

Non-Mass Action Modeling for the Binding of Phosphorylated Gli1 with Sufu

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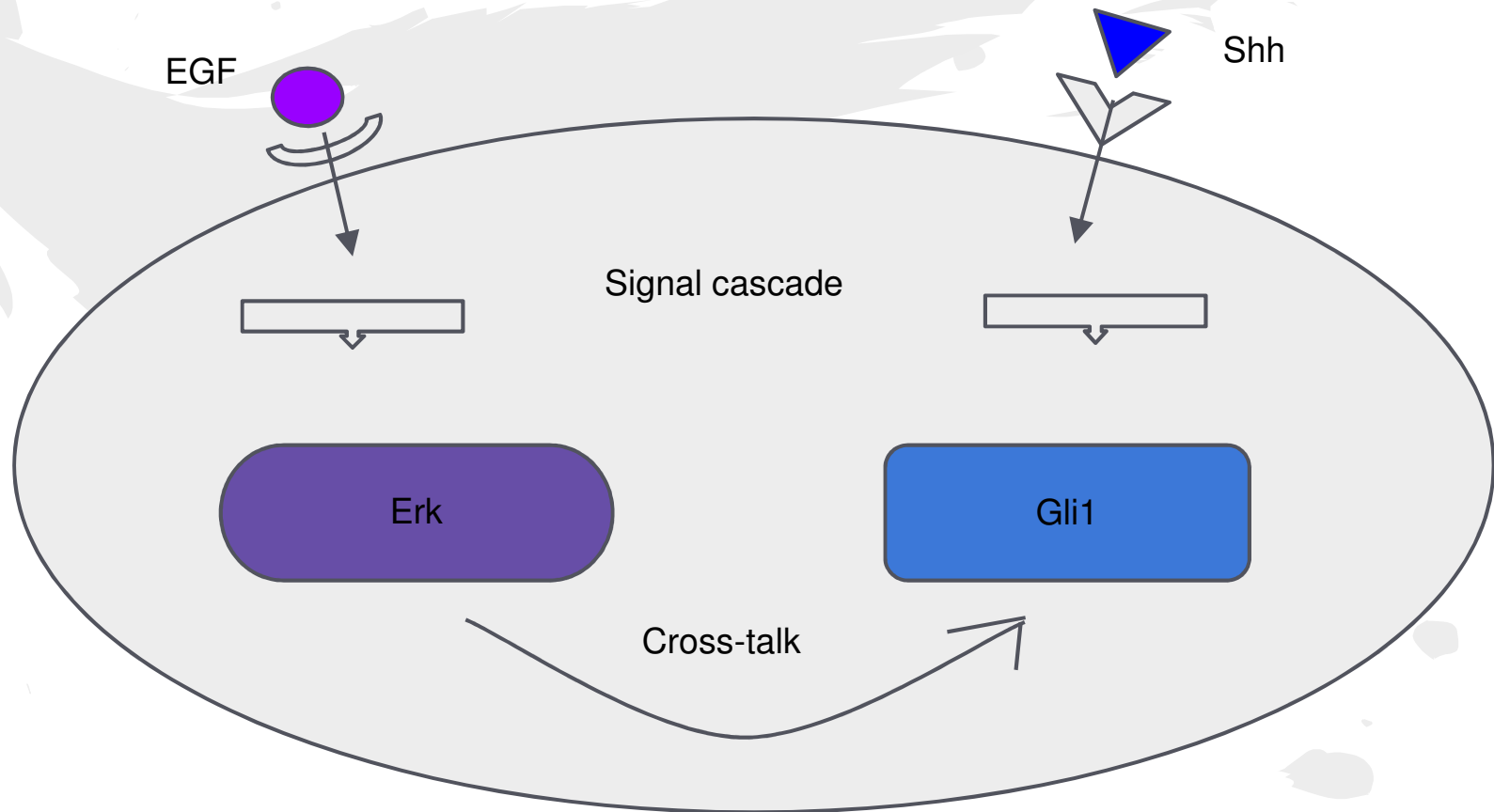
**What is
this?**



