


going back to p(44) simpler recursion relations with


matchas available (cents)

$$\text{eff} = (i, j) = \alpha_{j-1} \text{eff}(i, j+1)l + \sum_{k \leq j} (\text{eff } z^{\text{FP}}(k, j)) l_{\text{BP}} + z^{\text{BP}}(i, j) \textcircled{Q_{\text{BP}}(l)}$$

Can we identify  $\mathcal{L}_{BP} \Rightarrow \overline{ab}$   
 $\mathcal{L} \Rightarrow \overline{bc}$   
 $\mathcal{L}_{INT} \Rightarrow \overline{c}$  } i. multiloop rule

  $\rightarrow z = \mathcal{L}_{INT} \mathcal{L}_{BP}^3 / K_a^{BP^*}$

~~cut~~  
 $\Delta f = a + b(\# \text{ cuts})$   
 $+ c(\# \text{ poles})?$


 $\rightarrow Z = \frac{(5+1) \times 2 \times \frac{1}{2} \times \frac{1}{2}}{1 \times 2}$

$$\text{define } l_{\text{eff}} = (l_{\text{ex}}/L)$$
$$l_{\alpha\beta} = l'_{\alpha\beta} \rightarrow$$

try on  
next page

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$$l_{BP} = \frac{Q}{l_{BP}^*} = \frac{l_{BP}^*}{l_{BP}}$$

Further revisiting of "simplified" recursion:  $l_{BP} = l_{BP}^*$

$$Z^{BP}(i,j) = \alpha_i \alpha_j \frac{C_{off}(i+1,j-1) l_{BP}^2}{l_{BP}^*} + \sum_{\text{outpoint}} \frac{C_{sta}}{l_{BP}^*} Z_{\text{linea}}(i,k) Z_{\text{linea}}(k,j)$$

$$C_{off}(i,j) = \alpha_{j-1} C_{off}(i,j-1) Q + C_{int}^{BP}(i,j) l_{BP}^* + \sum_{i < k < j} C_{off}(i,k-1) l_{BP}^* Z^{BP}(k,j)$$

$$Z_{\text{linea}}(i,j) = \alpha_{j-1} Z_{\text{linea}}(i,j-1) + Z^{BP}(i,j) + \sum_{i < k < j} Z_{\text{linea}}(i,k-1) Z_{BP}(k,j)$$

Init size:

$$Z_{\text{linea}}(i,i) = 1$$

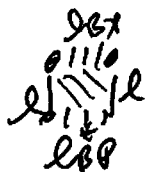
$$C_{off}(i,i) = C_{int}$$

Finalize

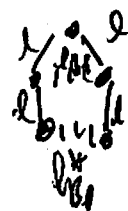
$$Z_{\text{full}}(i,i) = \begin{cases} C_{off}(i,i-1) l_{BP}^* + \sum_{i \leq c \leq i+1} Z_{\text{linea}}(i,c) Z_{\text{linea}}(c,i-1) & \text{if } i-1 \dots i \text{ connected.} \\ Z_{\text{linea}}(i,i-1) l_{BP}^* & \text{if } i-1 \dots i \text{ outpoint} \end{cases}$$



$$Z = \frac{C_{int} l_{BP}^4}{l_{BP}^*} = C_{int} \frac{l_{BP}^4}{l_{BP}^*}$$



$$Z = \left( \frac{C_{sta} l_{BP}}{l_{BP}^*} \right) \times \frac{l_{BP}^{C_{int}}}{l_{BP}^*} = \frac{C_{int} l_{BP}^2 C_{int}}{l_{BP}^* l_{BP}^*}$$



$$Z = C_{int} \left( \frac{l_{BP}}{l_{BP}^*} \right)^4 = \frac{C_{int} l_{BP}^4}{l_{BP}^{*2}} = \left( \frac{C_{int} l_{BP}}{l_{BP}^*} \right) \left( \frac{C_{int} l_{BP}}{l_{BP}^*} \right)$$

Wich makes to full from to full

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How to add outside state [rough]

new state

$$Z^{Coax}(i, j) = \sum_{i < k < j} Z_{BP}(i, k-1) Z_{BP}(k, j) K_{Coax}$$

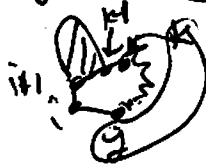
~~BP~~  
k-1 k

$$C_{eff}(i, j) += Z^{Coax}(i, j) L_{Coax} + \sum_{i < k < j} C_{eff}(i, k-1) L_{Coax} Z^{Coax}(k, j) L_{Coax}$$

$$Z^{linear}(i, j) += Z^{Coax}(i, j) + \sum_{i < k < j} Z^{linear}(i, k-1) Z^{Coax}(k, j)$$

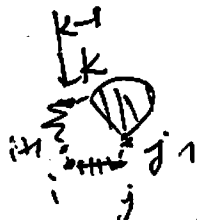
?

$$Z^{BP}(i, j) += \sum_{i < k < j} Z^{BP}(i, k-1) C_{eff}(k, j) L_{BP}^*$$



$k < j-1$  [or current]

$$+ \sum_{i < k < j} L_{Coax} C_{eff}(i, k-1) L_{BP}^* Z^{BP}(k, j-1) L_{Coax}$$



$k > i+1$  [or current]

write clearly on next page.

not

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If no constraints - COAX  $\Leftrightarrow$  BP, i.e.  $\downarrow$   
 bulgeless seq allowed,  $\rightarrow$

$$Z^{COAX}(i,j) = \sum_{i < k, k-1} Z_{BP}(i, k-1) Z_{BP}(k, j) + \sum_{i < c < k-1} \frac{C_{sta}}{K_d} Z_{BP}(i, k-1) Z_{BP}(k, j)$$

$Z > 1$  Bound

Supplement expression - PSD  
 $Z_{BP}(i,j) +=$

Cut?

$$Z_{BP}(i, j) \leftarrow Z_{BP}(i, j) + \frac{Z_{BP}(i, k) Z_{BP}(k, j)}{K_{BP}^{COAX}}$$

$$\sum_{i < k < j} C_{eff}(i, k-1) Z_{BP}(k, j)$$

$i \rightarrow j-1$   
 $j$

$$+ \sum_{i < k < j} [Z_{BP}(i, k) Z_{BP}(k, j)]$$

$k$   
 $i \rightarrow j$

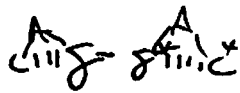
$$+ \sum_{c = \text{cutpoint}} \frac{C_{sta}}{K_{BP}^{COAX}} \left\{ Z_{BP}(i, c) Z_{linear}(c, j-1) + Z_{linear}(i, c) Z_{BP}(c+1, j) \right\} K_{COAX}$$

