

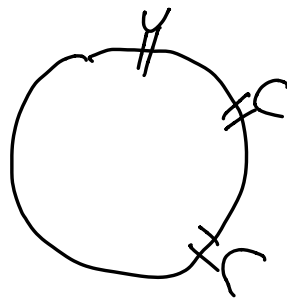
Models (and principles) of early immune recognition and antagonism

G.A.B (NIH)

J.B. Lelanne

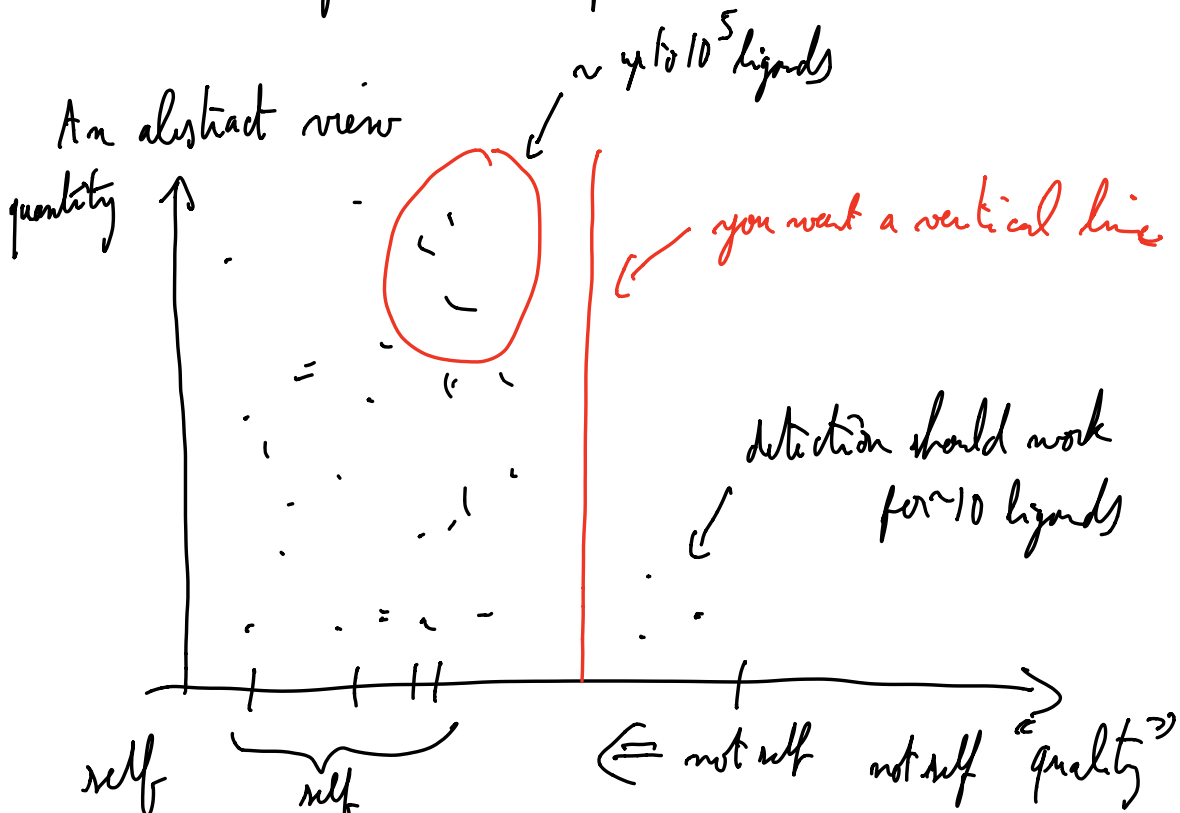
Early detection: 1st 5 minutes

T cells



← scan for environment
"ligands"
30 000 receptors

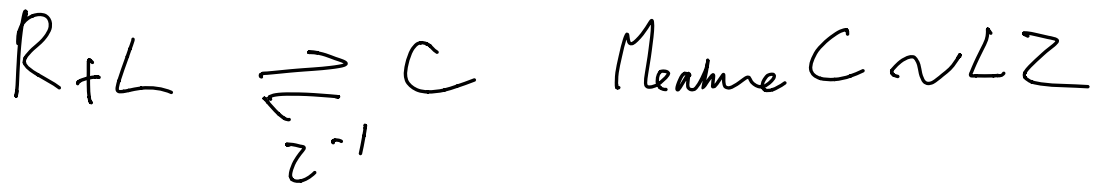
Problem: self vs not self.