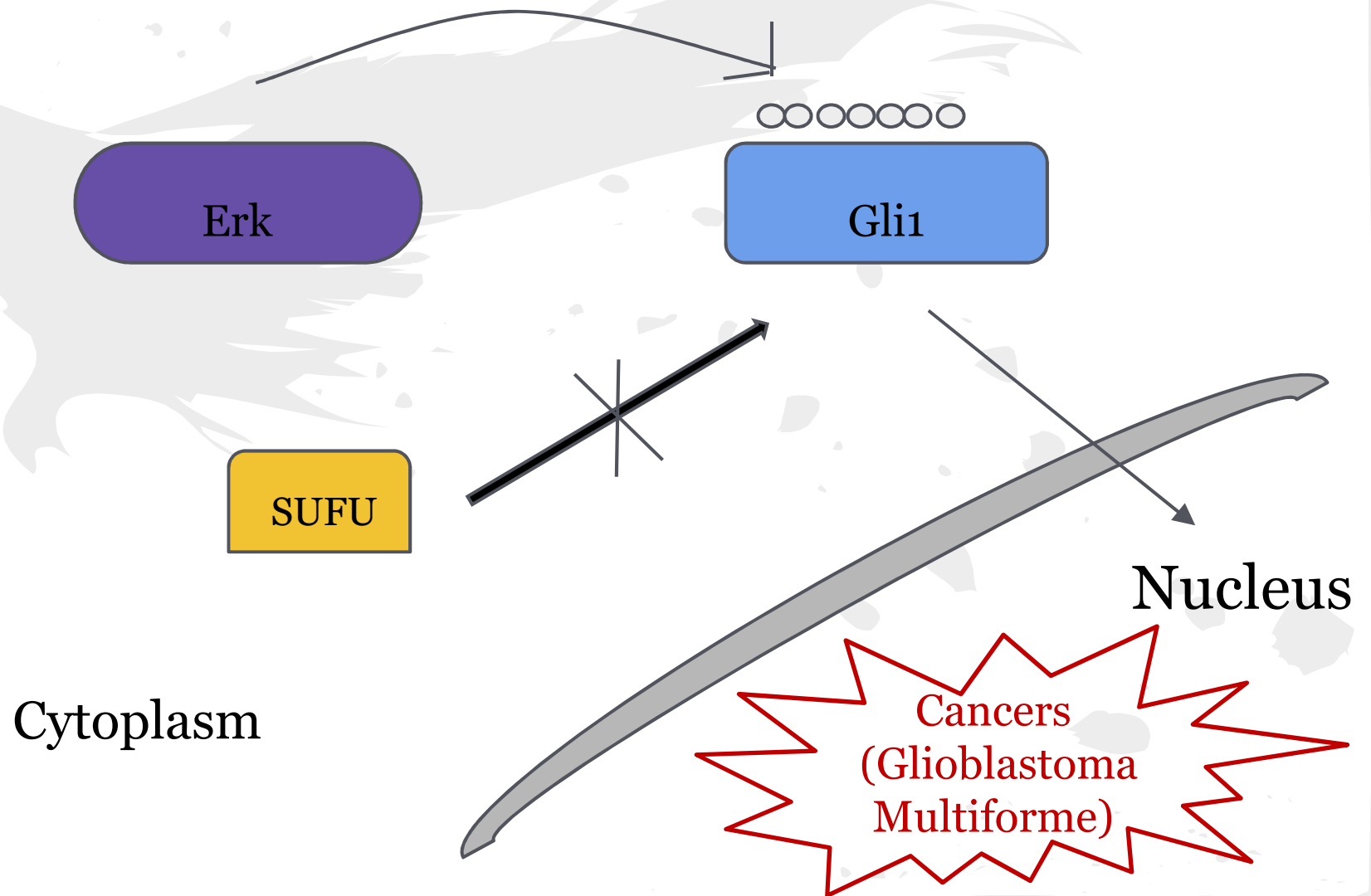
Hedgehog Pathway!



