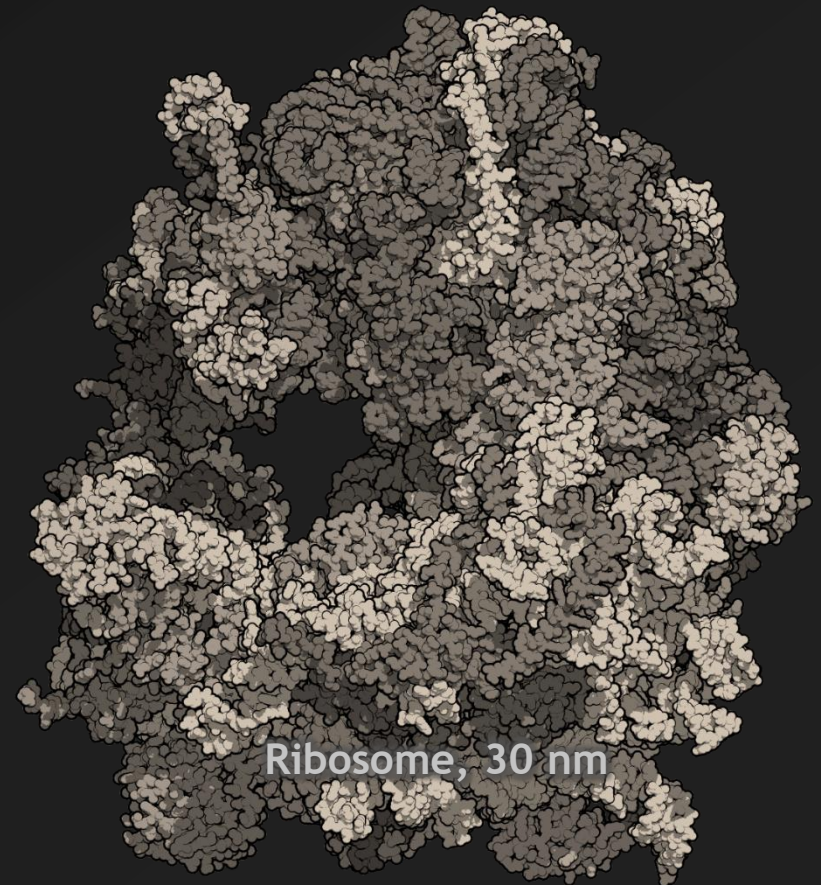
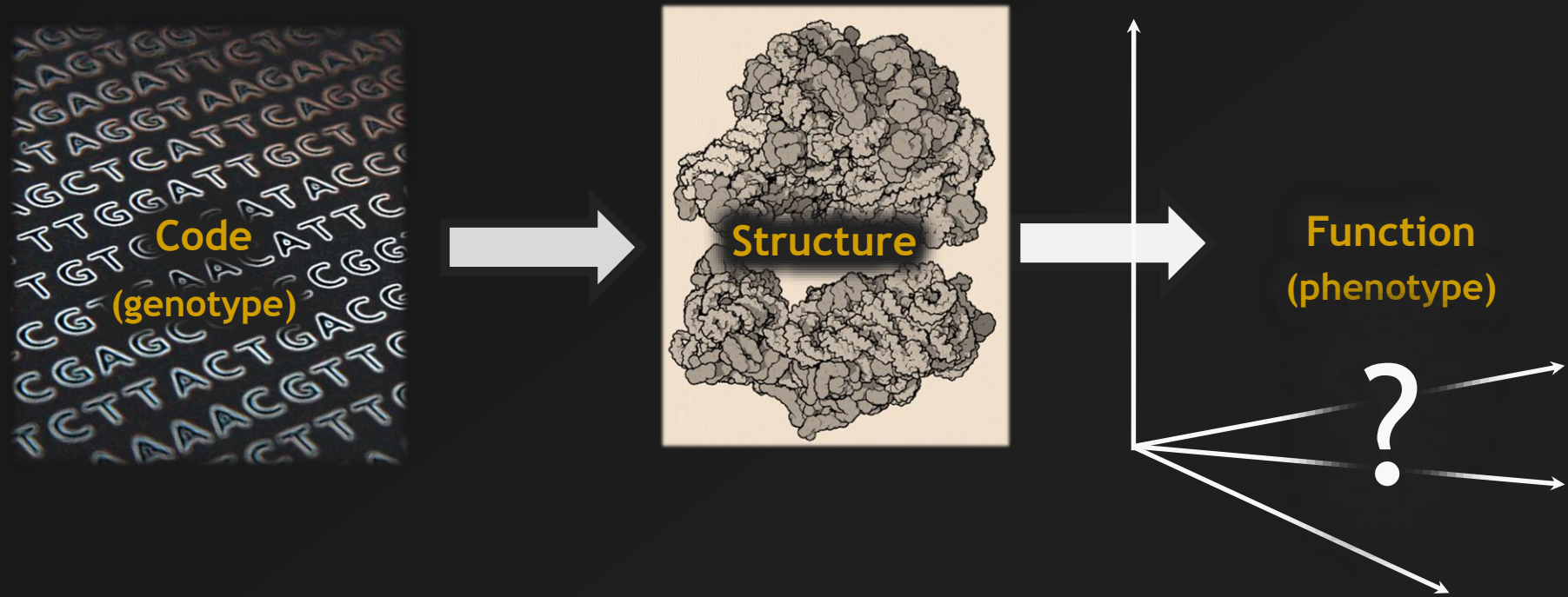


What is the dimension of molecular recognition?



Yoni Savir & TT

WAM seminar, Harvard 2015



What are the relevant degrees-of-freedom?

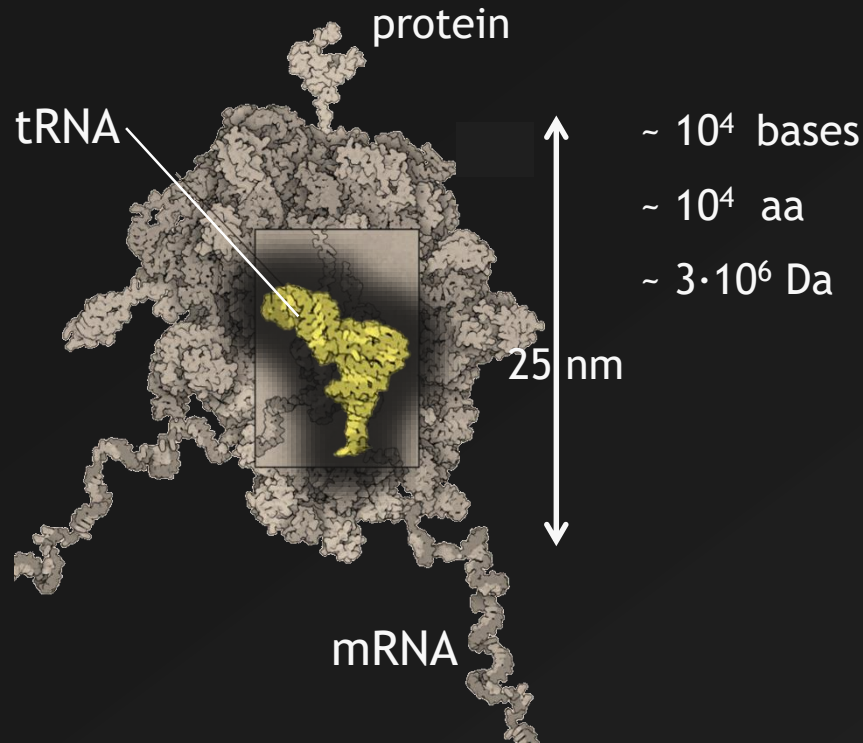
Outline: are molecular phenotype spaces “flat”?

$\text{dim}_{\text{geno}} \gg \text{dim}_{\text{pheno}}$



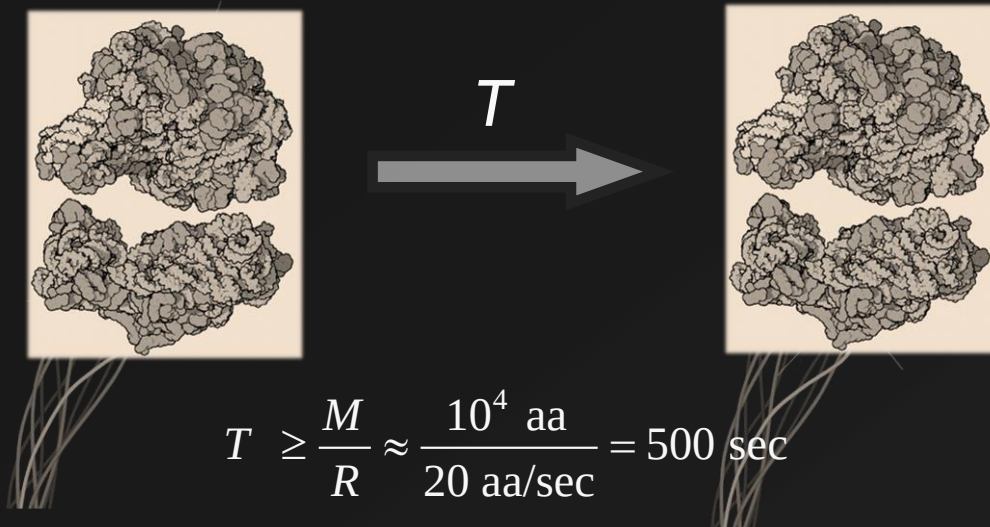
- **Molecular recognition:** “fitness”? conformational changes?
- **Generic principles:** conformational tuning (induced fit).
- **Physical basis?** ...