~~$C_{eff}(i,j) = \dots$~~

$$+= C_{Int} Z^{COAX}(i,j) L_{COAX} + \sum_{i < k < j} C_{eff}(i, k-1) L_{COAX} Z^{COAX}(k, j) L_{COAX}$$

Need to check through some examples

$$Z_{linear}(i,j) = Z^{COAX}(i,j) + \sum_{i < k < j} Z_{linear}(i, k-1) \times Z^{COAX}(k, j)$$



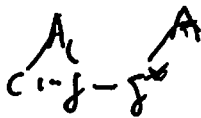
(normal  $c \rightarrow s$ ,  $s \rightarrow c$  for now)

C  
 $z_{BP} \rightarrow 0$   
 $z_{max} \rightarrow 0$   
 $C_{eff} = C_{int}$   
 $z_{linear} = 1$

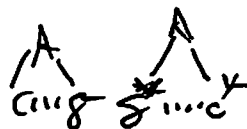
CA  
 $z_{BP} \rightarrow 0$   
 $z_{max} \rightarrow 0$   
 $C_{eff} = C_{int}$   
 $z_{linear} = 1$

CA  
 $z_{BP} = \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$   
 $z_{max} \rightarrow 0$   
 $C_{eff} = C_{int}^2 + \frac{C_{int}^2 l_{BP}^{2k}}{k_{BP}^k}$   
 $z_{linear} = 1 + \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$

CA  
 $z_{BP} \rightarrow 0$   
 $z_{max} \rightarrow 0$   
 $C_{eff} = C_{int} + \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$   
 $z_{linear} = 1 + \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$



CA  
 $z_{BP} \rightarrow 0$   
 $z_{max} \rightarrow 0$   
 $C_{eff} = C_{int} + \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$   
 $z_{linear} = 1 + \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$



CA  
 $z_{BP} \rightarrow 0$   
 $z_{max} = \left[ \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k} \right]^2 \frac{l_{BP}^{2k}}{k_{BP}^k} k_{BP}^{2k}$   
 $C_{eff} = C_{int}^5 + \frac{C_{int}^2 l_{BP}^{2k}}{k_{BP}^k} +$   
 $\left[ C_{int}^2 + \frac{C_{int}^2 l_{BP}^{2k}}{k_{BP}^k} \right] \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$   
 $+ C_{int} l_{BP} \left[ \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k} \right]^2 \frac{l_{BP}^{2k}}{k_{BP}^k}$

~~$z_{linear} = 1$~~   
 $z_{linear} = 1 + \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k} +$   
 $\left( 1 + \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k} \right) \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k}$   
 $+ \left[ \frac{C_{int} l_{BP}^{2k}}{k_{BP}^k} \right]^2 \frac{l_{BP}^{2k}}{k_{BP}^k} k_{BP}^{2k}$

Need to calculate  
 $z_{final}$  5 ways

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Return to motifs.

Same Recursion as on p-53, but add motif term to ~~z~~  $z^{OP}$ .

$$z^{BP}(i,j) = \alpha_i \alpha_j \frac{C_{off}(n, j-1) z^{OP}}{K_{BP}} + \sum_{c = \text{cutpoint}} \frac{C_{std}}{K_{BP}} z_{linear}(i, k_c) z_{linear}(k_c, j) + \alpha_i \alpha_j \frac{C_{off}}{K_{BP}} \frac{\text{motif}(k_l, j, i)}{K_{BP}} z^{OP}(k_l)$$

*(with arrows pointing from C<sub>off</sub> to C<sub>std</sub> and from motif to C<sub>off</sub>)*

pretty simple.

or WAIT:

$$z_{(i,j)}^{Final} = \lim \left[ \frac{C_{off}^{(i,j)}}{C_{std}} + \sum_{i < j < (i-1) + N} z_{linear}(i, j) z_{linear}(j, i) \right] + (1 - \alpha_i) [z_{linear}(i, i-1)]$$

[see p-52 for definition]

+  $\sum_{i < j < (i-1) + N} \alpha_j z_{BP}(i, j) z_{BP}(j, i)$  *(with arrows pointing from z<sub>linear</sub> to z<sub>BP</sub>)*

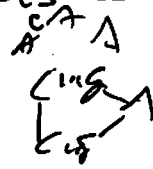
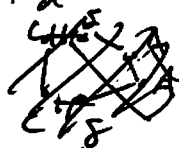
Let's ask a question...

Questions in inference

- fix  $d = d_{kp} = 1$

$$G_{int}/k_a^{op} = \exp[-4.1 \text{ kcal/mol / } kT] ?$$

- Simulate a bunch of  $K_d$  measurements to a motif like



as "msl" Rainier

- what's the "spectrum" of  $K_d$ ?

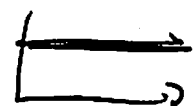
TEST 1 - if we have a bunch of measurements of  $\log(K_d)$  with 1 kcal/mol accuracy, could we infer  $\log(G_{int}/k_a^{op})$ ?



TEST 2 - then add in  $C_{off}$  for

back pair step (assume constant)

Same question is 2D XAN



TEST 3 - Suppose we could measure  $\Delta F$  for a bunch of designed  $\Delta F$  constructs - infer  $\log(G_{int}/k_a^{op})$ ?



TEST 4 - Suppose we could design a bunch of sequences with one ensemble defect, what if we optimize  $\sum_{\text{sequences}} \Delta F(\text{MFC}) - \Delta F(\text{all structures})$  target structure

TEST 5 - what if we simulate "chemical mapping" profiles could we make inference?

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- Can we do de novo inference?

① New kind of base pairs that build off w/c?