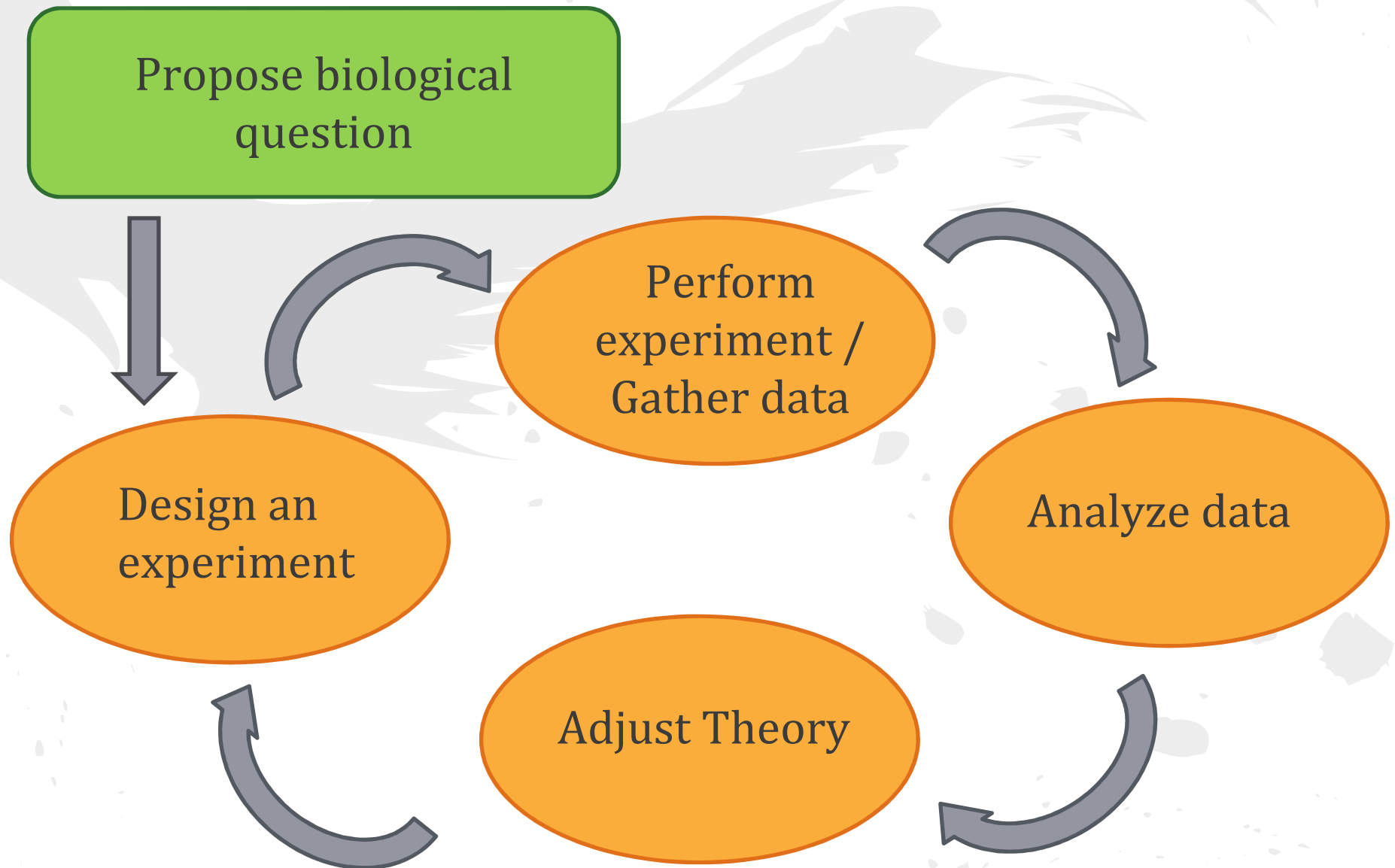
Hedgehog & MAPK Cross-Talk



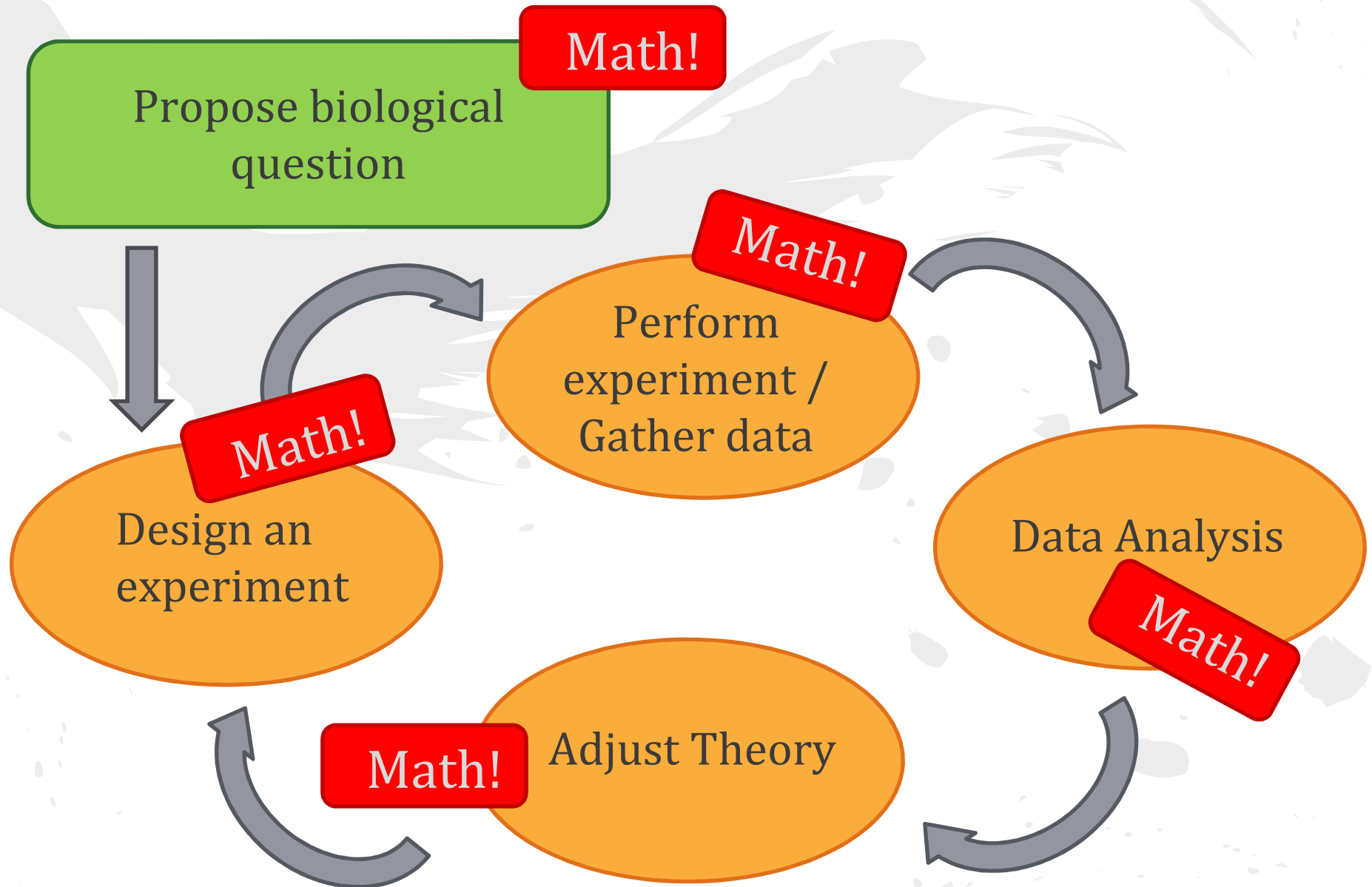
The Biological Question



Where is the math?



Where is the math?