Ribosomes are engines of self-reproduction whose performance is critical to fitness



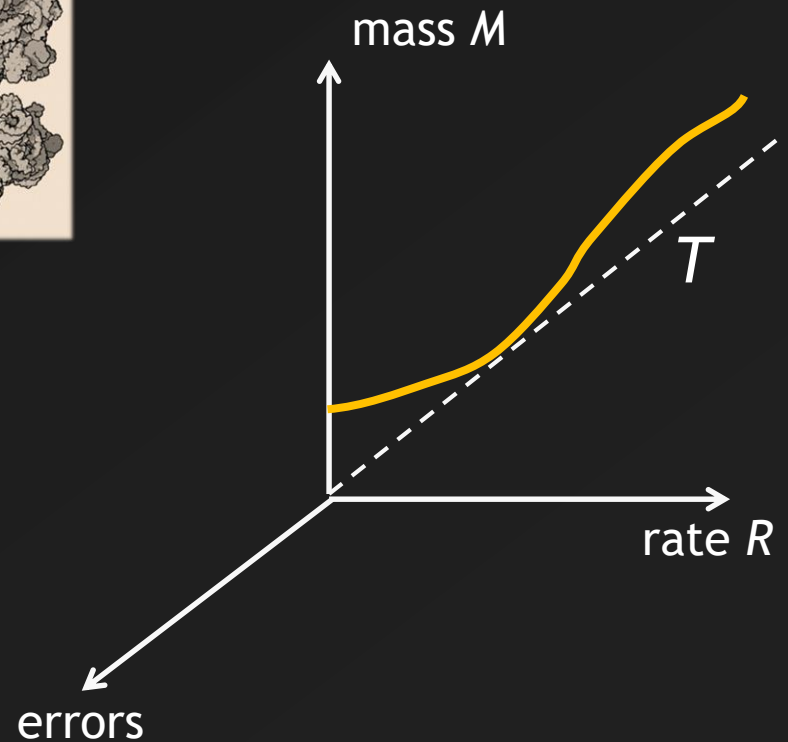
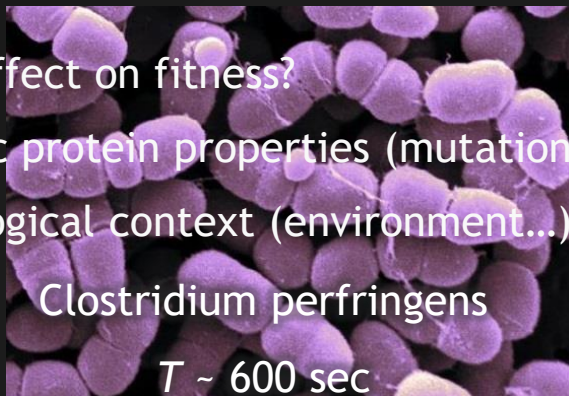
- Fast + accurate + small molecular decoder ?
- Relevant degrees of freedom ?

Ribosome limits self-replication rate

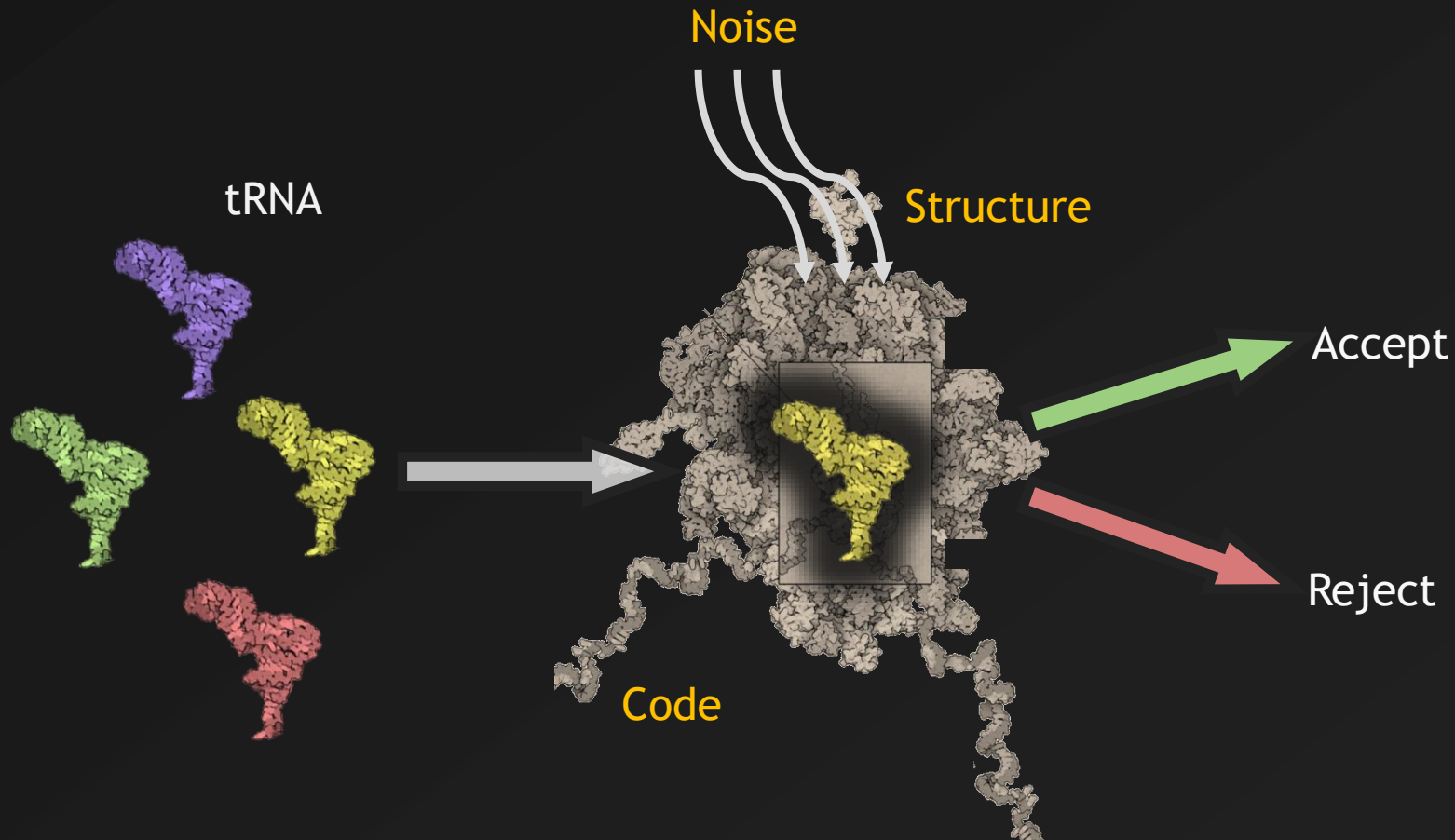


Errors effect on fitness?

- Basic protein properties (mutations...)
- Biological context (environment...)