$\left. \begin{matrix} A-U \\ G-C \\ G-U \end{matrix} \right\} \Rightarrow$  add:  $G-A$  w/c  
 $4 \times 3 = 12$  <sup>non</sup> Ceff param  
 $+ K_{dA}^{-1} = 1$  Ceff parameters

Could scan all  $4 \times 4 = 16$  variants

② Look for UMR turns  $\rightarrow$   $\left\{ \begin{matrix} K_{UCA} \\ L_{UCA} \end{matrix} \right.$  treat similar to "BP".  
In fact, could scan all 64 variants  
Early Faults

Need primitives to constrain searches,  
either naive ones, es,  $l_{BP} \sim l_{GC}$   
or sophisticated ones for es.

Ceff  $\left( \begin{matrix} S_{UC} \\ S_{UA} \end{matrix} \right)$ ,  
from Picta.



pre - (5.7) A

②



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$$\left( \frac{P_{\text{OVS}}}{P_{\text{FFC}}} \right) \approx \left[ \frac{C_{\text{INT}}}{C_{\text{OVS}}} \right] \frac{k_{\text{OVS}}}{k_{\text{FFC}}}$$







A circuit diagram of a common-emitter amplifier. It features a base resistor \$R\_{B1}\$ connected to a voltage divider consisting of two resistors in series between \$V\_{CC}\$ and ground. The emitter is connected to ground through a resistor \$R\_E\$. The collector is connected to \$V\_{CC}\$ through a load resistor \$R\_L\$ and a feedback resistor \$R\_F\$. A capacitor \$C\_{out}\$ is connected from the collector to ground.

Diagram illustrating the calculation of the effective impedance  $Z_{eff}$  for a transmission line with a load  $Z_L$  and a series impedance  $Z_s$ .

The diagram shows a transmission line with a load  $Z_L$  and a series impedance  $Z_s$ . The input impedance  $Z_{in}$  is calculated as:

$$Z_{in} = Z_s + Z_L$$

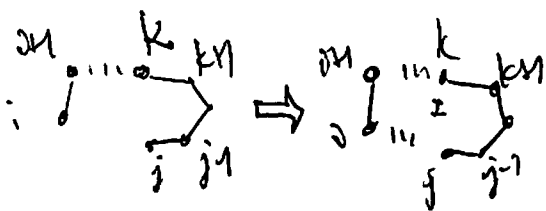
The diagram also shows the calculation of the effective impedance  $Z_{eff}$  for a transmission line with a load  $Z_L$  and a series impedance  $Z_s$ .

$$Z_{eff} = Z_s + Z_L$$

The diagram includes a note: "Digital signal processing".

$$= z_{k_i, j} z_{k_j, i} = \text{local coeff}(k_i, j) l$$

(sta) ← [to match connection]

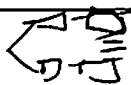


$$Z = \left( \frac{G_{Hd}}{K_d} \right) \times \left( \frac{L_{oxz}}{L_d} \right) \times K_{oxz} = \frac{Z_{ox} (G_{Hd}) \times \left( \frac{L_{oxz}}{L_d} \right) \times K_{oxz}}{K_d}$$

See Latin -  
many relative to  
depth of

(5)

# Alternative scaling



define  $K_d^{max}$   $\rightarrow$   $K_{max} = \frac{C_{init} l_{bp}^2 l_{exp}}{K_{loss}}$   $W?$

then,

$$K_{loss} = \frac{C_{init} l_{bp}^2 l_{exp}}{K_{d^{max}}} = \frac{C_{eff}^{BP}}{C_{init} l_{exp}}$$

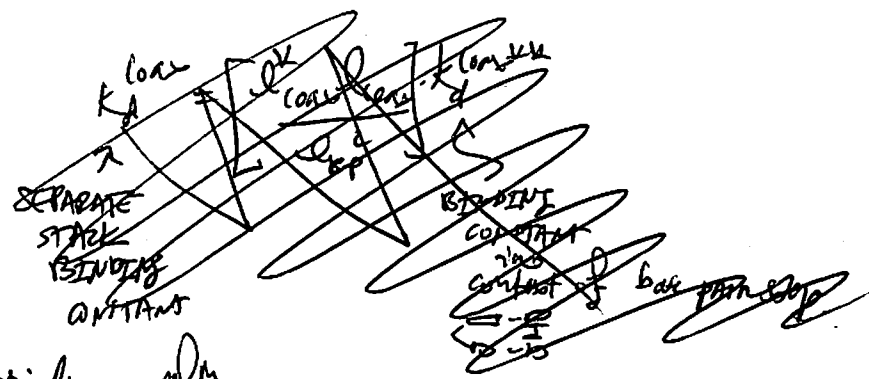
$$K_d^{loss*} = \left[ \frac{C_{eff}^{BP}}{C_{init} l_{bp}^2} \right] \left( \frac{1}{l_{exp} C_{init}} \right) = \frac{C_{eff}^{BP}}{C_{init} l_{bp}^2 l_{exp} C_{init}}$$

and

$$K_d^{loss} = \frac{C_{init} l_{exp}}{K_{loss}} = \frac{C_{init} l_{exp}}{\frac{C_{eff}^{BP}}{C_{init} l_{bp}^2 l_{exp} C_{init}}} = \frac{C_{init}^2 l_{exp}^2 l_{bp}^2}{C_{eff}^{BP}}$$

then

$$K_{loss} = K_d^{loss*} = \left[ \frac{C_{eff}^{BP}}{C_{init} l_{bp}^2} \right] \frac{1}{C_{init}}$$



Alternatively



then

$$K_d^{loss*} = \frac{C_{init} l_{exp}}{K_{loss}}$$

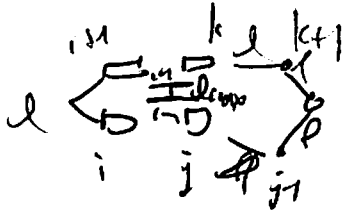
$$K_{loss} = \frac{C_{init} l_{exp}}{K_d^{loss*}}$$

$$\frac{C_{init} l_{exp}}{K_{loss}} = \frac{C_{eff}^{BP}}{C_{init} l_{exp}}$$

$$K_{loss} = \frac{C_{eff}^{BP}}{C_{init} l_{exp}^2}$$

and

$$K_d^{loss} = \frac{C_{init} l_{exp}}{K_{loss}} = \left[ \frac{l_{exp}}{C_{init}} \right] K_d^{loss*}$$



$$Z = \text{Unit Std } K_{\text{Ox}} P^{\frac{1}{2}}_{\text{Ox}} / K_a^2$$

$$= \text{Gross Cost} \frac{\text{Gross Weight}}{\text{Kd}} \cdot d^3 \text{ Gross / } K_d^{BPZ}$$

$$= \left( \frac{C_{SH}}{L_{BP}} \right) \times \left( \frac{C_{INT} L_{COAS}}{K_{BP}} \right) \left( \frac{C_{INT} L_{COAS}}{L_{COAS}} \right)$$

$\uparrow$                    $\uparrow d$                    $\uparrow$

IN, LO      FROM AXE      COST OF DERIV  
PARTIALIZING      AND LOSS?