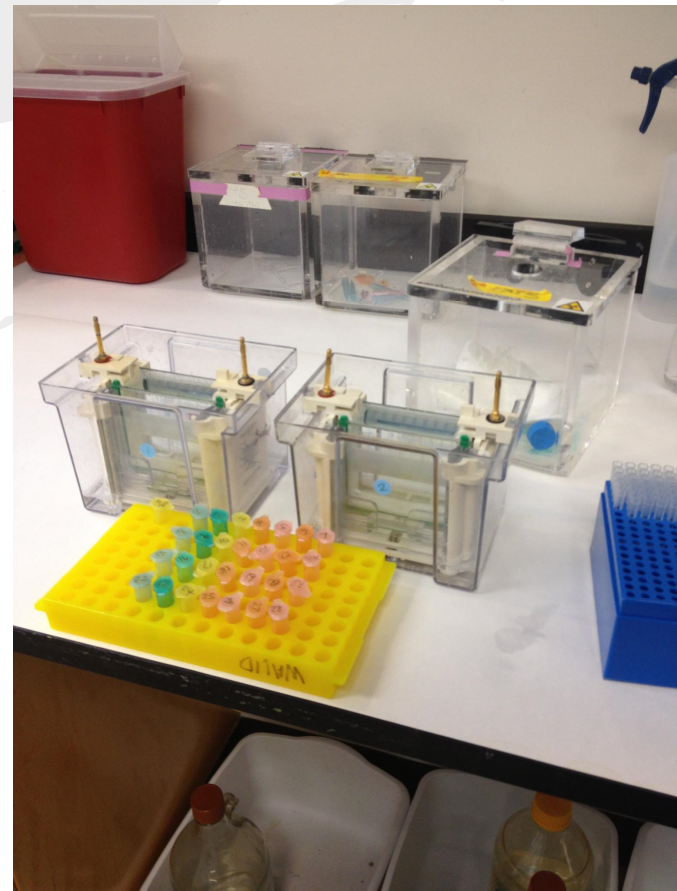


Our Research

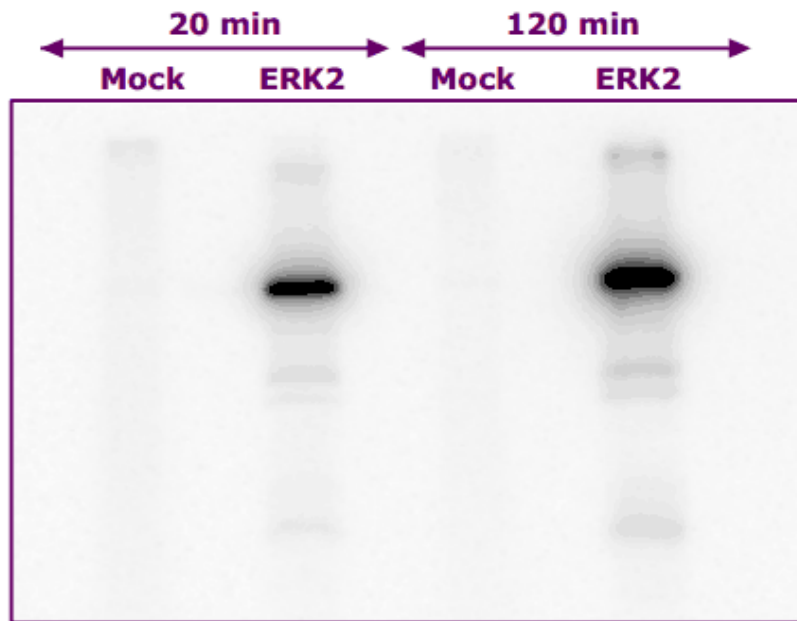
In order to understand the Erk2-Gli1-Sufu system we propose a novel method of biochemical modeling using Holling Type-II non-linear interactions from ecology.

Experimental Design – Bardwell Lab

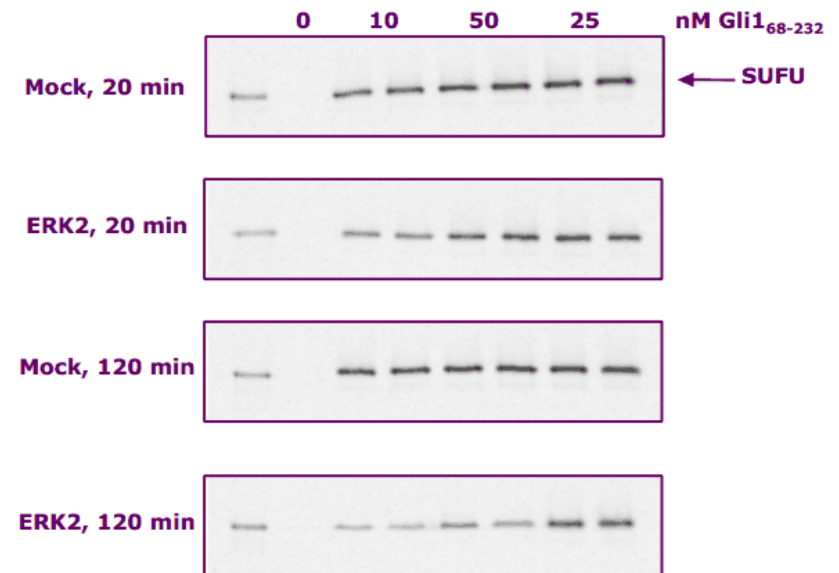
- Want Gli1₆₈₋₂₃₂ (has Sufu binding site)
 - CDNA Cloning
 - Radioactive tagging Sufu
- Titrations of Gli1 and Erk1
- Analyze with protein binding assays
 - Gel electrophoresis