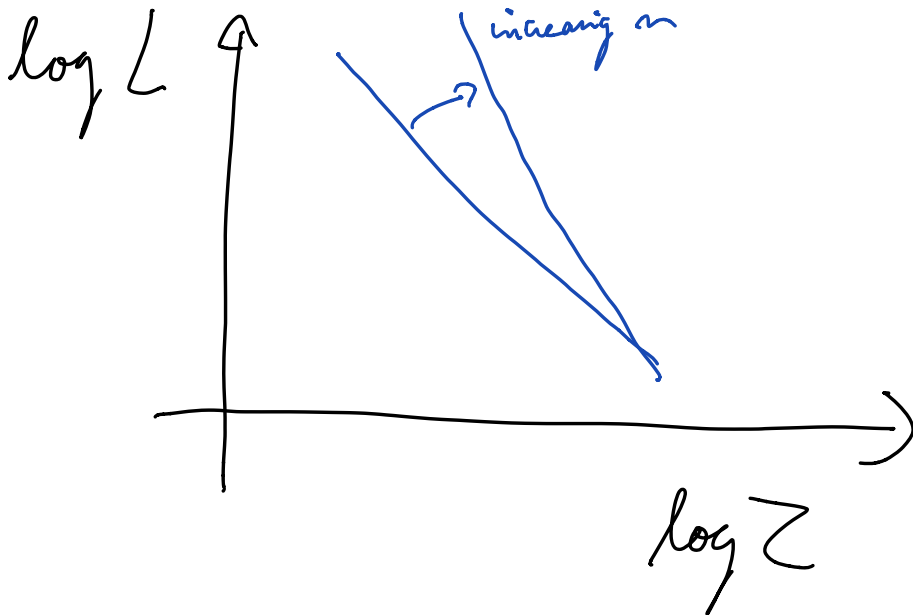
Here quality \equiv binding time of ligand to receptor
threshold around 4s.

Vertical line: tough problem.

I imagine the following



L "compensates" for z



Mc Keithan 95'

KPR

LZ^m

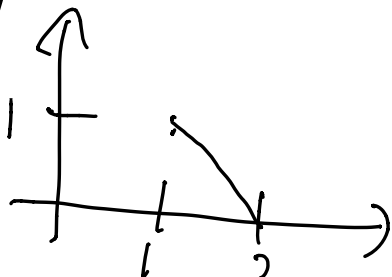
G.A.B and Gemin $n \sim 10000$ + low.

\Rightarrow very complex models (~ 100 equations, μ eq)

Explains a lot of other features such as:

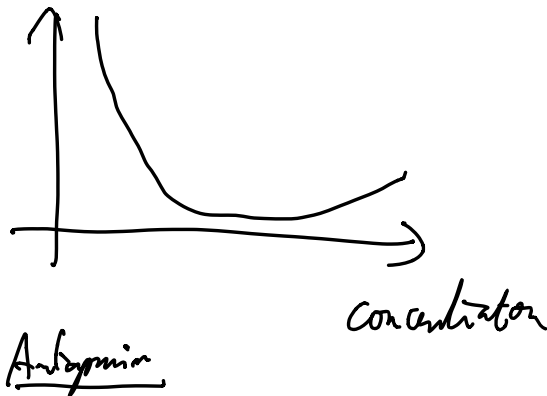
- time of response (~ 5 min) + non monotonic dependency of response time

digital vs analog effect
regenerative cells



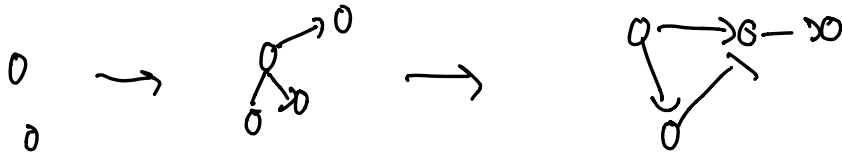
SHP-1 $\xrightarrow{\text{w/ Adiponin}}$ not up

response time



How to make a theory of this?