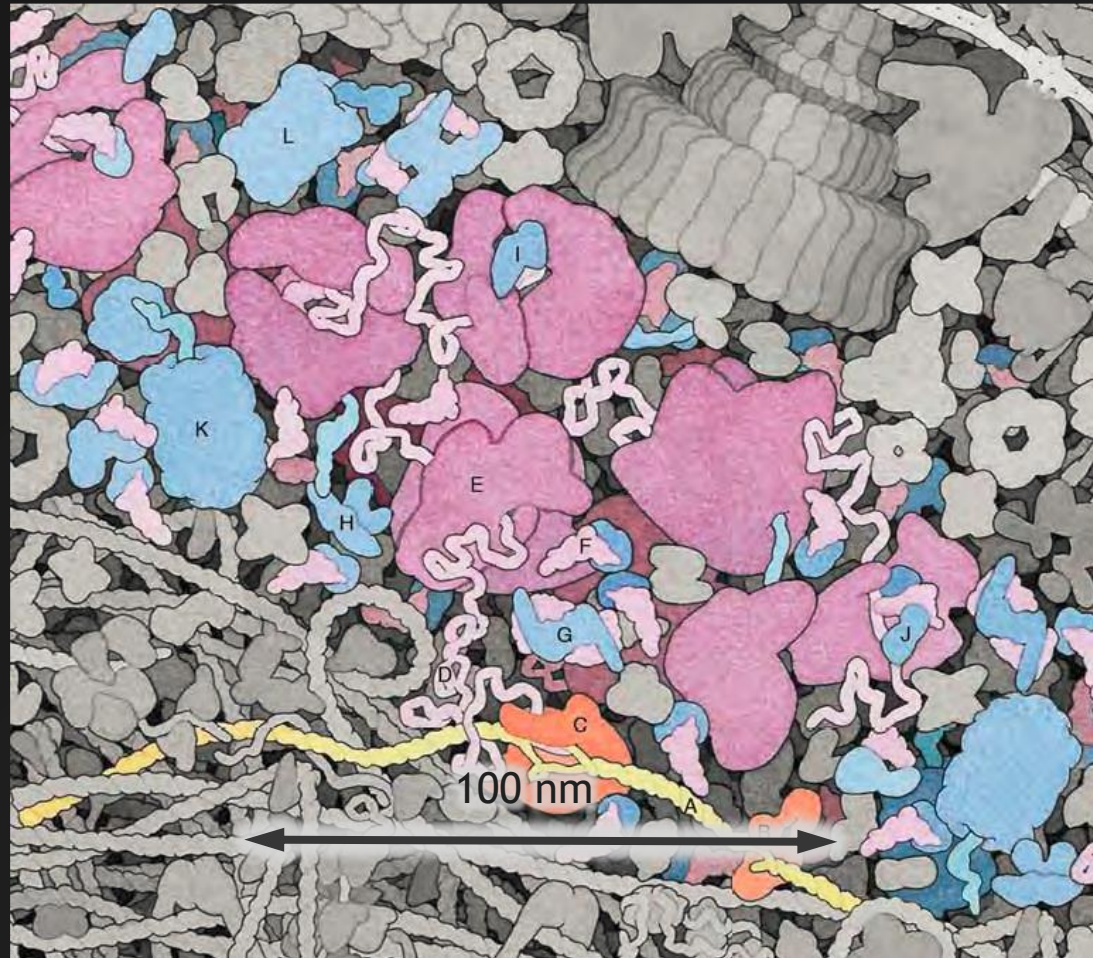
Decoding at the ribosome is a molecular decision problem



- How to evolve molecules that recognize in a noisy environment?

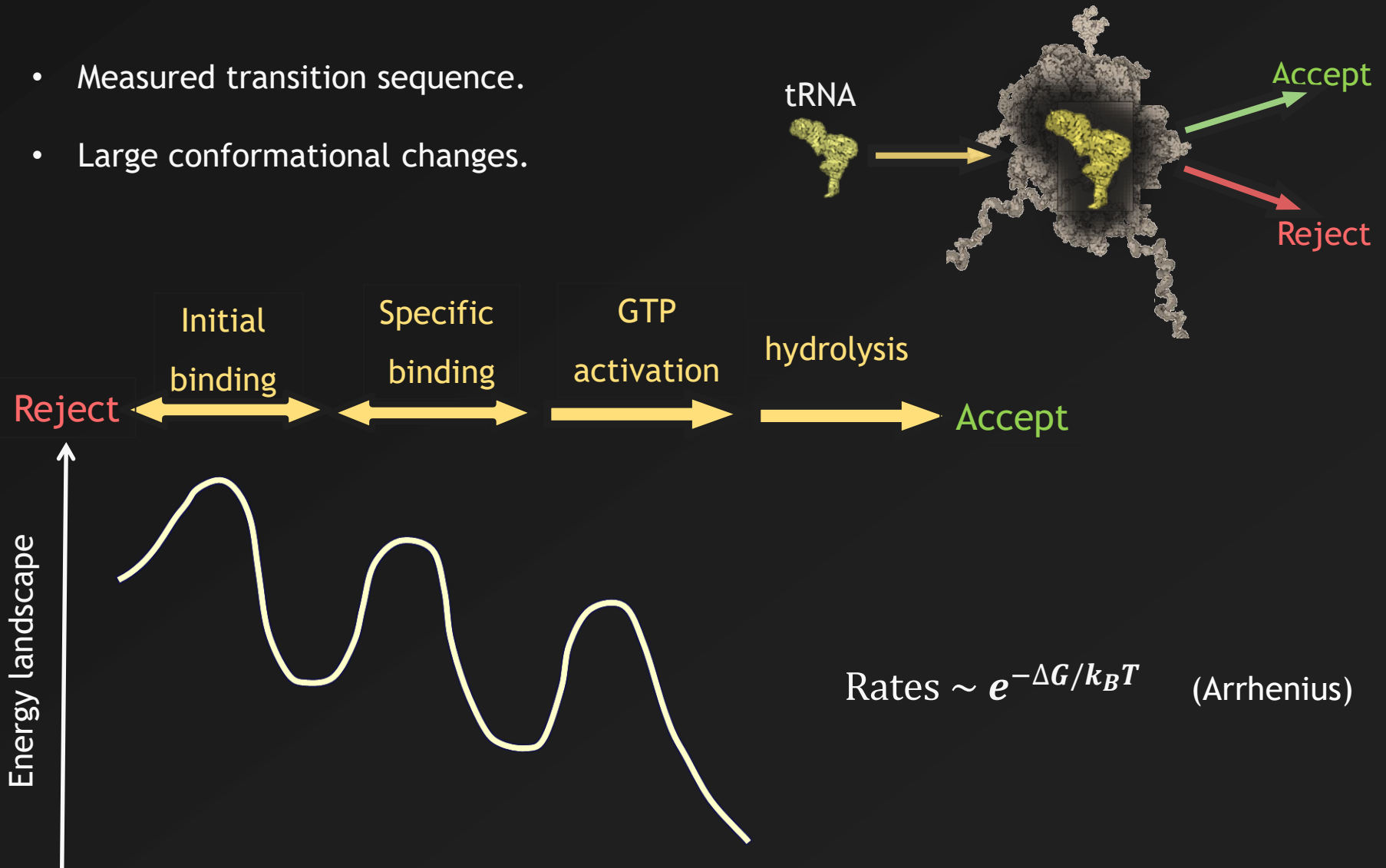
Recognizing molecular targets is a formidable task

- tRNA $\sim 10^6$
- Ribosomes $\sim 10^5$
- tRNA species = 41
- $D_{\text{tRNA}} \sim 10^2 \mu\text{m}^2/\text{sec}$
- tRNA-ribosome $\sim 10^4/\text{sec}$
- Only $\sim 1/10$ tRNAs is correct.



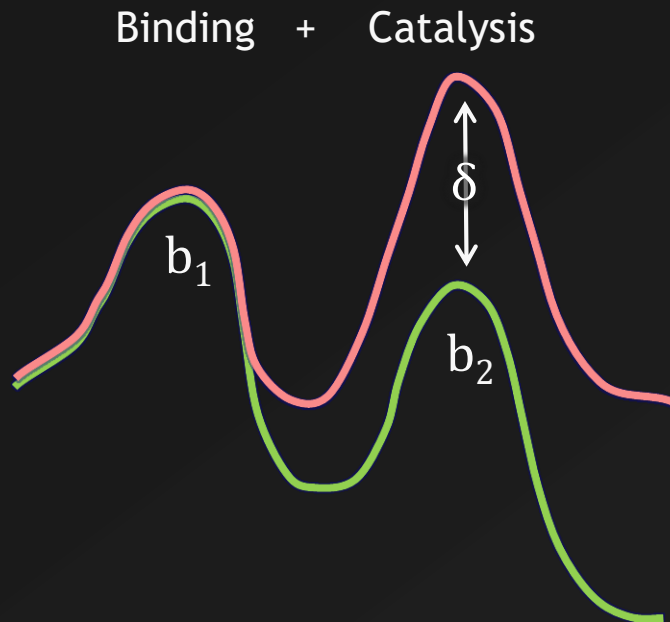
Decoding governed by energy landscapes of tRNA recognition

- Measured transition sequence.
- Large conformational changes.



Ribosome decoding problem is an example of recognition in the presence of competition

- Generic enzyme kinetics (MM):



$$R_B \sim \frac{1}{e^{b_1} + e^{b_2 + \delta}}$$

$$R_G \sim \frac{1}{e^{b_1} + e^{b_2}}$$

- The enzyme dilemma: rate R_G vs. specificity R_G/R_B ?

Given the phenomenology of tRNA decoding:

- Recognition performance (“fitness”) ?
- Effective dimension?
- Role of conformational changes?