Enter answer on next page (52)

Let's try to write out newstone clearly:

$$Z_{BP}(i, j) = \alpha_i \alpha_j \frac{C_{BP}(i+1, j-1)}{K_{BP}} + \sum_{i < c < j} \frac{C_{std}}{K_{BP}} Z_{linear}(i+1, c) Z_{linear}(c+1, j-1)$$

i.e. force cutpoint

Scatter order

$$+ \sum_{i+1 < k < j-1} \alpha_i \alpha_k \alpha_j \frac{Z_{BP}(i+1, k) C_{BP}(k+1, j-1)}{K_{BP}} + \sum_{i+1 < k < j-1} \alpha_i \alpha_k \alpha_j \frac{Z_{BP}(i+1, k) C_{std}}{K_{BP}}$$

Could combine with temp release of  $Z_{cut}(j, j-1) = 1$

Replace with:

$$Z_{cut}(k+1, j-1) = \sum_{k < c < j-1} \frac{C_{std}}{K_{BP}} Z_{linear}(k+1, c) Z_{linear}(c+1, j-1)$$

$$+ (1-\alpha_j) \alpha_j Z_{BP}(i+1, j-1) \frac{C_{std}}{K_{BP}} + \sum_{i+1 < k < j-1} \alpha_i \alpha_k \alpha_j \frac{Z_{BP}(k+1, j-1) C_{std}}{K_{BP}} + \sum_{i+1 < k < j-1} \alpha_i \alpha_k \alpha_j \frac{Z_{BP}(k+1, j-1) C_{BP}(i+1, k-1)}{K_{BP}}$$

$$+ \alpha_i \alpha_j \frac{C_{BP}(i+1, i+1, j-1)}{K_{BP}}$$

$$C_{BP}^{std} \sim C_{cut}^{std} K_{BP}$$

IN FACT USE THIS AS FIRST ESTIMATE OF  $K_{BP}$

(12)

Let's REORDER and simplify ZBP recursion.

First Redefn:

$$Z_{cut}(i, j) = \sum_{i \leq k < j} Z_{linear}(i+1, k) Z_{linear}(k+1, j-1) \underbrace{[1 - \alpha_c]}_{\text{cutpoint}}$$

So ... Single Cut DEFINING  $Z_{linear}(i+1, i) \equiv 1$ . (check:  $Z_{cut}(i, i+1) = 1$  cutpoint)

$$Z^{BP}(i, j) = \alpha_i \alpha_{j-1} C_{off}^{BP}(i+1, j-1) Z^{BP}/K_{BP} + \frac{C_{ad}}{K_{BP}} Z_{cut}(i, j)$$

$$+ \alpha_i \alpha_{j-1} C_{off}^{BPS}(i, j-1; i+1, j-1) / K_{BP}$$

STACKED BASE PAIR

$$+ \sum_{i+1 < k < j-1} \alpha_i \alpha_{j-1} Z_{BP}(i+1, k) C_{off}(k+1, j-1) Z_{BP}^{(i, j-1; k+1, j-1)} / K_{BP}$$

$$+ \sum_{i+1 < k < j-1} \alpha_i \alpha_{j-1} \alpha_{k+1} Z_{BP}(k+1, j-1) Z_{BP}^{(i, j-1; k+1, j-1)} / K_{BP}$$

$$+ \sum_{i+1 < k < j-1} \alpha_i Z_{BP}(i+1, k) C_{ad} K_{cut} Z_{cut}(k+1, j-1) / K_{BP}$$

$$+ \sum_{i < k < j-1} \alpha_{j-1} Z_{cut}(i, k) Z_{BP}(k+1, j-1) C_{ad} K_{cut} / K_{BP}$$

$$K_{BP}^{(BP)} [C_{off}^{BPS}(BP, BP) / C_{cut} L_{coop}] \rightarrow \text{first approximation.}$$

OVER



Rest of the recursions...

$Z_{cut} \rightarrow$  see prev. page (62)

$Z_{BP} \rightarrow$  see prev. page (62)

$$Z_{linear}^{(i,j)} = \sum_{i < k < j-1} \alpha_k Z_{BP}^{(i,k)} Z_{BP}^{(k,j)} K_{cons}^{(k,i;j,k,j)} \left( \frac{k-i-1}{i} \right) \left( \frac{j-k-1}{j} \right)$$

$$C_{off}^{(i,j)} = \alpha_{j-1} C_{off}^{(i,j-1)} l + \sum_{i < k < j} C_{off}^{(i,k-1)} l Z_{BP}^{(k,j)} l_{BP} + \sum_{i < k < j} C_{off}^{(i,k-1)} l Z_{cons}^{(k,j)} l_{cons}$$

$$Z_{linear}^{(i,j)} = \alpha_{j-1} Z_{linear}^{(i,j-1)} + Z_{BP}^{(i,j)} + \sum_{i < k < j} Z_{linear}^{(i,k-1)} Z_{BP}^{(k,j)} + Z_{cons}^{(i,j)} + \sum_{i < k < j} Z_{linear}^{(i,k-1)} Z_{cons}^{(k,j)}$$

$$Z_{fine}^{(i,i)} = \begin{cases} Z_{linear}^{(i,i-1)} & \text{if } i-1-i \text{ is cutpoint} \\ C_{off}^{(i,i-1)} l + \sum_{i < k < j-1} Z_{linear}^{(i,k-1)} Z_{linear}^{(k,j)} + \sum_{i < j < k < j+1} \alpha_j Z_{BP}^{(i,j)} Z_{BP}^{(j+1,i-1)} C_{off}^{(j+1,j,i-1)} + \sum_{i < j < k < j+1} Z_{BP}^{(i,j)} Z_{BP}^{(k,i-1)} K_{cons}^{(j,i;j+1,i-1)} Z_{cons}^{(j+1,k)} + \sum_{i < j < k < j+1} Z_{BP}^{(i,j)} Z_{BP}^{(k,i-1)} K_{cons}^{(j,i;j+1,i-1)} C_{off}^{(j+1,k)} l_{BP} \end{cases}$$

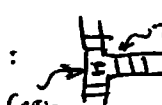
more TO DO:

- check  $\alpha_{j,k}$  in last term of  $Z_{fine}$
- code up  $C_{off}$  no  $l_{BP}$  sheet
- code up  $C_{off}$  no  $l_{BP}$  sheet
- PARAP modeling
- Run - is it work?