Tool: "in silico" evolution
(machine learning)



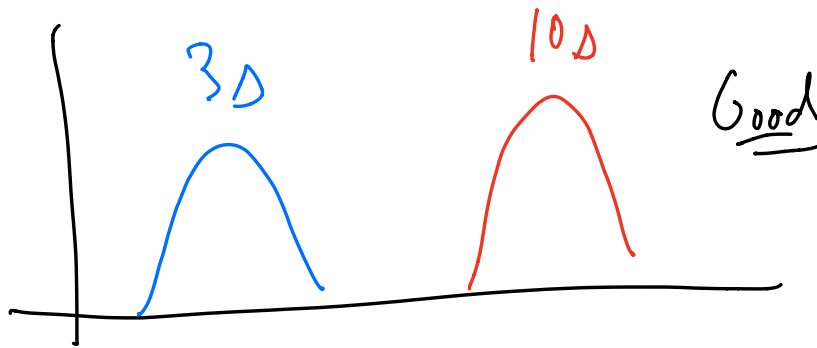
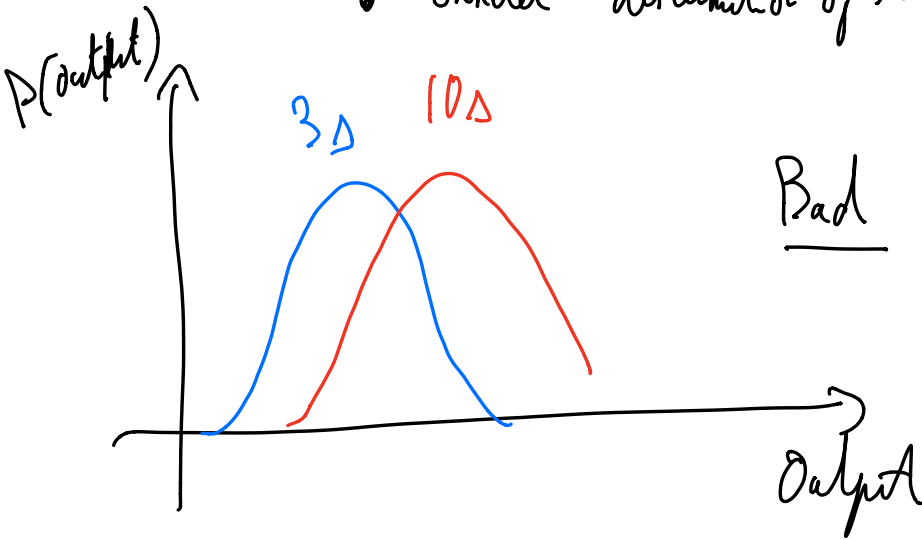
fitness increase

Here fitness quantifies this absolute discrimination



One way to do it: • imagine you take random L connections
 with # bridging times (say 3 or 10s)