Recognition fitness has generic features

- “Fitness” $F(R_G, R_B)$ is context-dependent.
- “Biologically reasonable”: $\frac{\partial F}{\partial R_G} \geq 0, \quad \frac{\partial F}{\partial R_B} \leq 0$.

$$R_G \sim \frac{1}{e^{b_1} + e^{b_2}}$$

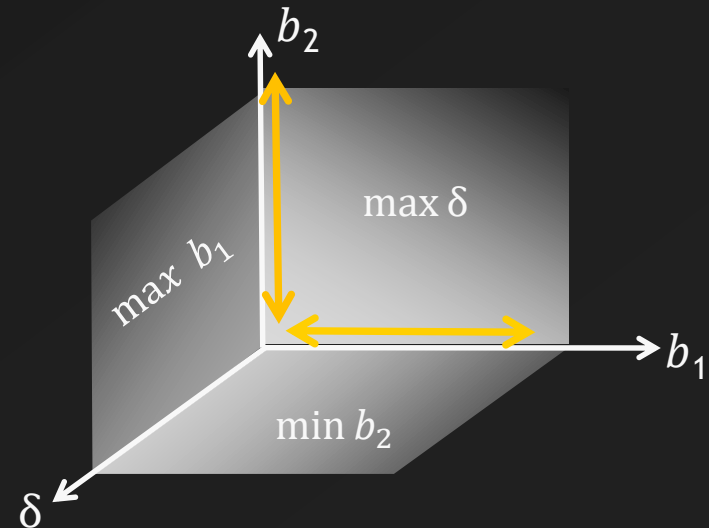
$$R_B \sim \frac{1}{e^{b_1} + e^{b_2 + \delta}}$$

- Optima in landscape (b_1, b_2, δ) space:

– $\delta \rightarrow \text{limit.}$ $\frac{\partial F}{\partial \delta} \geq 0$

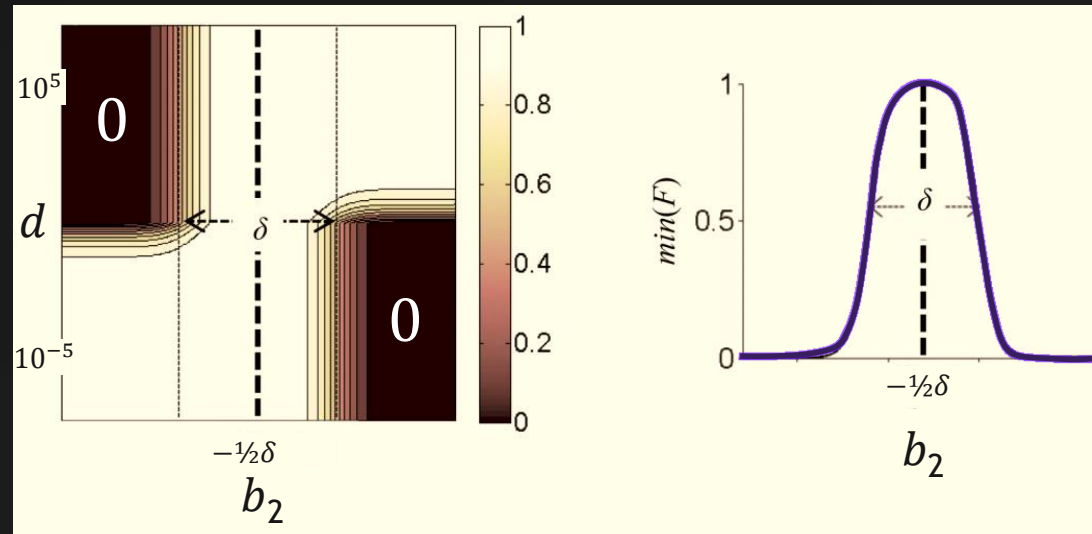
- Optimization is effectively 1D.

$$\left\{ \frac{\partial F}{\partial B} = 0 \ \& \ \frac{\partial F}{\partial b_3} \geq 0 \right\} \text{ or } \left\{ \frac{\partial F}{\partial B} \leq 0 \ \& \ \frac{\partial F}{\partial b_3} = 0 \right\}$$



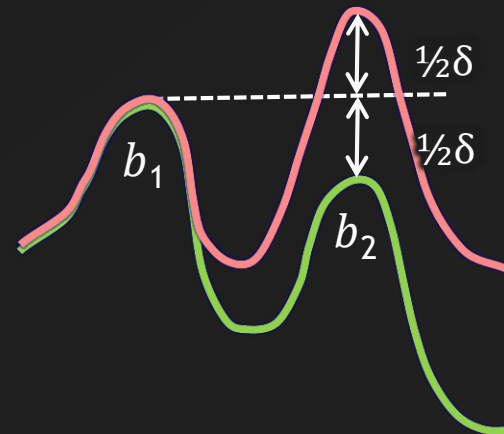
Adaptation to worst case yields symmetric design

- Example: $F = R_G - d \cdot R_B$.
- “Worst case scenario” $\max(\min(F))$.



- Max-Min solution is symmetric:

$$\Delta = b_2 - b_1 \approx -\frac{1}{2}\delta$$



Solution is valid for any “reasonable” fitness function

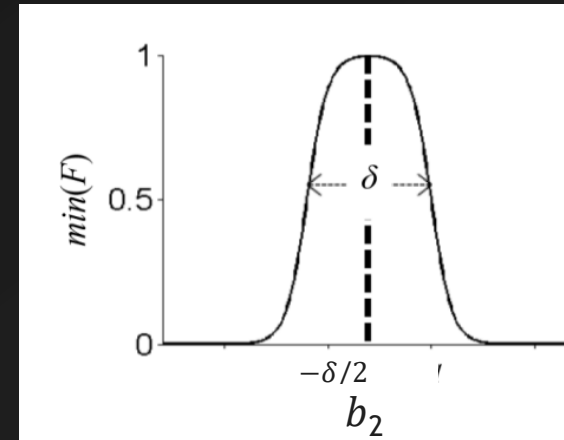
- Ribosome optimal in wide region:

$$10^{-5} = e^{-\delta} \leq d \leq e^{\delta} = 10^5.$$

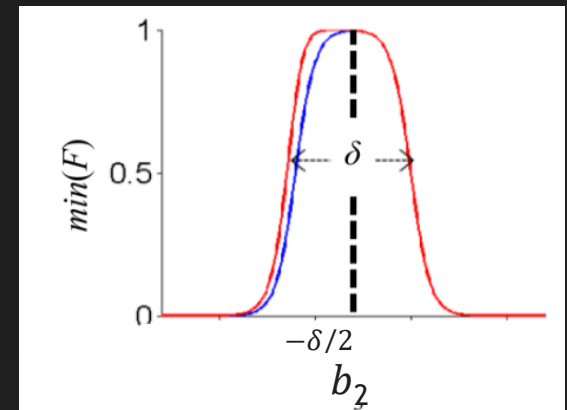
- Any $F(R_G, R_B)$ exhibits maximum as long as both rates are “relevant”.

$$e^{-\delta} < \left| \frac{\partial F}{\partial R_B} / \frac{\partial F}{\partial R_G} \right| < e^{\delta}$$

$$F = R_G - d \cdot R_B$$

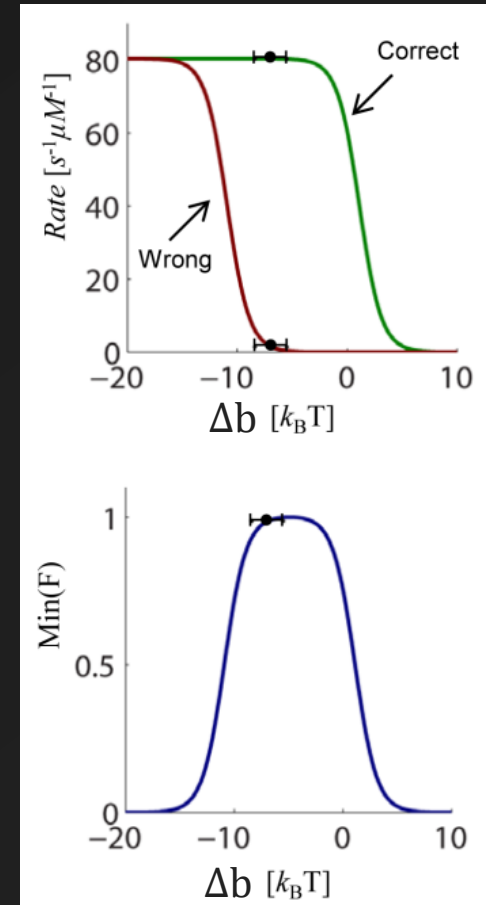
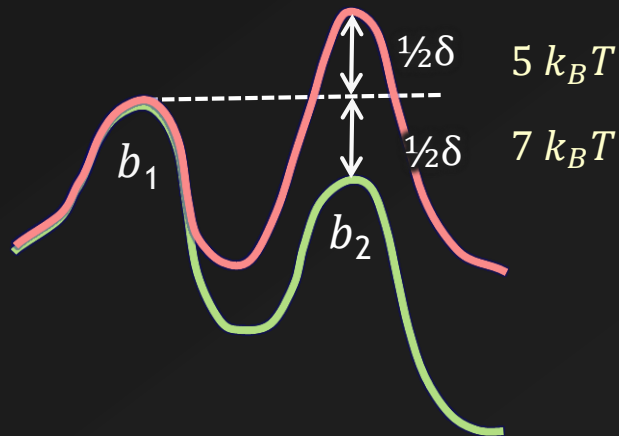