TO DO

- check cutpoint (MSTP) (54)
- check written here (53)
- MPO or how to do all this
- code up  $Z_{cut}$
- try to...

(64)

Notes on how to NSALLOW: 

In expression for  $z^{BP}$  [p. 61], make following replacements:

$$\begin{aligned}
 z^{BP} &= \alpha_i \alpha_{j-1} \overbrace{c_{eff}^{BP}(i+1, j-1)}^{\text{no coarse stretch}} l_{coarse}^{BP} + \dots \\
 &+ \sum_{i+k < j-1} \alpha_i \alpha_k \alpha_{j-1} z^{BP}(i+1, k) \overbrace{c_{eff}^{BP}(k+1, j-1)}^{\text{no BP stretch}} l_{coarse}^{BP} K_{coarse}(i, j; i+1, k) / K_{BP} \\
 &+ \sum_{i+k < j-1} \alpha_i \alpha_k \alpha_{j-1} \overbrace{c_{eff}^{BP}(i+1, k-1)}^{\text{no BP stretch}} z^{BP}(k, j-1) l_{coarse}^{BP} K_{coarse}(i, j; k, j-1) / K_{BP}
 \end{aligned}$$

where:

$$\overbrace{c_{eff}^{BP}(i, j)}^{\text{no coarse stretch}} = c_{eff}^{BP}(i, j) - c_{fine/coarse} z^{coarse}(i, j)$$

$$\text{and } \overbrace{c_{eff}^{BP}(i, j)}^{\text{no BP stretch}} = c_{eff}^{BP}(i, j) - c_{fine/BP} z^{BP}(i, j) \quad \text{"subtraction" expressions}$$

pretty easy. might actually just track

$$c_{eff}^{BP}(i, j) = \alpha_{j-1} c_{eff}^{BP}(i, j-1) l + \sum_{i+k < j} c_{eff}^{BP}(i, k) l z^{BP}(k, j) l_{BP} + \sum_{i+k < j} c_{eff}^{BP}(k-1, j) l z^{BP}(i, k) l_{coarse}$$

$$\overbrace{c_{eff}^{BP}(i, j)}^{\text{no coarse stretch}} = c_{eff}^{BP}(i, j) + c_{fine/BP} z^{BP}(i, j)$$

$$\overbrace{c_{eff}^{BP}(i, j)}^{\text{no BP stretch}} = c_{eff}^{BP}(i, j) + c_{fine/coarse} z^{BP}(i, j)$$

$$c_{eff}^{BP}(i, j) = \overbrace{c_{eff}^{BP}(i, j)}^{\text{no coarse stretch}} + \overbrace{c_{eff}^{BP}(i, j)}^{\text{no BP stretch}} + c_{fine/coarse} z^{BP}(i, j)$$

would also for simpler back tracking than "subtraction" expressions above

Later realized that I also need to extract out contribution from joint BP singlet + one singlet & Z<sub>BP</sub>

⇒ see p. 68

(65)

Revisiting

Start out  
4  
A

4  
A - x  
5 5

$$Z_{BP} \rightarrow \infty$$

$$Z_{BP} \rightarrow \infty$$

$$Z_{load} = 0$$

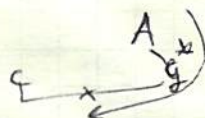
$$Z_{load} = 0$$

$$C_{eff} = C_{input}$$

$$C_{eff} = C_{input}$$

$$Z_{linear} = 1$$

$$Z_{linear} = 1$$



$$Z_{BP} \rightarrow \infty$$

$$Z_{BP} \rightarrow \infty$$

$$Z_{load} \rightarrow \infty$$

$$Z_{load} \rightarrow \infty$$

$$C_{eff} \rightarrow \infty$$

$$Z_{linear} = 0$$

(trying to debug implementation in alpha field)



$$Z_{BP} \rightarrow \infty$$

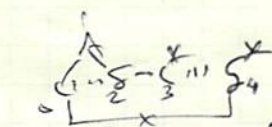
$$Z_{load} = 0$$

$$C_{eff} = \infty$$

$$Z_{linear} = \frac{C_{st} + 1}{K_{BP}} \left( 1 + \frac{K_{load}}{K_{BP}} \right)$$

$$Z_{BP} = \frac{C_{input} K_{BP}}{K_{BP}}$$

$$Z_{BP} = \frac{C_{st} + 1}{K_{BP}}$$



$$Z_{BP} = \frac{C_{st} + 1}{K_{BP}} Z_{linear} [C_{st} + 1] Z_{linear} [5] + \frac{K_{load}}{K_{BP}} Z_{BP} [C_{st} + 1] Z_{linear} [5]$$

$$= \frac{C_{st} + 1}{K_{BP}} \times \left( 1 + \frac{C_{input} K_{BP}}{K_{BP}} \right) + \frac{C_{load} K_{load} K_{BP}}{K_{BP} K_{BP}}$$

$$\left[ \frac{C_{input} K_{BP}}{K_{BP}} \right] \left[ \frac{C_{st} + 1}{K_{BP}} \right] \times K_{load}$$



(Ah, was missing this term.)