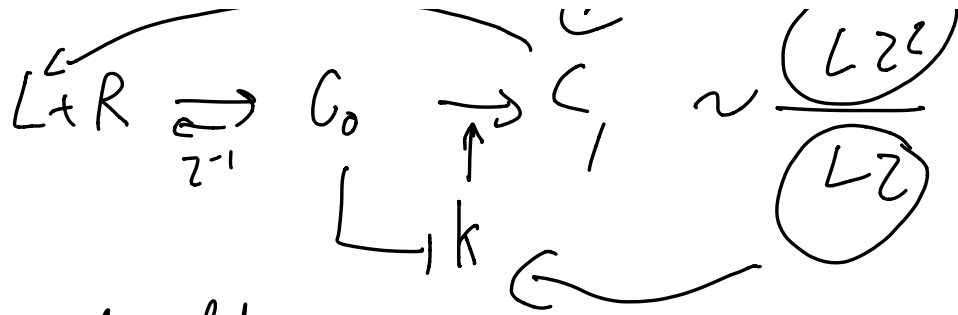
• consider distribution of some output



Mutual Info is a way to do this

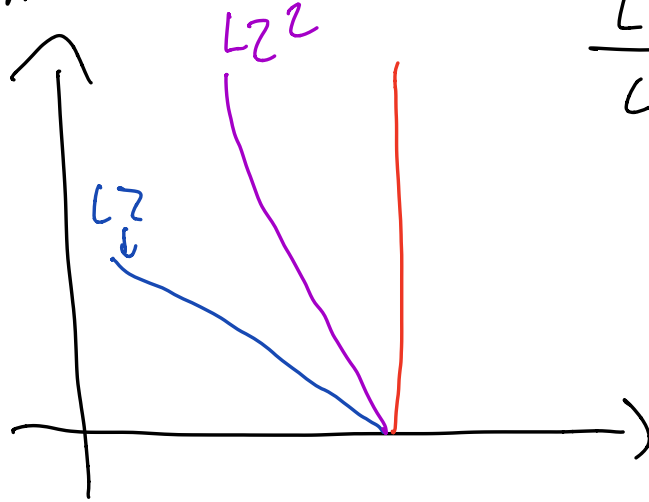
Results

2^{-1} 



Very simple solution

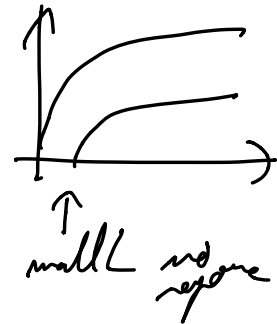
What happens?