Data Collection



Gli1₆₈₋₂₃₂ kinase with Erk2



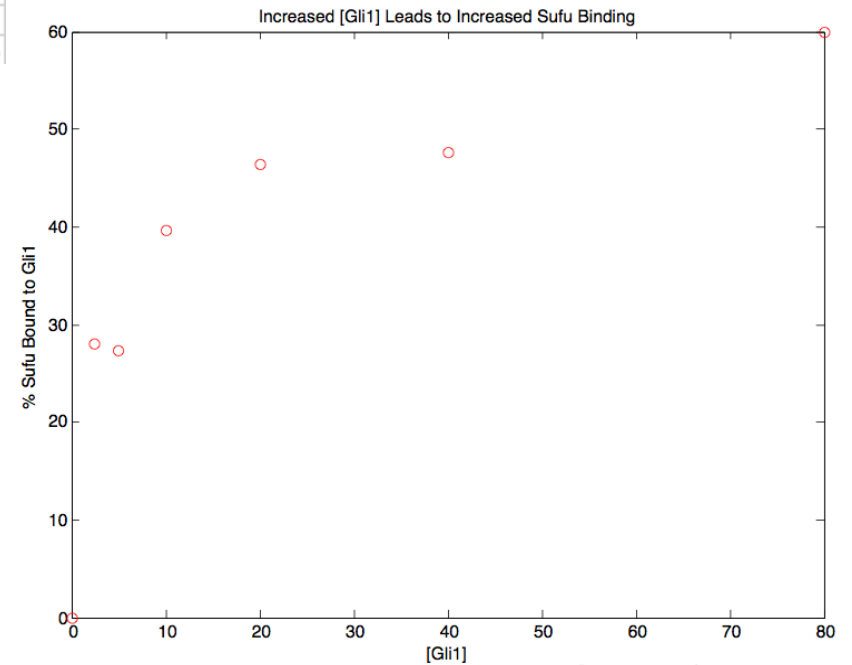
Gli1₆₈₋₂₃₂ binding to Sufu

Data Translation

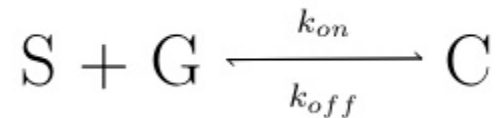
Mock, 20 min



Substrate	Treatment	Time	nM Gli1 ₆₈₋₂₃₂	raw	bkqd adj	% pptd
GLI1 ₆₈₋₂₃₂	Mock	120 min	10	1950344	1707458	43.12
GLI1 ₆₈₋₂₃₂	ERK2 10 u	120 min	20	1743213	1500327	37.89
GLI1 ₆₈₋₂₃₂	ERK2 20 u	120 min	80	1928816	1685930	42.58
GLI1 ₆₈₋₂₃₂	ERK2 50 u	120 min	2.5	487568	244682	6.18
GLI1 ₆₈₋₂₃₂	ERK2 100 u	120 min	0	246405	3519	0.09
					3959758	100%