$$R_G - d \cdot R_B^2 \quad R_C - d \cdot (R_B/R_G)$$

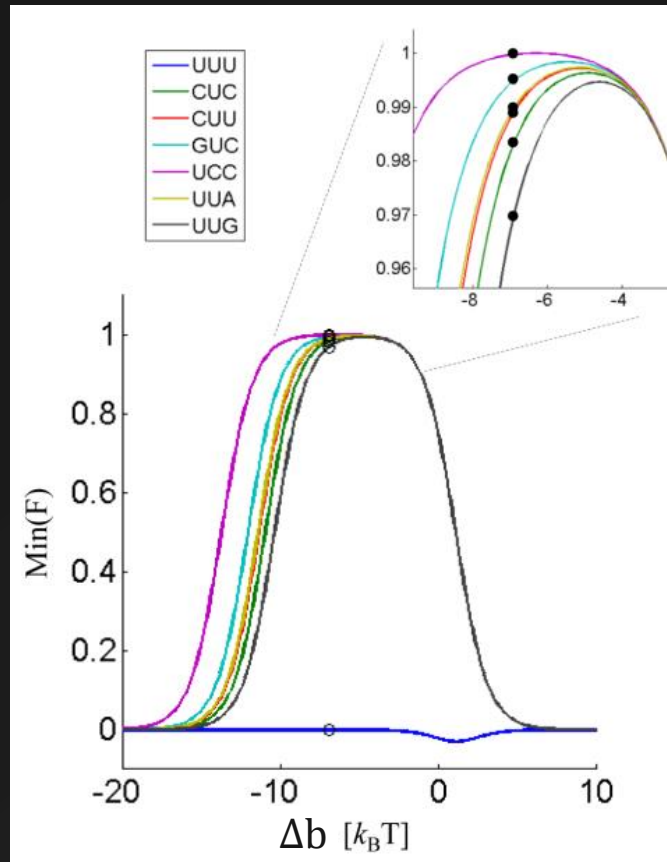


Ribosome energy barriers appear adapted

- Predicted symmetry: $\Delta b = b_2 - b_1 \approx -\frac{1}{2}\delta$.
- Measurements: $\delta \approx 12k_B T$.

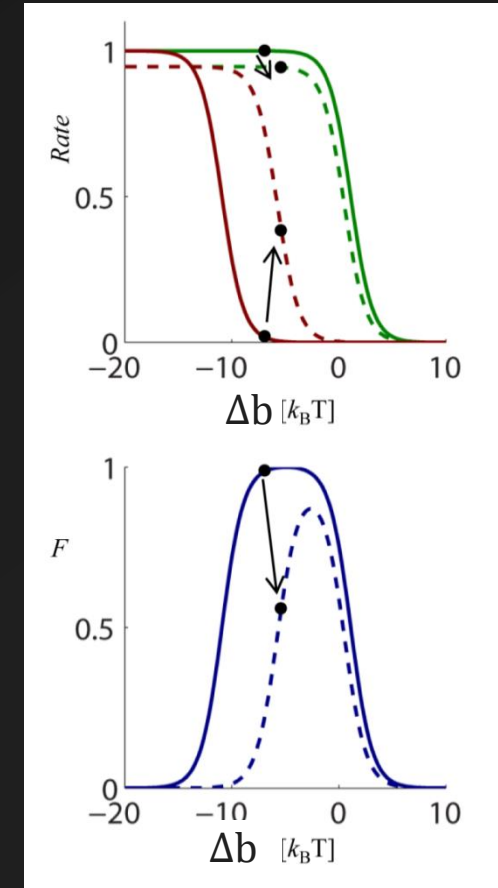
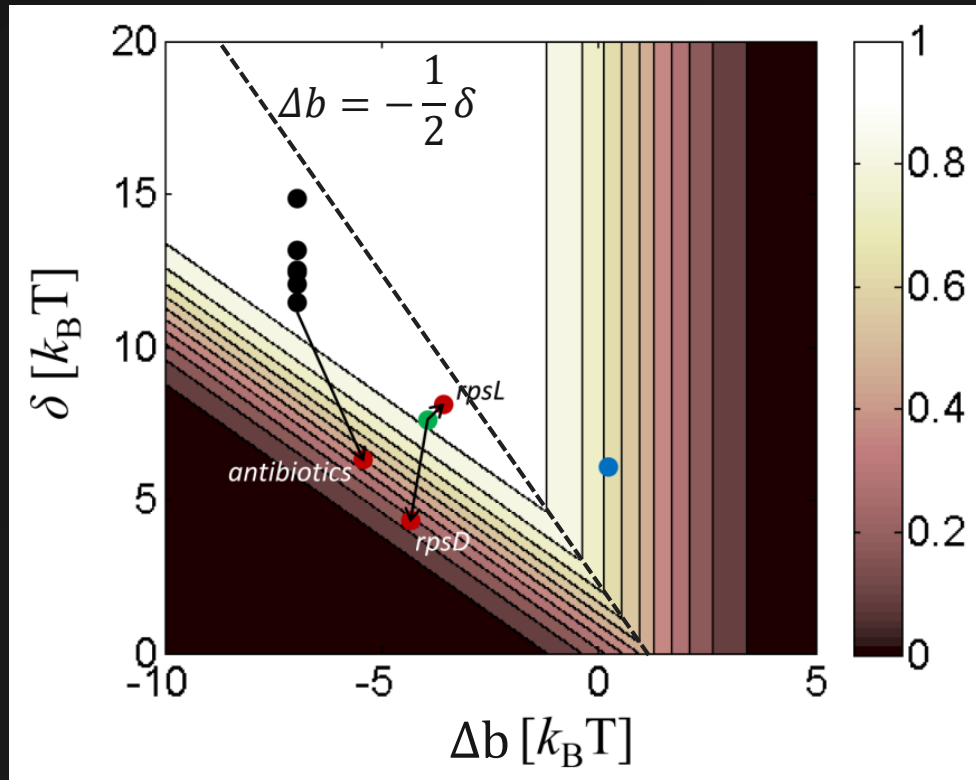


Decoding is adapted for all measured tRNAs



- Except for UUC: $\{\text{UUC}, \text{UUU}\} \rightarrow \text{phenylalanine}$.

Predicted adaptation regime of other ribosomes

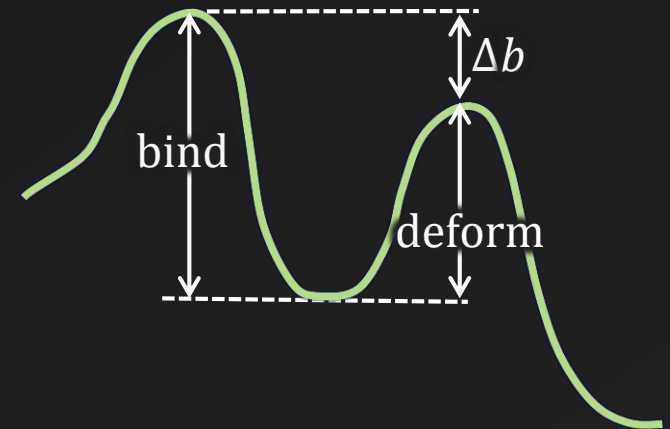


- Adaptation regime in ribosome phenotype space, $-\delta \leq \Delta b \leq 0$.
- Mutations\antibiotics reduce adaptation.

What is the role of conformational changes?

- Energy barrier $\Delta b = \text{deform} - \text{bind} = -\frac{1}{2}\delta \approx -6 k_B T$.

- For $\text{bind} \geq 6 k_B T$, $\text{deform} \geq 0$.



- **Conformational proofreading:** deformation improves tRNA recognition.
- Energy barrier that selects right targets among competitors.

Energy barriers for optimal recognition may be a general design principle of recognition systems with competition



extension energy of dsDNA.