$$\frac{LZ^2}{LZ} = (Z)$$

↓
you can detect this

Realistic Dranes



Ultra-simple solution but... not so simple

ANT AGONISM

what happens with mixtures

10 act self vs 10^5 self

WITHIN

More

add this

Pure ligands

$$z_1 > z_c > z_2$$

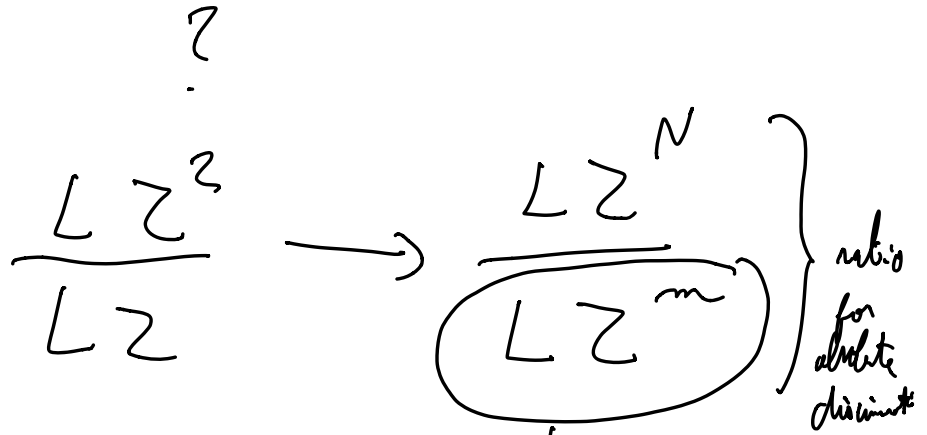
$$\frac{Lz_1^2}{Lz_1} > z_c$$

$$z_1 > \frac{L_1 z_1^2 + L_2 z_2^2}{L_1 z_1 + L_2 z_2}$$

If enough $L_2 \Rightarrow$ get lower than threshold

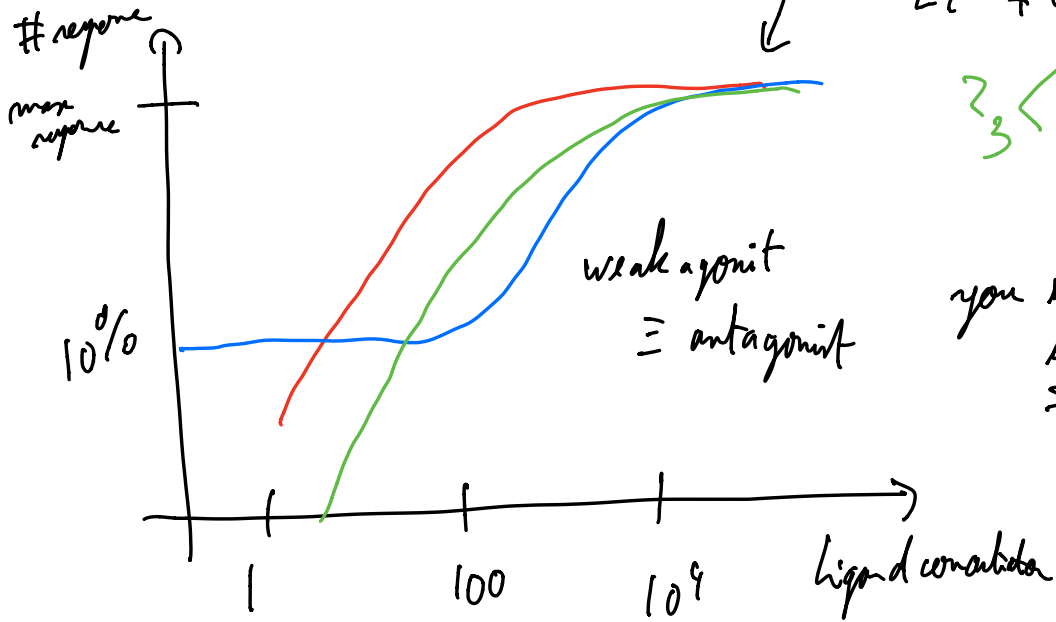
Simple adaptive voting

10^5 self at 0.1 s \Rightarrow kill immune response



proofreading to amplify the strength of agonist vs antagonist
 \Rightarrow removes self influence

Antagonism is the smoking gun



$$\frac{LZ^N}{LZ^m + \epsilon} + \frac{L_e Z_e^N}{L_e Z_e^m}$$

$\epsilon < Z_e < Z$
 $\epsilon < Z_e < Z$

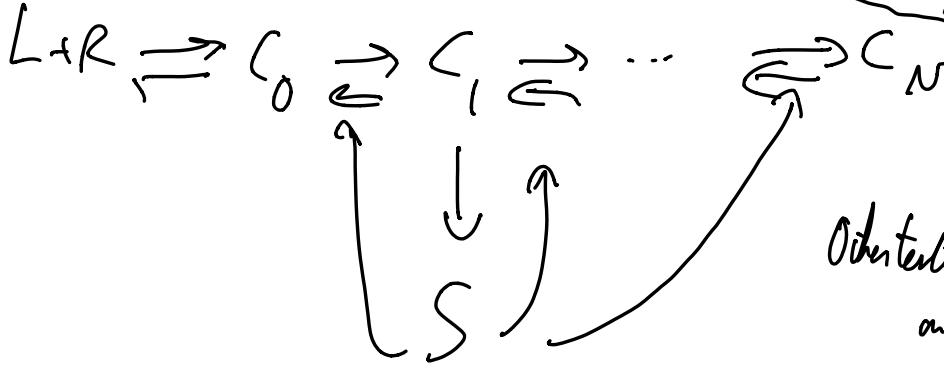
you see this madly!!