Initial Model: Mass-Action



$S = [\text{Sufu}]$

$G = [\text{Gli1}]$

$C = [\text{Bound Complex}]$

$k_{on}, k_{off} = \text{rate constants}$

$$G' = -k_{on}SG + k_{off}C$$

$$S' = -k_{on}SG + k_{off}C$$

$$C' = k_{on}SG - k_{off}C$$

Steady State Analysis: Mass Action

$$G' = -k_{on}SG + k_{off}C = 0$$

$$-k_{on}(S_t - C)G + k_{off}C = 0$$

$$-(S_t - C)G + k_dC = 0$$

$$k_dC = (S_t - C)G$$

$$k_dC + GC = S_tG$$

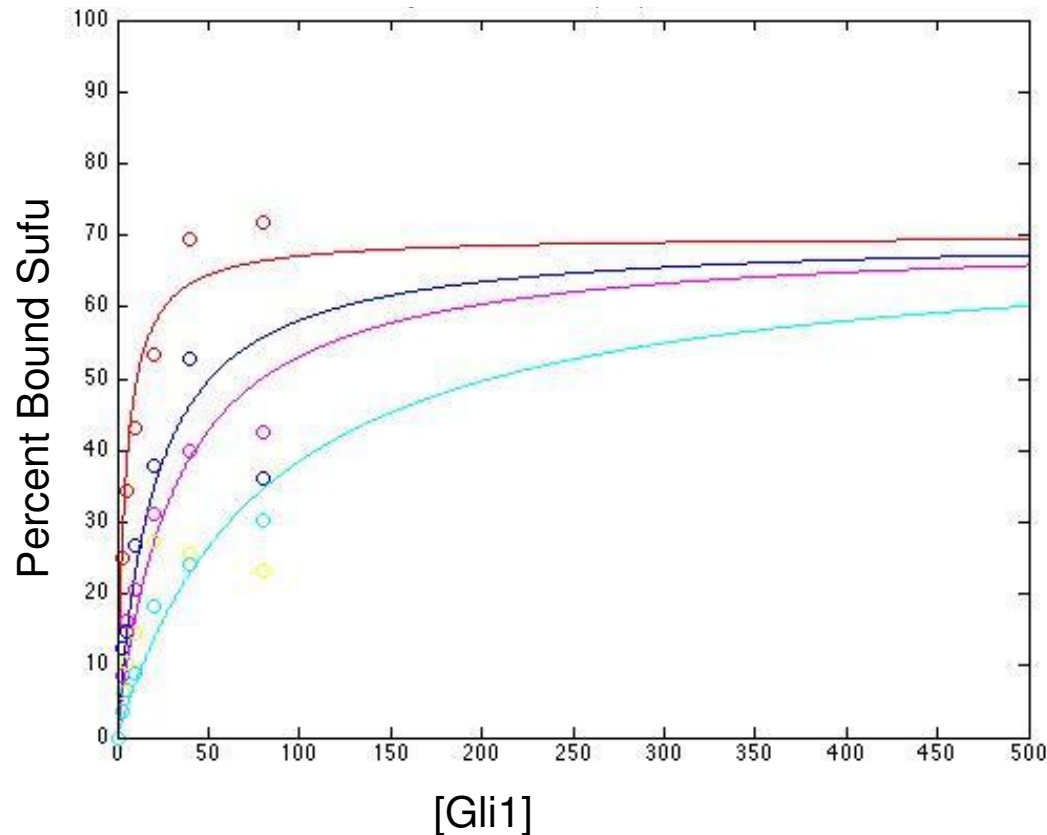
$$C(k_d + G) = S_tG$$

$$C = \frac{S_tG}{(k_d + G)}$$

$$\%SufuBound = \frac{C}{S_t} = \frac{G}{(k_d + G)}$$

Curve fitting: Mass-Action

Sufu binding to ERK 100u Phosphorylated Gli1 10min



Equal saturation: $R > 0.64$

Variable saturation: $R > 0.93$

Least-squares fit
(Matlab) on:

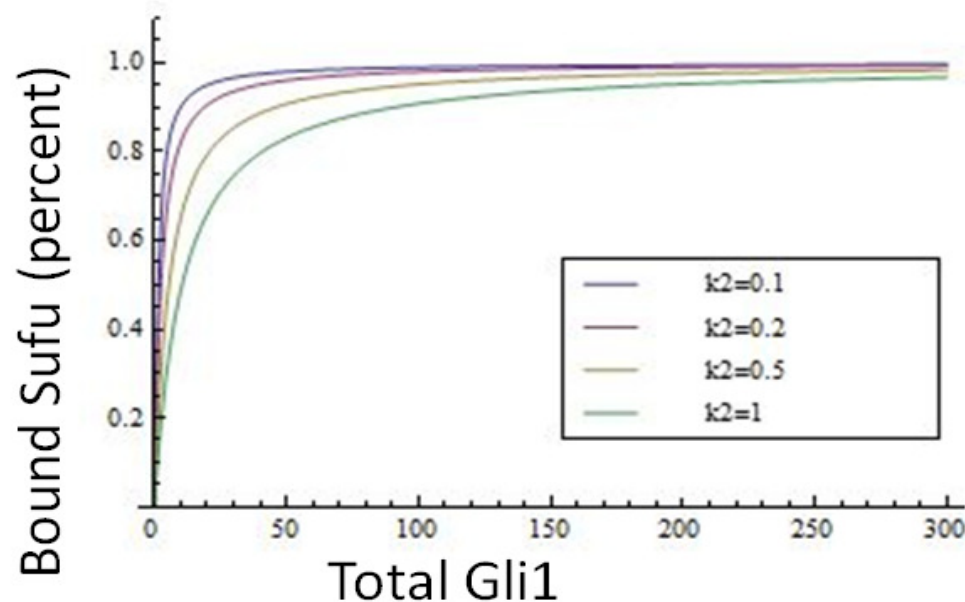
$$\frac{G}{(k_d + G)}$$

This requires forced
saturation level,
“ S_{\max} ”, to be
identified. We tested

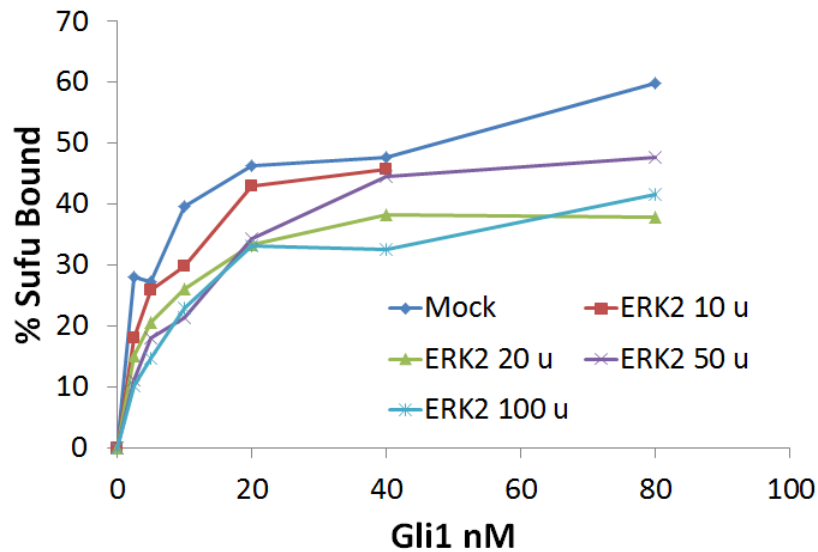
- Equal saturation
- Variable saturation