①

OCT 21, 2018

# WRITE IT ALL DOWN

Stuck in window locks



$z = z_1 + z_2$  ← allows only for single BASE PARA to stab →  
 (diagram of handle)  $z_{ep} = \frac{1}{k_3} (k_1 (z_1 + z_2) + k_2 (z_1 + z_2))$   
 (diagram of handle)  $z_{ep} = \frac{1}{k_3} (k_1 (z_1 + z_2) + k_2 (z_1 + z_2))$   
 (diagram of handle)  $z_{ep} = \frac{1}{k_3} (k_1 (z_1 + z_2) + k_2 (z_1 + z_2))$



CANDLEWOOD

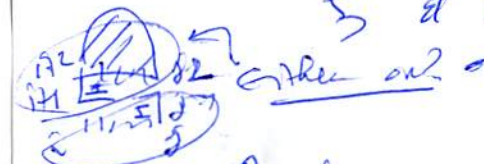
SUITES  
AN IHG® HOTEL

②

# WRITE IT ALL DOWN

$$z_{ep} = \frac{1}{k_3} (k_1 (z_1 + z_2) + k_2 (z_1 + z_2))$$

Interesting in  $\pm$  think  $\pm$  could get the all the equations to work out self consistently with just 3' PAGES  
 - But what if we have  $\pm$  of  $\pm$  dangle?



$$z_{ep}(i,j) = \frac{1}{k_3} (k_1 (z_1 + z_2) + k_2 (z_1 + z_2))$$



CANDLEWOOD

SUITES  
AN IHG® HOTEL





Exactly same as Markov 2004 model.

→ NOTE THAT A NUCLEOTIDE CANNOT BE BOTH 5' & 3' DANGLE BPV  
→ NOTE THAT DANGLES act as "caps" → cannot continue into mutations, only into eff loops.  
model in dangles's terms to add into recursions on p(62)-(63)

$$\begin{aligned} Z_{BP}(i,j) &= \alpha_i \alpha_{i+1} \alpha_j \left( \text{Coff}_{\text{dangle}}(i+2, j-1) \ell_{\text{dangle}}^2 \ell_{BP} / K_d \right. \\ &+ \left. \frac{1}{\sqrt{\pi}} \alpha_i \alpha_{i+1} \alpha_j \text{Csta} K_{\text{dangle}}(i,j;i+1) Z_{\text{cut}}(i+1, j) / K_d \right. \\ &+ \alpha_i \alpha_j \alpha_{j-1} K_{\text{dangle}}(i, j; j-1) \text{Ceff}(i+1, j-2) \ell_{\text{dangle}}^2 \ell_{BP} / K_d \\ &+ \left. \alpha_i \alpha_j \alpha_{j-1} K_{\text{dangle}}(i, j; j-1) Z_{\text{cut}}(i, j-1) / K_d \right) \end{aligned}$$

Could also write  $K_3' = \text{Ceff} \ell_{\text{dangle}}^2 / K_d$   $K_5' = \text{Csta} \ell_{\text{dangle}} / K_d$

REVISION ON (NEXT PAGE)

$$Z_{\text{dangle}}(i,j) = K_{\text{dangle}}(j,i;i) \times Z_{BP}(i+1, j)$$

$$Z_{\text{dangle}}(i,j) = K_{\text{dangle}}(j,i;j) Z_{BP}(i, j-1) \quad i \rightarrow j$$

$$\begin{aligned} \text{Ceff}(i,j) &= \text{Ceff}_{\text{dangle}} Z_{\text{dangle}}(i,j) + \sum_{i < k < j} \text{Ceff}(i,k) \ell_{\text{dangle}}^2 \ell_{\text{dangle}} \\ &+ \text{Ceff}_{\text{dangle}} Z_{\text{dangle}}(i,j) + \sum_{i < k < j} \text{Ceff}(i,k-1) \ell_{\text{dangle}}^2 \ell_{\text{dangle}} \end{aligned}$$

$$\begin{aligned} Z_{\text{dangle}}(i,j) &= Z_{\text{dangle}}(i,j) + \sum_{i < k < j} \frac{\text{Ceff}(i,k-1)}{\ell_{\text{dangle}}} Z_{\text{dangle}}(k,j) \\ &+ Z_{\text{dangle}}(i,j) + \sum_{i < k < j} \text{Ceff}(i,k) Z_{\text{dangle}}(i,k-1) Z_{\text{dangle}}(k,j) \end{aligned}$$

$$\begin{aligned} Z_{\text{final}}(i,i) &= \sum_{i-2 \leq j \leq i} \alpha_j \alpha_{i-2} K_{\text{dangle}}(j,i;i-1) \ell_{\text{dangle}}^2 \text{Ceff}(j+1, i-2) Z_{BP}(i,j) \\ &+ \sum_{i-2 \leq j \leq i-1} K_{\text{dangle}}(j,i;i-1) Z_{BP}(i,j) Z_{\text{cut}}(j, i-1) \\ &+ \sum_{i-1 \leq j \leq i} \alpha_i \alpha_{j-1} K_{\text{dangle}}(i,j;j) \text{Ceff}(i+1, j-1) \ell_{\text{dangle}}^2 Z_{BP}(i, j-1) \\ &+ \sum_{i-1 \leq j \leq i} K_{\text{dangle}}(i,j;j) Z_{\text{cut}}(i,j) Z_{BP}(j, i-1) \end{aligned}$$

I ~~was~~ ~~the~~ was previously following this line of reasoning:

$$\text{in } \frac{1}{2} \left[ \frac{1}{2} \right]_j \quad K_{3'} = \text{Cart} \frac{1}{2} K_d'$$

$$\text{in } \frac{1}{2} \left[ \frac{1}{2} \right]_j \quad Z = K_5 \times \text{Cart} \frac{1}{2} K_d' \approx \text{Cart} \frac{1}{2} K_d'$$

$$\text{is identity } \text{Cart} \frac{1}{2} K_d' = \text{Cart} \frac{1}{2} K_d'$$

$$\text{and analogously } \text{Cart} \frac{1}{2} K_5' = \text{Cart} \frac{1}{2} K_5'$$

$$K_5' l_5' = l_5' K_5' \quad (?)$$

$$l_5' = K_5' l_5' / l_5' \quad \text{O.K.}$$

mismatch/  
BASE PAIR STACK  
Cart  $\frac{1}{2} K_d'$

in approximation  
(likely incorrect)  
that 1' and 3'  
stacks are  
not conflicting.

TO DO:







① CASE for with periodic Boundary Condition → check

Other de leix ~~de~~ cherb fctn.