More "realistic" model

Central idea

ting of war LZ^N vs LZ^m

But you could get
 $h(A)f(z)$ vs $h^k(z)g(z)$
 new phenomena
 h h really
 nice



but not
 always

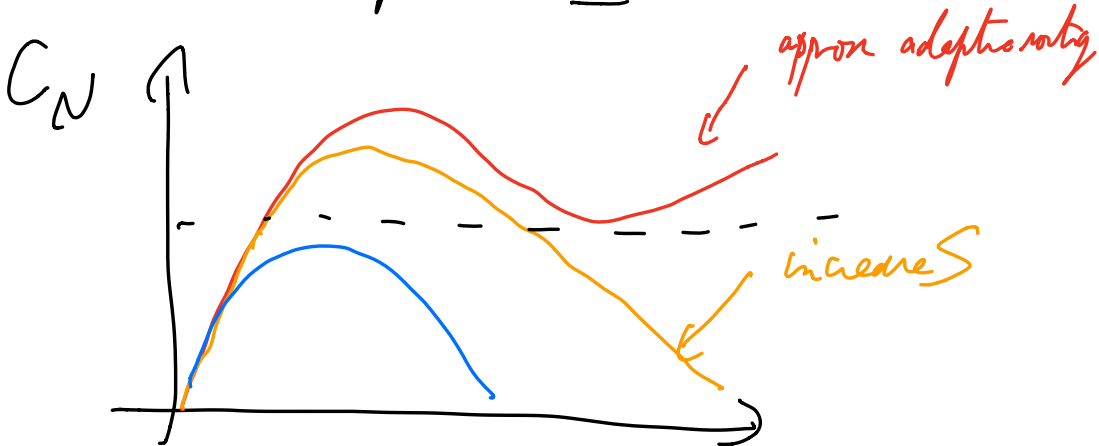
Other text: SMP-1
 antagonizes other
 side of
 the cell

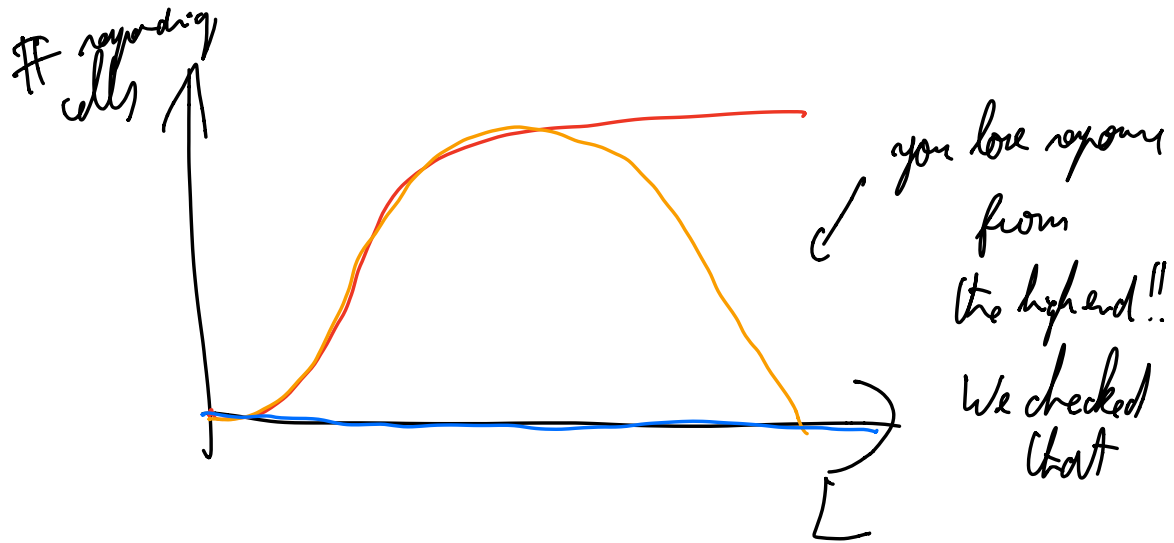
This gives you 2 properties I mentioned:

NMDR

digital vs analog

They are related





Can we get rid of antagonism? NO!!