Mass-Action Dynamics

- Assumes all curves saturate to the same level, 100% of S_{\max}
- Occurs regardless of k_{on} , k_{off} parameters

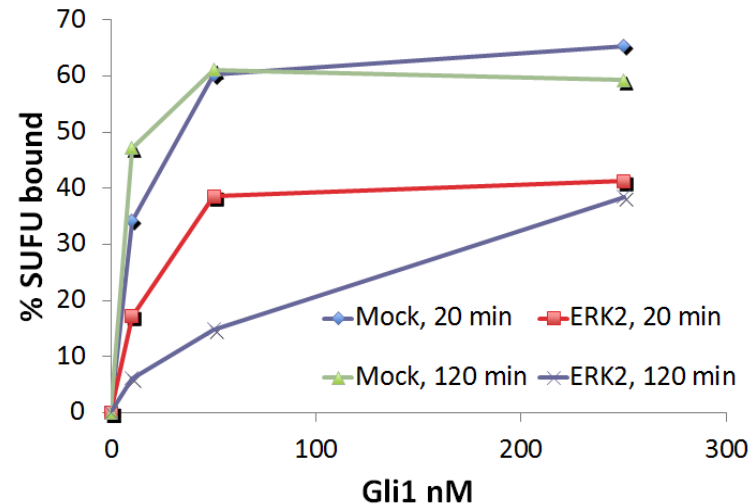


Identifying Saturation Levels



Directed Experiment:
find saturation levels
decrease when Gli1
phosphorylated by Erk.
Consistent through time

Original data:
Might all saturate at
equal levels



Explaining Saturation Levels

- Simple mass-action is not reasonable
- Try dimerization model:
 - Same basis as simple mass-action, however Gli1 is able to form a dimer, D
 - Hope Gli1 dimerization lowers binding with Sufu



$$S' = -k_1 SG + k_2 C$$

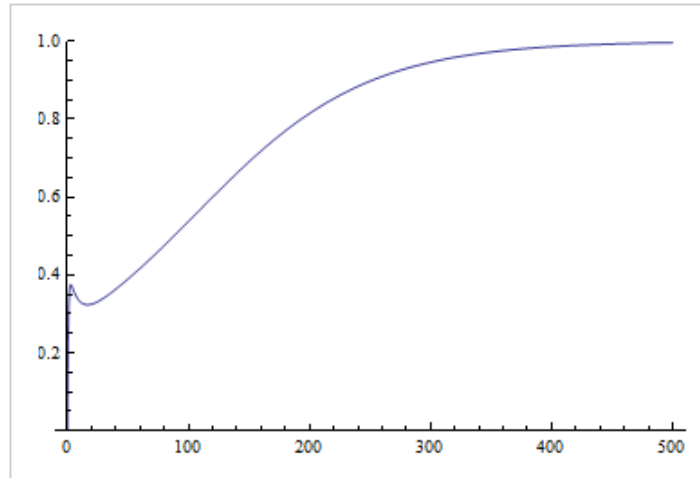
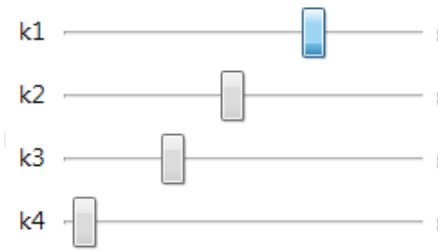
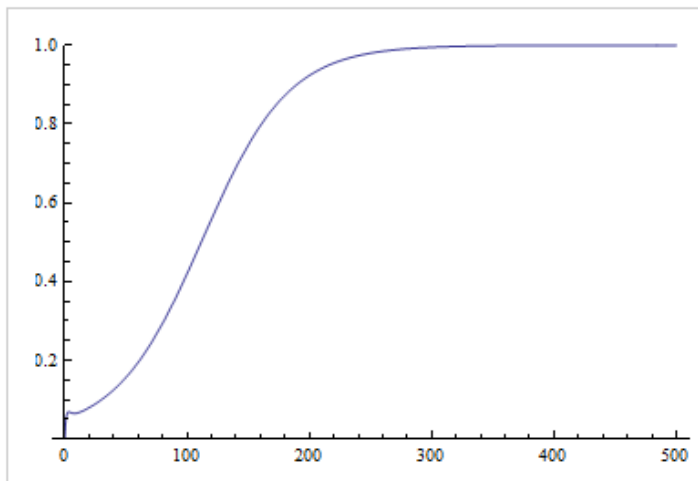
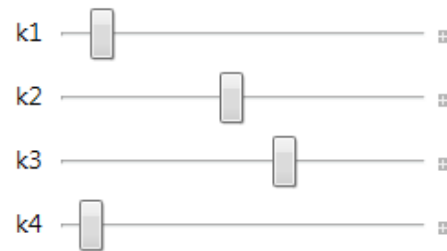
$$G' = -k_1 SG - k_3 G^2 + 2k_4 D + k_2 C$$

$$C' = -k_2 C + k_1 SG$$

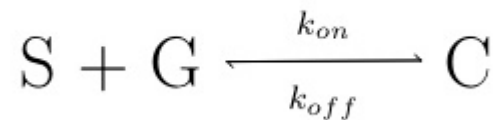
$$D' = -k_4 D + k_3 G^2$$

Steady-State Analysis: Dimerization

- Curve shape changes, but saturation remains at 100%
- Not a sufficient explanation of varying saturation levels



