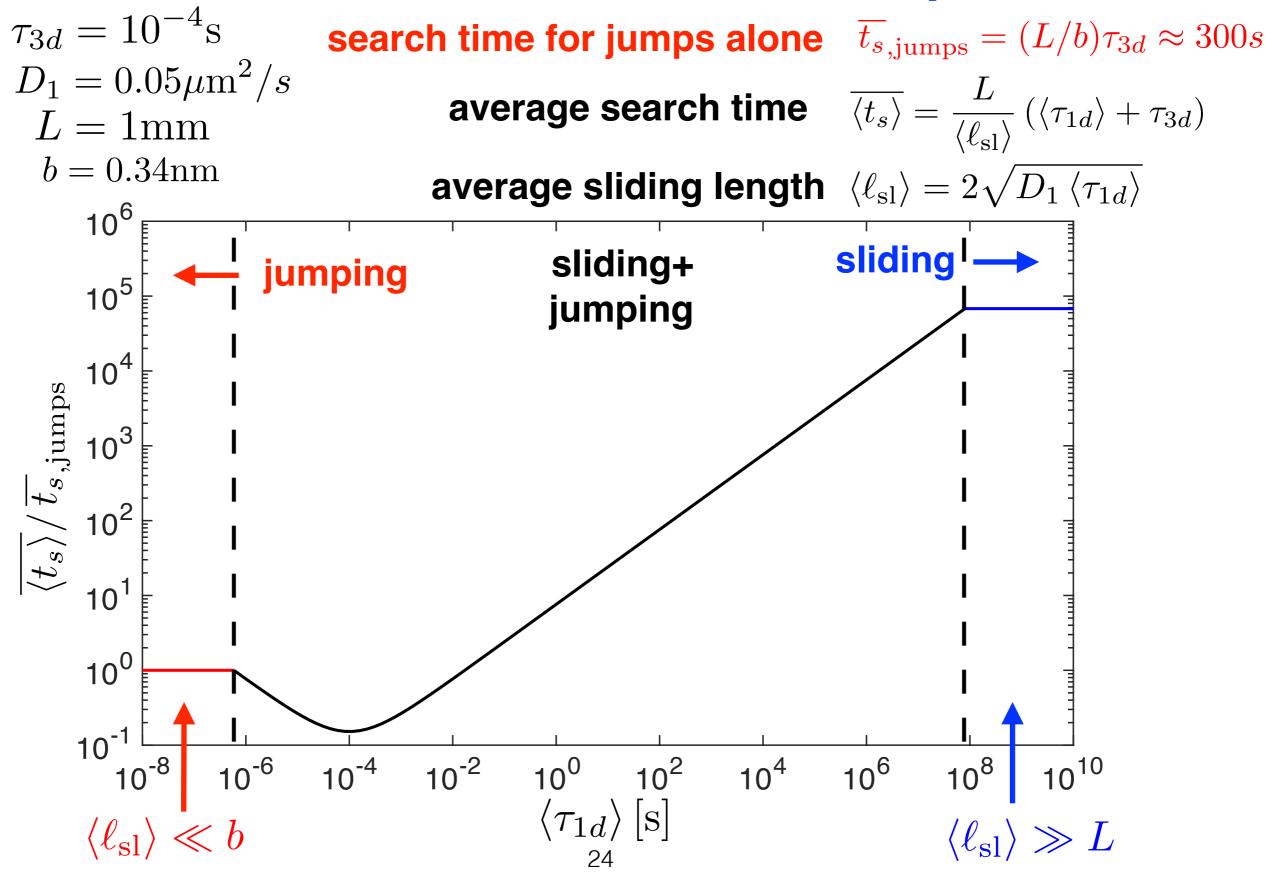
Average search time

$$\overline{\langle t_s \rangle} = \overline{\langle N_R \rangle} \left(\langle \tau_{1d} \rangle + \tau_{3d} \right)$$
$$\overline{\langle t_s \rangle} = \frac{L}{2\sqrt{D_1 \langle \tau_{1d} \rangle}} \left(\langle \tau_{1d} \rangle + \tau_{3d} \right)$$

Facilitated diffusion

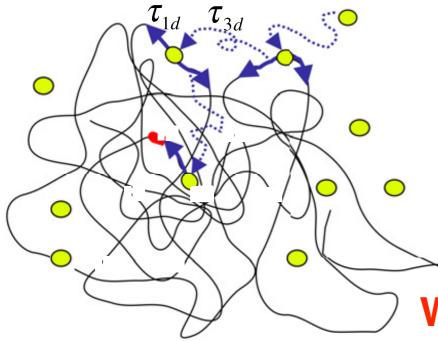


Example: search time for target site in bacteria on DNA with 10⁶ base pairs



Sing/Itaneous search for target site by multiple proteins

Interactions and collisions between proteins are ignored



Search times for target site by individual proteins are exponentially distributed

 $p_1(t_s) = \frac{1}{\overline{\langle t_s \rangle}} e^{-t_s/\overline{\langle t_s \rangle}}$

What is the typical search time for the fastest of *n* independently searching proteins?

(Extreme value distributions)

$$p_n(t_s) = n \times p_1(t_s) \times \left(\int_{t_s}^{\infty} dt' \ p_1(t') \right)^{n-1} = \frac{n}{\langle t_s \rangle} e^{-nt_s/\langle t_s \rangle}$$

n proteins finds the target site at time *t*_s

probability that one of probability that other n-1 proteins take longer time to find the target site

Average search time is reduced by factor *n*

 $\int dt_s t_s p_n(t_s) = \frac{\langle t_s \rangle}{n}$