Theorem

Absolute discrimination

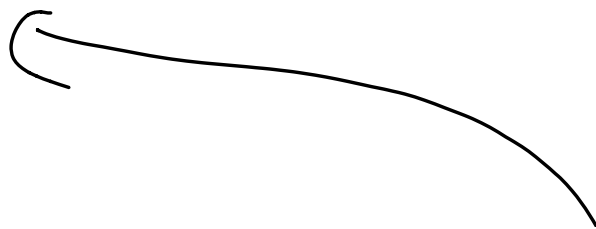


\Rightarrow antagonism

Why? Assume thresholding of continuous variable

$$T(\{L_i, z_i\})$$

$$\left. \frac{\partial T}{\partial z_i} \right|_{z_i = z_c} \gg 0$$



On the line $T(L, z_c) = 0$ by definition

Taylor expand on the line

$$T(\{L_i, z_i\}) = T(\sum L_i, z_c) + \sum \frac{\partial T(L, z_c)}{\partial L_i} L_i (z_i - z_c)$$

$$T(\{L_1, z_c; L_2, z_c - dz\}) = T(L_1 + L_2, z_c) - dz \frac{\partial T(L, z_c)}{\partial L_2} L_2$$

$$= T(L_1, z_c) - dz \frac{\partial T(L, z_c)}{\partial L_2} L_2$$

$< 0 \Rightarrow$ antigenic

linear term, now you can build many \neq models at the non linear order

ex:

$$L + R \xrightarrow{K} D \xrightarrow{\mu^{-1}} \phi$$

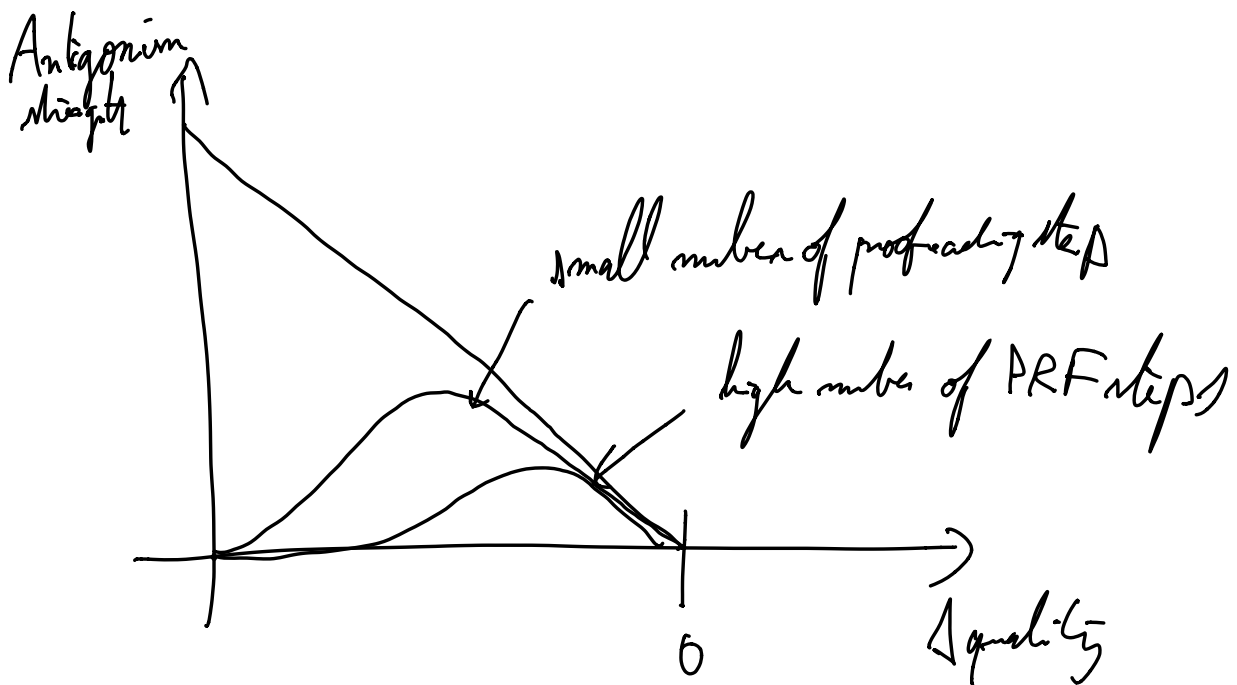
$$\dot{R} = 1 - k \sum L_i R$$

$$\dot{D}_i = k L_i R - \mu_i^{-1} D_i$$

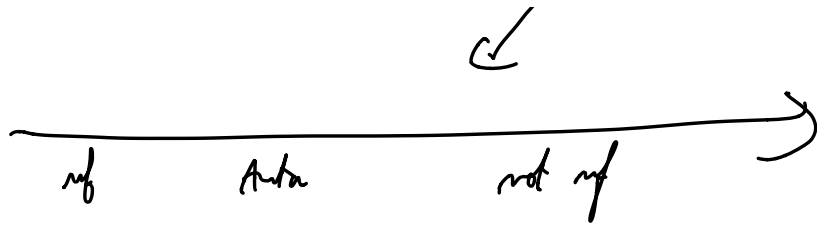
$$\sum D_i = \left(\sum \mu_i^{-1} L_i \right) R = \frac{\sum \mu_i^{-1} L_i}{\sum L_i} R$$

=> average of μ_i

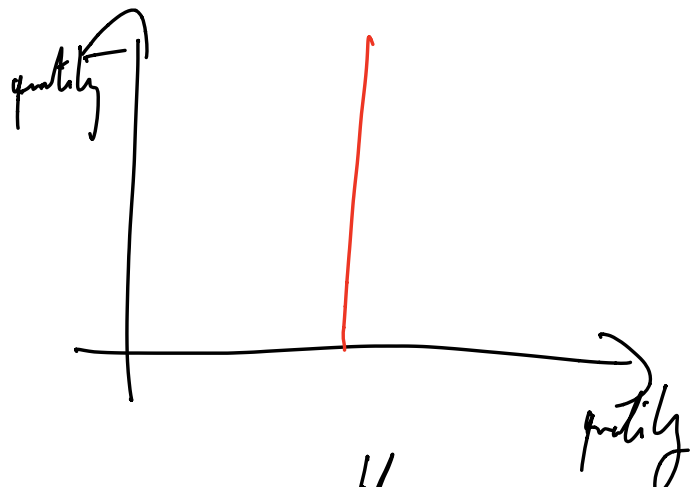
Mean antagonism for small μ_i



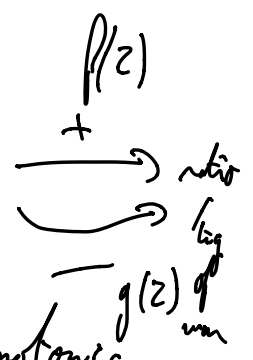
No antagonism of self \equiv highly nonlinear and not precise



Further

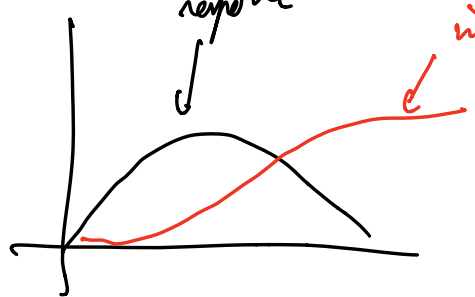


probably very generic theory
 ("preaching" for ligands)



General tests: via antagonism and/or non monotonic

dose response curve
 response



internal negative feedback

↓
 or induced
 by weaker but
 more abundant
 ligands

● Torigoe et al, Science 1998

FC ERI_D (other kind of mGluR receptor)

antagonism via deactivation of kinase \Rightarrow exactly AS^V

• Reduction of model

Prinz - Girdleau et al,

Biophys J 2017

Concluding remarks

Evolution or ... computer aided network discovery

- coarse-grained fitnesses
 - Many "phenotypes" come for free. This is probably the same in nature.
 - Simple networks give rich dynamics (see also Kimistry)
 - hypothesis, mechanism
- We were kind of lost in the immense haystack, now we have the "good" class of models