③ then, do accumulations of contributions →  $\left\{ \begin{array}{l} \text{Cumulative} \\ \text{Backtrack} \\ \text{Planned} \leftarrow \begin{array}{l} \text{Stochastic} \\ \text{Backtrack} \\ \text{MFE} \end{array} \end{array} \right.$

(c) Then, port parameters from Vienna<sup>18.5</sup>~~/RPA/SPAA/RAA/PAA~~

ie, Turner 1999  
 Uredo on +RVA, incl. F. m. R. 2;  
~~Log~~ S. NATIVE

Q. Do we want chemical mapping data for every TRA?

→ (5) Excellent ~~takeoff~~ point for ~~optimization~~ testing 3<sup>rd</sup> ~~stage~~ for TPRN & SA Bar  
mats. CU-tune  
PARK

⑥ then ~~log~~  $\log N$  - logs

⑦ Then code up hct to optimize  
[w/ hyper binding]

250072.5

Peripheral [giving "issues"]

⑧ Dangler

⑨ ~~file~~ <sup>file</sup> ~~does~~ <sup>converts</sup> into class, and write "compile" to translate into fast python or C++ code.

A way to make the code even more readable...

Consider a term:

$$Z_{BP}(i,j) = \sum_{i+1 < k < j-1} Z_{BP}(i+1, k) C_{eff}(k, j-1) \underbrace{\frac{2}{Z_{BP}(k, j-1)}}_{Z_{BP}^{CONTRIB}}$$

Then derivative w.r.t. any parameter  $\lambda$  is given by Chain Rule:

$$\frac{1}{Z_{BP}(i,j)} \frac{\partial Z_{BP}(i,j)}{\partial \lambda} = \frac{\partial Z_{BP}^{CONTRIB}(i,j)}{\partial \lambda} = \frac{\partial Z_{BP}(i+1, k)}{\partial \lambda} + \frac{1}{C_{eff}(k, j-1)} \frac{\partial C_{eff}(k, j-1)}{\partial \lambda} + 3 \frac{1}{2} \frac{\partial}{\partial \lambda} + 2 \frac{1}{2} \frac{\partial \log Z_{BP}}{\partial \lambda} + \frac{1}{k_{cons}} \frac{\partial k_{cons}}{\partial \lambda} - \frac{1}{k_{cons}} \frac{\partial \lambda}{\partial \lambda}$$

$$\text{and } \frac{\partial Z_{BP}(i,j)}{\partial \lambda} = \sum_{i+1 < k < j-1} \frac{\partial Z_{BP}^{CONTRIB}}{\partial \lambda}(i, k, j)$$

could also write in terms of DERIVATIVES:

$$\begin{aligned} \frac{\partial}{\partial \lambda} \log Z_{BP}(i,j) &= \frac{1}{Z_{BP}(i,j)} \frac{\partial Z_{BP}(i,j)}{\partial \lambda} \\ &= \frac{1}{Z_{BP}(i,j)} \left[ \sum_{i+1 < k < j-1} Z_{BP}^{CONTRIB}(i, k, j) \times \left[ \frac{\partial}{\partial \lambda} \log Z_{BP}(i+1, k) + \frac{\partial}{\partial \lambda} \log C_{eff}(k, j-1) \right] \right] \end{aligned}$$



Probably easier to write out  
chain rule for product:

$$Z_{BP}(ij) = \sum_{i+1 \leq k \leq j-1} Z(i+1, k) c_{\text{eff}}(k+1, j-1) \text{lll}_{BP} \text{lll}_{BP} k_{\text{con}} k_a^{BP} \dots$$

$$\begin{aligned} \frac{\partial}{\partial \lambda} Z_{BP}(ij) = & \sum_{i+1 \leq k \leq j-1} \left[ \frac{\partial Z_{BP}(i+1, k)}{\partial \lambda} c_{\text{eff}}(k+1, j-1) \text{lll}_{BP} \text{lll}_{BP} k_{\text{con}} k_a^{BP} + \right. \\ & + Z_{BP}(i+1, k) \frac{\partial c_{\text{eff}}(k+1, j-1)}{\partial \lambda} \text{lll}_{BP} \text{lll}_{BP} k_{\text{con}} k_a^{BP} + \\ & + Z_{BP}(i+1, k) c_{\text{eff}}(k+1, j-1) \frac{\partial \text{lll}_{BP} \text{lll}_{BP} k_{\text{con}} k_a^{BP}}{\partial \lambda} + \\ & \left. + Z_{BP}(i+1, k) c_{\text{eff}}(k+1, j-1) \text{lll}_{BP} \text{lll}_{BP} k_{\text{con}} \frac{\partial k_a^{BP}}{\partial \lambda} \right] \end{aligned}$$

So we could actually keep track of ~~the~~ objects

$$\begin{pmatrix} Z_{BP} \\ \frac{\partial}{\partial \lambda} Z_{BP} \end{pmatrix} \begin{pmatrix} c_{\text{eff}} \\ \frac{\partial}{\partial \lambda} c_{\text{eff}} \end{pmatrix} \begin{pmatrix} \text{lll} \\ \frac{\partial}{\partial \lambda} \text{lll} \end{pmatrix}$$

and when we take product, use product rule:

$$\begin{pmatrix} A \\ \frac{\partial A}{\partial \lambda} \end{pmatrix} * \begin{pmatrix} B \\ \frac{\partial B}{\partial \lambda} \end{pmatrix} = \begin{pmatrix} A \cdot B \\ \frac{\partial A}{\partial \lambda} B + A \frac{\partial B}{\partial \lambda} \end{pmatrix}$$

Problem: I am seeing a lot of overhead  
in Python when I overload  
even the `--getitem--` class;  
e.g. to automatically wrap

$$Z_{BP}(i+1, k) \Rightarrow Z_{BP}(\text{mod}(i+1, N), \text{mod}(k, N))$$

Can't have these in inner loops of code.