[Dekker et al., *Mol Cell*, 2012]

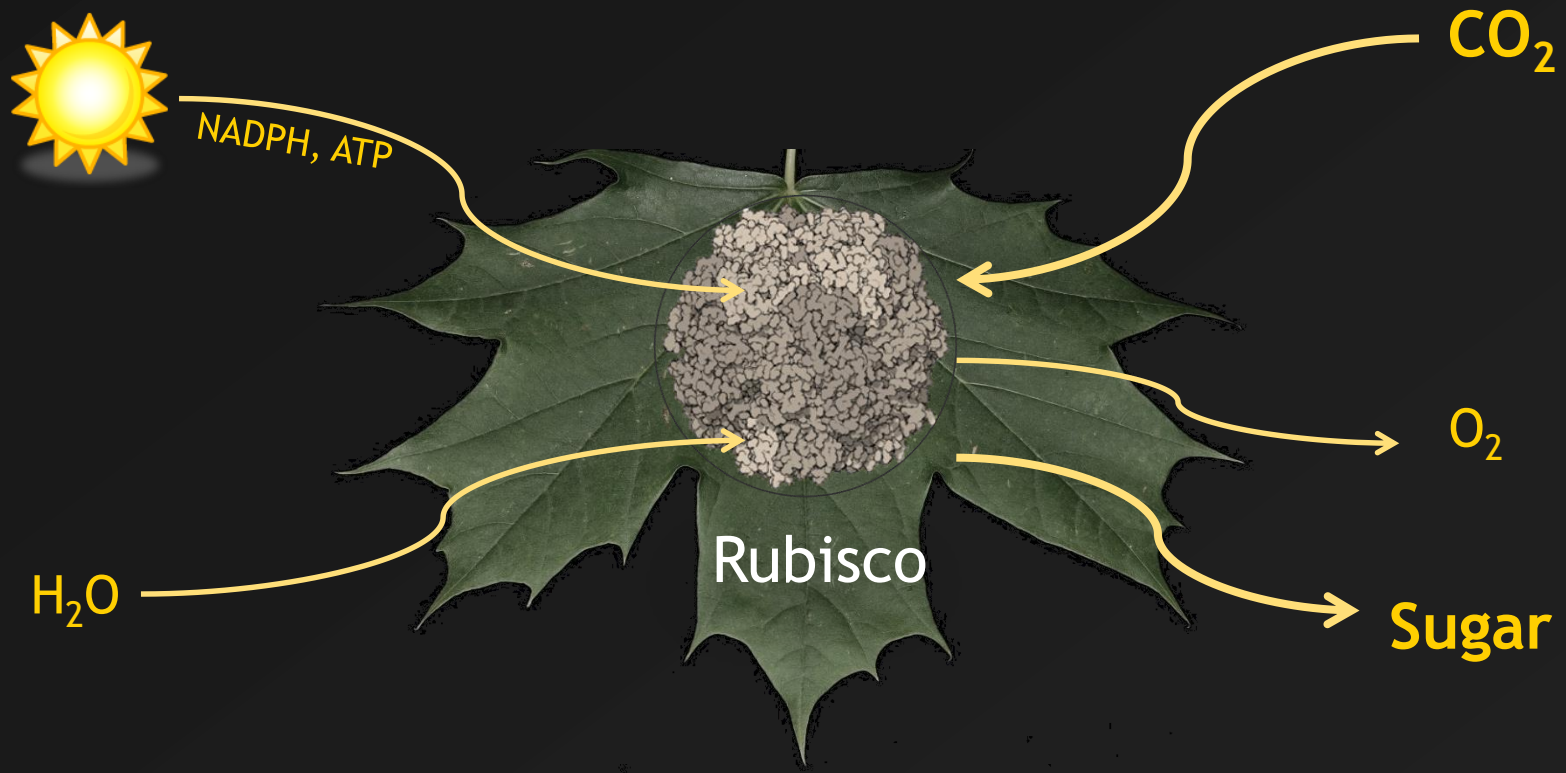
energy barriers of decoding

- Conformational proofreading:
optimization of information transfer function.
- Explains induced fit (Koshland 1958)?
- Applies to any enzymatic kinetics with competition...



- Generic properties of adaptation in function space.
- Why is coarse-grained transition-state description effective?

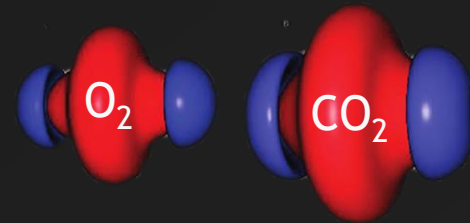
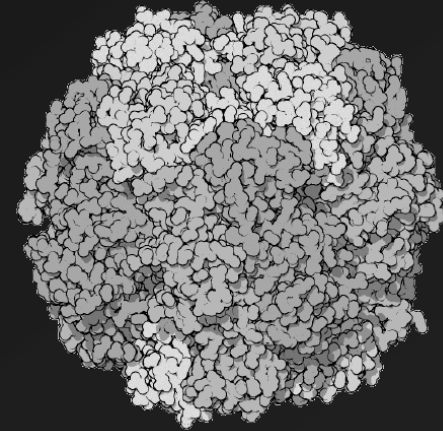
Rubisco captures atmospheric CO₂ to make sugar



- Photosynthesis fixes C into sugar.
- Rubisco captures CO₂.
- 8 large + 8 small subunits = 540 kDa.

Rubisco is thought to be a slow and confused enzyme

- #1 protein: most C fixation.
- Slow: $\sim 10 \text{ CO}_2 / \text{sec}$.
- Low specificity: $\text{O}=\text{C}=\text{O}$ vs. $\text{O}=\text{O}$.
- Can Rubisco be improved ?
- Already adapted by evolution?
- Inefficiency due to biochemical constraints?
- Is Rubisco **optimal, constrained** or both?



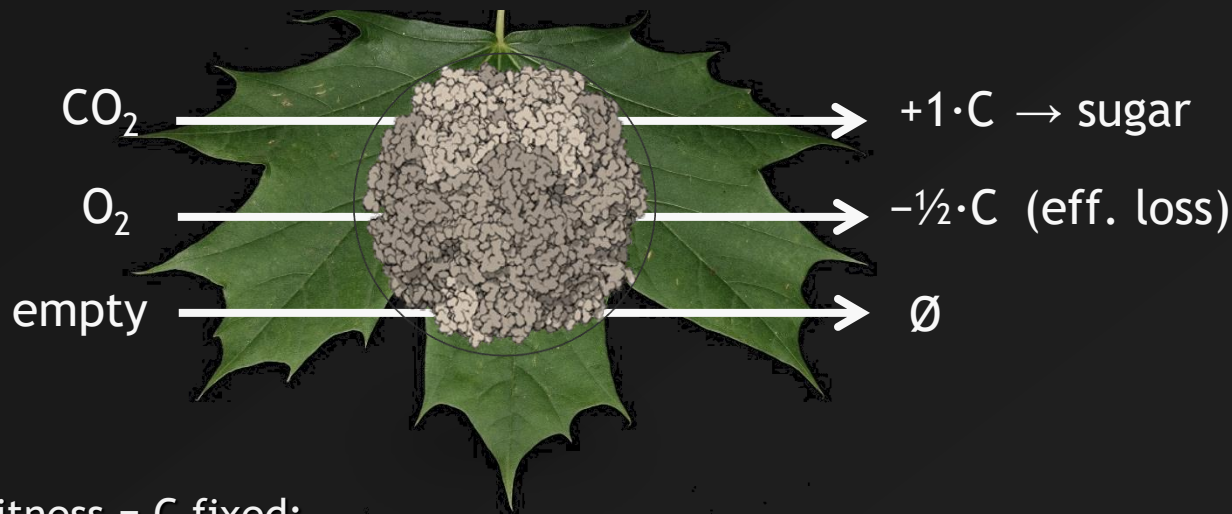
Rubisco needs to recognize CO₂ above a high background of O₂

- 2 affinities to CO₂ and O₂ : $\begin{Bmatrix} K_C & K_O \end{Bmatrix}$
+ 2 catalysis rates : $\begin{Bmatrix} v_C & v_O \end{Bmatrix}$

Atmosphere:

20% O₂

0.04% CO₂



- Fitness = C fixed:

$$F = v_C P_C - \frac{1}{2} v_O P_O + 0 \cdot P_{\text{empty}}$$

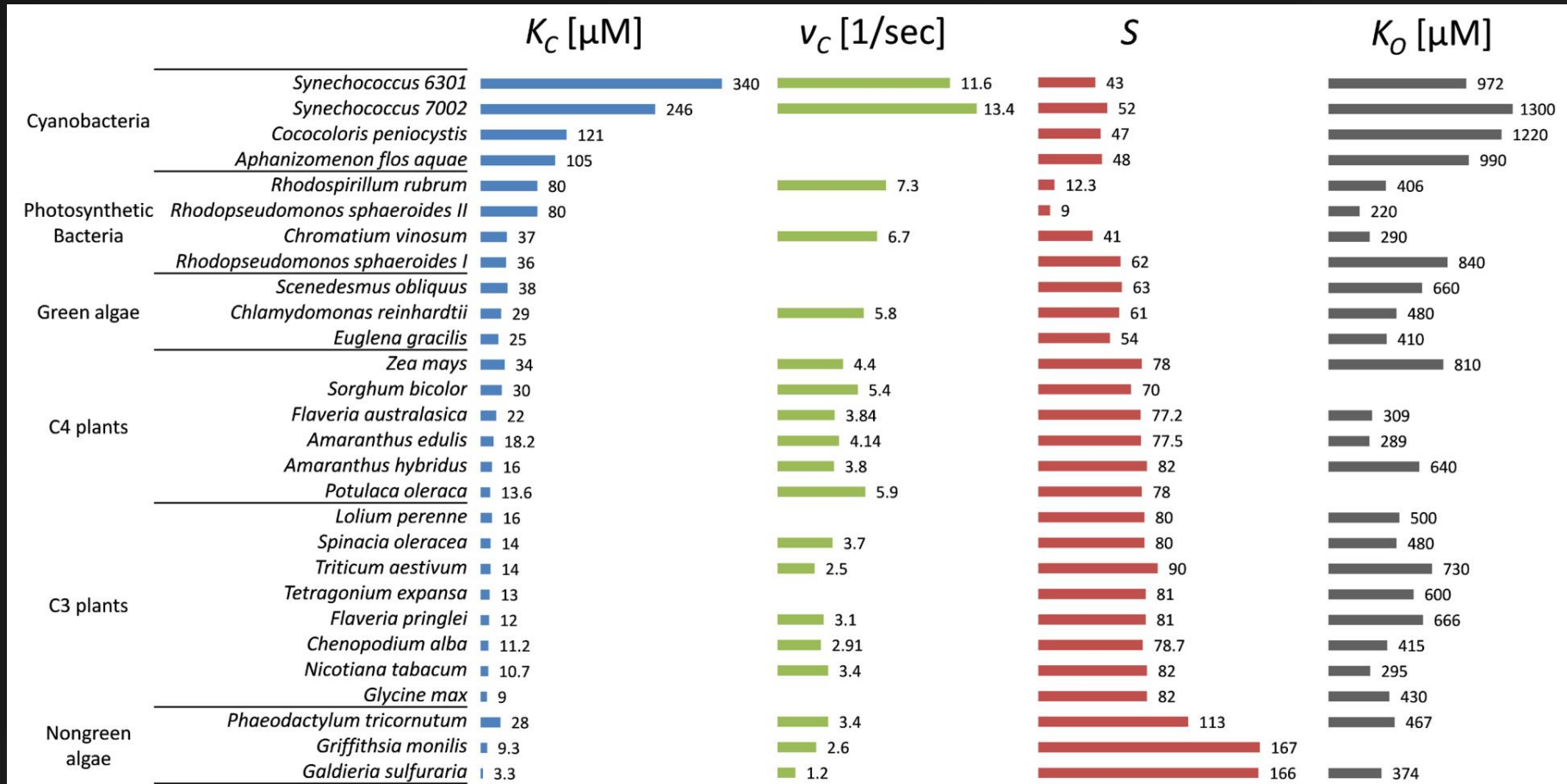
- Adapted Rubisco?

$$P_C = \frac{\frac{[CO_2]}{K_C}}{1 + \frac{[CO_2]}{K_C} + \frac{[O_2]}{K_O}}$$

$$P_O = \frac{\frac{[O_2]}{K_O}}{1 + \frac{[CO_2]}{K_C} + \frac{[O_2]}{K_O}}$$

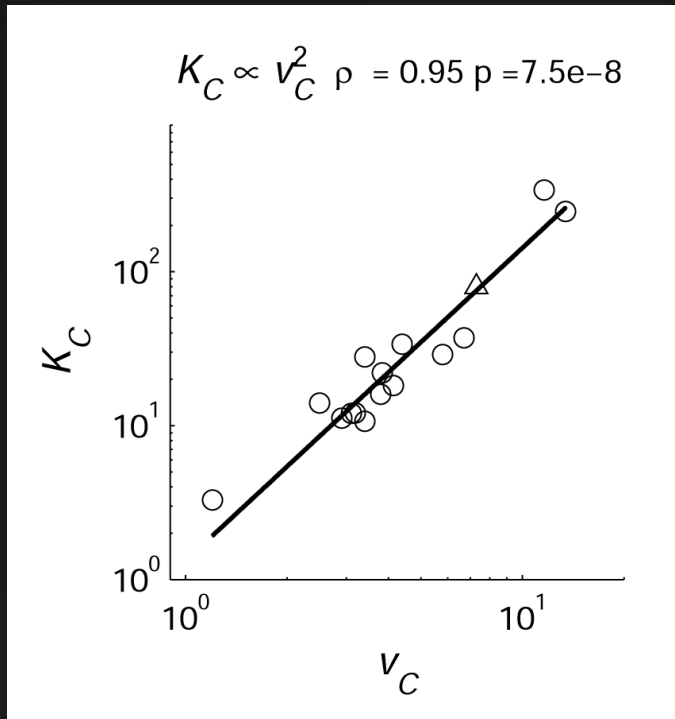
Rubisco kinetic parameters were measured in diverse organisms

- data from 28 eukaryotes and prokaryotes:



Rubisco's parameter space is effectively 1D

- 4D data space (S , v_C , K_C , K_O).
- constrained to 1D power law.



$$K_C \propto v_C^{2.0 \pm 0.2}$$

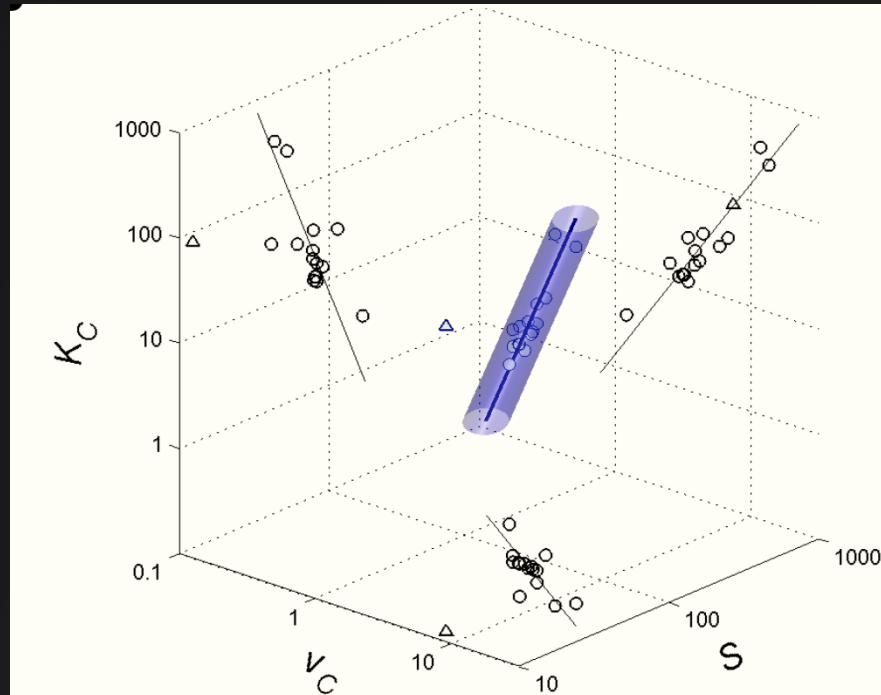
$$K_O \propto v_C^{0.5 \pm 0.1}$$

$$S \propto v_C^{-0.5 \pm 0.1}$$

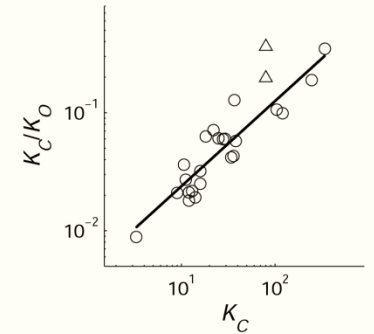
Dimensional reduction
("Constrained Plasticity")

Effective 1D landscape for Rubisco evolution

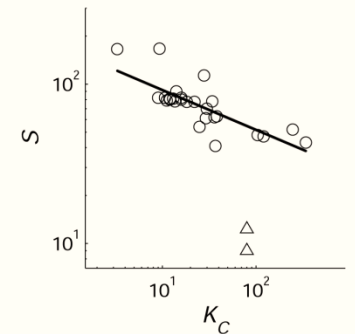
- All organisms scattered around a straight line in 4D parameter space.