Needed  
FOR  
MFG  
Stochastic  
BACKTRACKING

Could also accumulate separate  
contributions to each path-function  
through overloading addition:

$$Z = AB + CD$$

~~23~~

List of  
Contributions

$$\begin{aligned} \begin{pmatrix} Z \\ \frac{\partial Z}{\partial A} \\ \frac{\partial Z}{\partial B} \\ Z_{CONTRIB} \end{pmatrix} &= \begin{pmatrix} A \\ \frac{\partial A}{\partial A} \\ \frac{\partial A}{\partial B} \\ A_{CONTRIB} \end{pmatrix} \begin{pmatrix} B \\ \frac{\partial B}{\partial A} \\ \frac{\partial B}{\partial B} \\ B_{CONTRIB} \end{pmatrix} + \begin{pmatrix} C \\ \frac{\partial C}{\partial A} \\ \frac{\partial C}{\partial B} \\ C_{CONTRIB} \end{pmatrix} \begin{pmatrix} D \\ \frac{\partial D}{\partial A} \\ \frac{\partial D}{\partial B} \\ D_{CONTRIB} \end{pmatrix} \\ &= \begin{pmatrix} AB \\ \frac{\partial A}{\partial A} B + A \frac{\partial B}{\partial A} \\ \frac{\partial A}{\partial B} B + A \frac{\partial B}{\partial B} \\ [AB] \end{pmatrix} + \begin{pmatrix} CD \\ \frac{\partial C}{\partial A} D + C \frac{\partial D}{\partial A} \\ \frac{\partial C}{\partial B} D + C \frac{\partial D}{\partial B} \\ [CD] \end{pmatrix} \\ &= \begin{pmatrix} AB + CD \\ \frac{\partial A}{\partial A} B + A \frac{\partial B}{\partial A} + \frac{\partial C}{\partial A} D + C \frac{\partial D}{\partial A} \\ \frac{\partial A}{\partial B} B + A \frac{\partial B}{\partial B} + \frac{\partial C}{\partial B} D + C \frac{\partial D}{\partial B} \\ [AB, CD] \end{pmatrix} \end{aligned}$$

- Again, don't want the accumulation in inner loops if code, due to list of overloading.
- In addition, keeping track of contributions will cost  $N^3$  memory (will probably limit  $N \ll 1000$ ) and incur overhead in dynamic updating of lists.
- ⇒ instead only do it if we're backtracking and perhaps even repeat calculations at those steps.



To solve code readability problems, and code copying  
 could simply write a little "compiler"  
 that goes from a text file <sup>py</sup> recursions.txt  
 of commented recursions to a  
 single python file:  
 derived-recursions.py [for alphafold]  
 and/or

- derived-recursions.h [for C++]
- (Alternative would be MACROS but I find <sup>alphafold</sup> those cryptic)
- In fact, as stepping stone (or sanity check),

could also code up objects to ~~test~~  
 "rigorously" do derives and contributions too,  
 even at the expense of speed.

# Thoughts on convexity

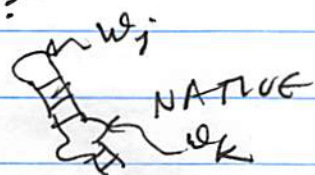
$$Z = \sum_{\text{structures } s} e^{\sum_j w_j n_j(s)}$$

$$\frac{\partial}{\partial w_j} \log Z = \frac{1}{Z} \sum_{\text{structures } s} n_j e^{-\sum_j w_j n_j(s)} = \langle n_j \rangle$$

$$\begin{aligned} \frac{\partial^2}{\partial w_j \partial w_k} \log Z &= \left( -\frac{1}{Z^2} \frac{\partial Z}{\partial w_k} \right) \sum_{\text{structures } s} n_j e^{-\sum_j w_j n_j(s)} \\ &\quad + \left( \frac{1}{Z} \right) \sum_{\text{structures } s} n_j n_k e^{-\sum_j w_j n_j(s)} \\ &= \langle n_j n_k \rangle - \langle n_j \rangle \langle n_k \rangle \end{aligned}$$

Suppose I have maximum likelihood target: [or minimal free energy gpt]

$$\Delta S_{\text{NATIVE}} = -\log Z_{\text{NATIVE}} + \log Z$$



$$\frac{\partial^2}{\partial w_j \partial w_k} \Delta S_{\text{NATIVE}} = \left[ \frac{\partial^2}{\partial w_j \partial w_k} (-\log Z_{\text{NATIVE}}) \right] + \left[ \langle n_j n_k \rangle - \langle n_j \rangle \langle n_k \rangle \right]$$

$$= \langle n_j n_k \rangle - \langle n_j \rangle \langle n_k \rangle$$

Is this covariance matrix always positive definite?



(75)

Note :- KARATSUBA

Return to "simple recursive"

$$Z_{br}^{(i,j)} = \cancel{C_{eff}(i,j-1)} \cancel{Z_{br}^{(i,j-1)}} \cancel{Z_{br}^{(i,j-1)}} + \sum_{i \leq k < j} Z_{linear}(i,k) Z_{linear}(k+1,j-1) \cancel{Z_{br}^{(i,j-1)}} \cancel{Z_{br}^{(i,j-1)}}$$

$$C_{eff}^{(i,j)} = C_{eff}(i,j-1) \cancel{Z_{br}^{(i,j-1)}} + C_{int} Z_{br}^{(i,j)} \cancel{Z_{br}^{(i,j-1)}} + \sum_{i \leq k < j} \cancel{Z_{br}^{(i,j-1)}} C_{eff}(i,k-1) Z_{br}^{(k,j)}$$

$$Z_{linear}^{(i,j)} = Z_{linear}(i,j-1) + Z_{br}(i,j) + \sum_{i \leq k < j} Z_{linear}(i,k-1) Z_{br}(k,j)$$

$$Z_{find}^{(i)} = Z_{linear}(i,i-1) \quad \text{if it is output}$$





alphafold:

img to completion...

PAPER → Rank all existing RPT algorithms based on Eterna

SHAPE  
+  
Array fold  
measures  
measures

PAPER1 [AlphaFold Premiere]   
 •  $\mu$  to + bootstrapping → still need theory   
 • Rank-estimated state values for comp, etc.   
 • Evaluate success on RPT; Tunes  $\Delta G$ ; closed loop RT; m2 binding   
 • All structural motifs incl. Rosetta estimates   
 • logarithmic loop penalty   
 • non-canonical base pairs   
 • write out theory.   
 • check SA on test   
 • inspection   
 • loop states

PAPER2   
 • Partition function incl. 3' structure   
 • Evaluate success via m2 data + progressive experiments on alternative states   
 • Need  $\Delta G$  estimator (incl. normal modes)

PAPER3   
 • Ceff + SD distributions → what's the story?   
 • Is that even "inside" alphafold?