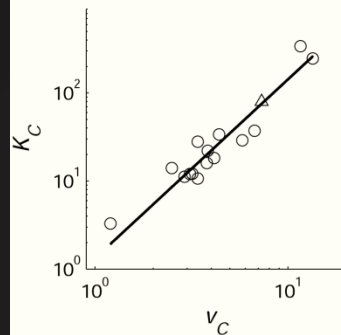
$$K_C/K_O \propto K_C^{0.28} \quad \rho = 0.92 \quad p = 6.3e-10$$



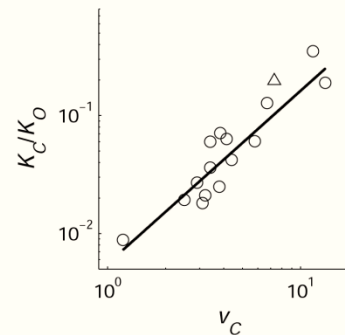
$$S \propto K_C^{-0.25} \quad \rho = 0.78 \quad p = 6.9e-4$$



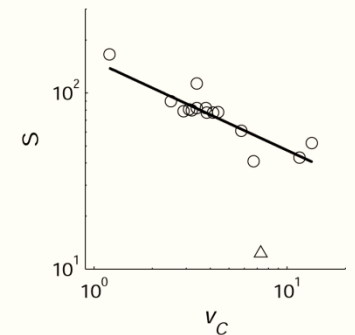
$$K_C \propto v_C^2 \quad \rho = 0.95 \quad p = 7.5e-8$$



$$K_C/K_O \propto v_C^{1.5} \quad \rho = 0.92 \quad p = 8.7e-7$$

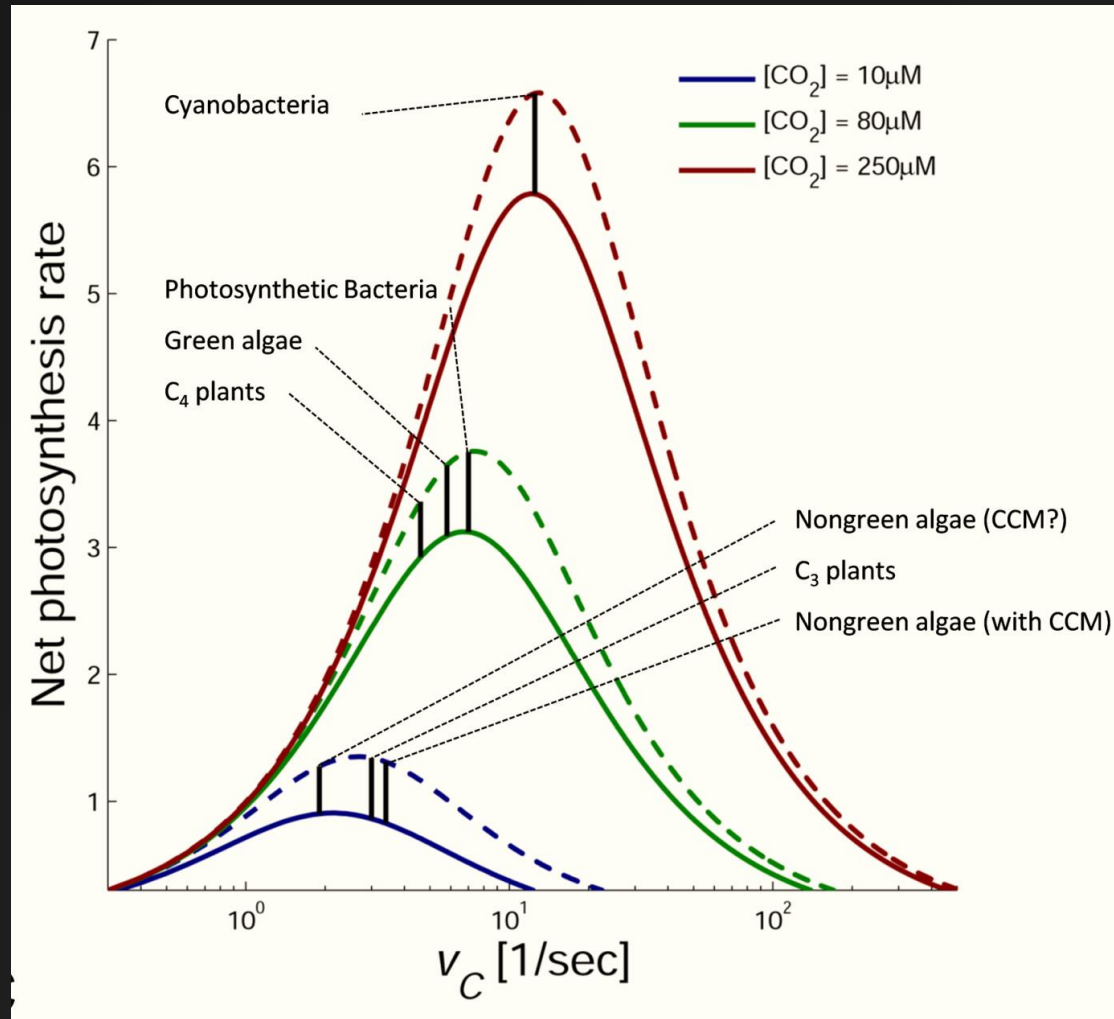


$$S \propto v_C^{-0.51} \quad \rho = 0.89 \quad p = 7.4e-6$$



Rubisco energy barriers are nearly optimal for recognition of CO₂ in their habitats

- Max(F) along power-laws in given habitat (O₂ + CO₂).

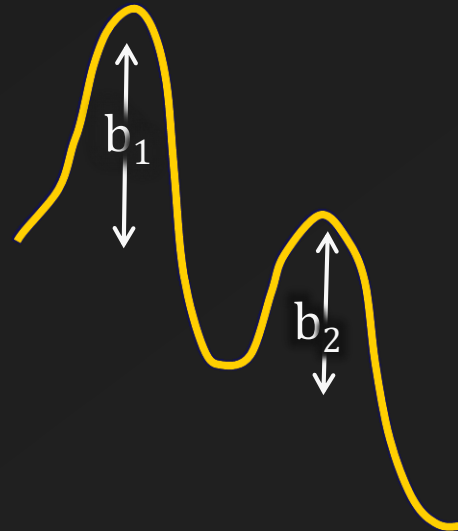


$$F(v_C) = \frac{v_C}{1 + 1.3 \cdot v_C^2 / [\text{CO}_2]} + f([\text{O}_2])$$



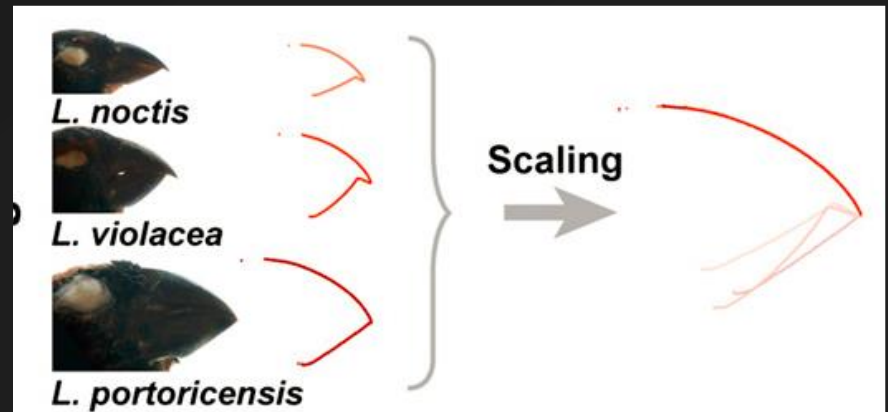
- Linear energy constraints?

$$K_C \sim v_C^2 \quad \leftrightarrow \quad b_1 + b_2 = \text{const.}$$



- Molecular phenotype spaces appear low-dimensional,
ribosome, RecA, rubisco... - **is it generic?**
- Simple, coarse-grained thinking about function may explain basic features.
- Conformational changes enhance recognition.
- **Dimensional reduction** in other living systems (development, behavior, neural...)

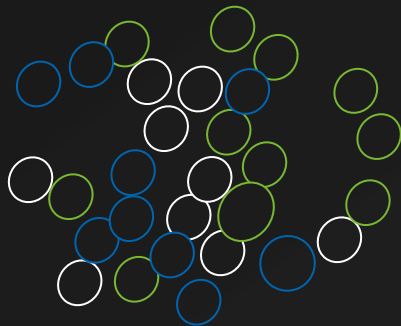
- General feature of mapping
genotype $\rightarrow$ phenotype?



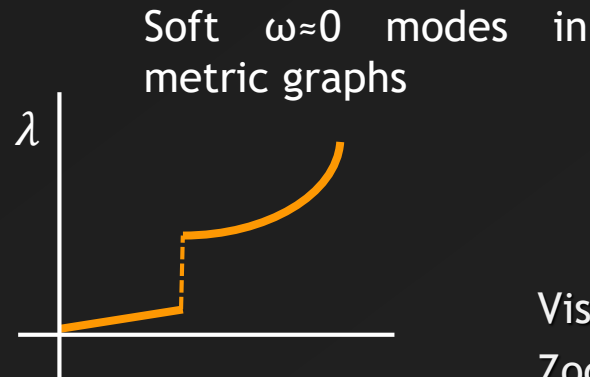
Mallarino, Campàs, Fritz, Burns, Weeks,
Brenner, Abzhanov (PNAS 2012)

What is the relation to physical properties of proteins as an evolvable state of matter?

- **Slow manifolds of low dimension** in various living systems.
- Hints: - transition states approximate kinetics.
 - Low modes dominate protein function (spectral gap?)
- **What are the biophysical degrees-of-freedom of *evolvable*, *information-rich soft matter*.** (work in progress...)



Soft modes in amorphous solids



Viscous response $\omega \leq 10^2$ Hz
Zocchi et al. (2012)

Thanks:

Yonatan Savir

Elad Noor, Ron Milo