Proposed Solution: Non-Linear Dynamical Model



$$G' = S' = -k_{on} \frac{G}{1 + \tau k_{on} G} S + k_{off} C$$

$$C' = -k_{off} C + k_{on} \frac{G}{1 + \tau k_{on} G} S$$

- General form remains the same
- Gli1 and Sufu binding rate taken to be non-linear
- Binding rate changes based on τ , time delay between Gli1-Sufu initial interaction and successful binding

Steady-State Analysis: Non-Mass Action

$$S' = -k_{on} \frac{G}{1 + \tau k_{on} G} S + k_{off} C$$

...

Equations are a little bit more complex:

$$C = S_t \frac{G_t k_{on} + k_{off} + k_{on} S_t + G_t k_{on} k_{off} \tau \pm \sqrt{4 G_t k_{on} S_t (-k_{on} - k_{on} k_{off} \tau) + (G_t k_{on} + k_{off} + k_{on} S_t + G_t k_{on} k_{off} \tau)^2}}{2(k_{on} + k_{on} k_{off} \tau)}$$

$$\%SufuBound = \frac{G_t k_{on} + k_{off} + k_{on} S_t + G_t k_{on} k_{off} \tau + \sqrt{4 G_t k_{on} S_t (-k_{on} - k_{on} k_{off} \tau) + (G_t k_{on} + k_{off} + k_{on} S_t + G_t k_{on} k_{off} \tau)^2}}{2(k_{on} + k_{on} k_{off} \tau)}$$

Steady-State Analysis: Non-Mass Action

$$S' = -k_{on} \frac{G}{1 + \tau k_{on} G} S + k_{off} C$$

...

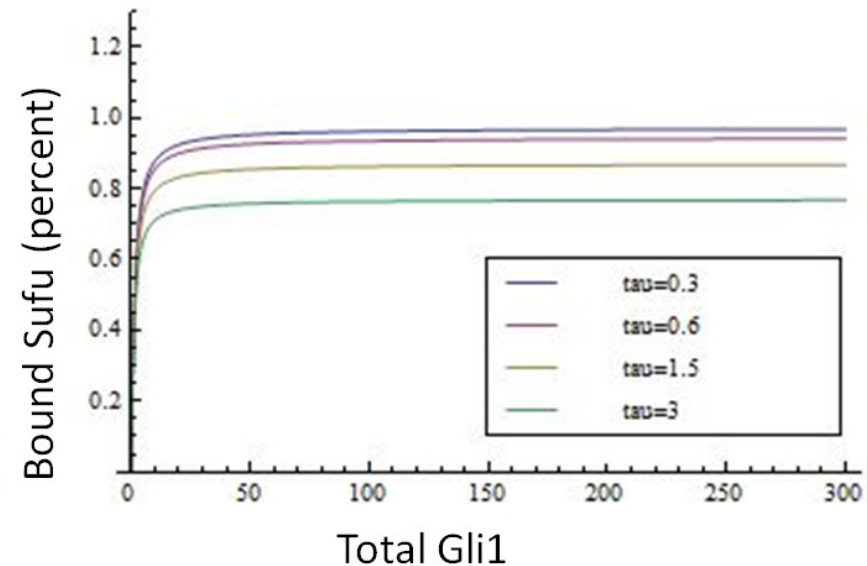
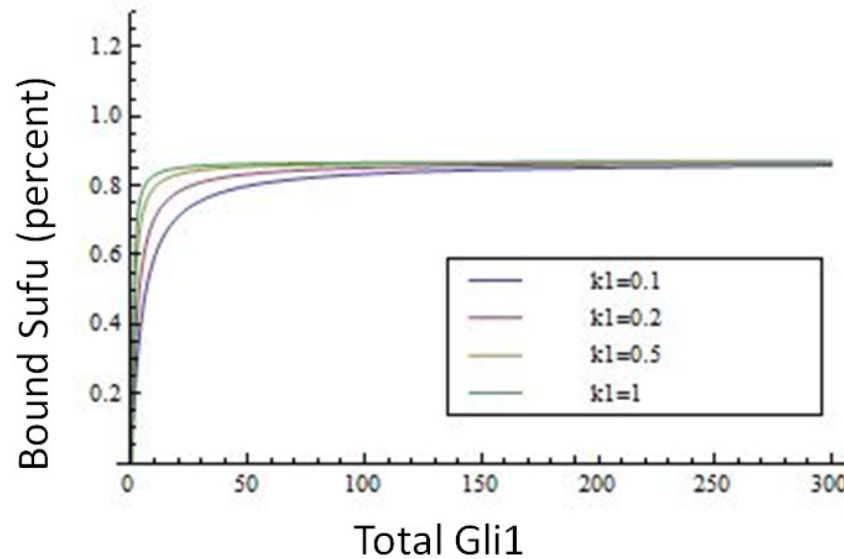
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Notice total Sufu becomes important, along
with individual k_{on} , k_{off}

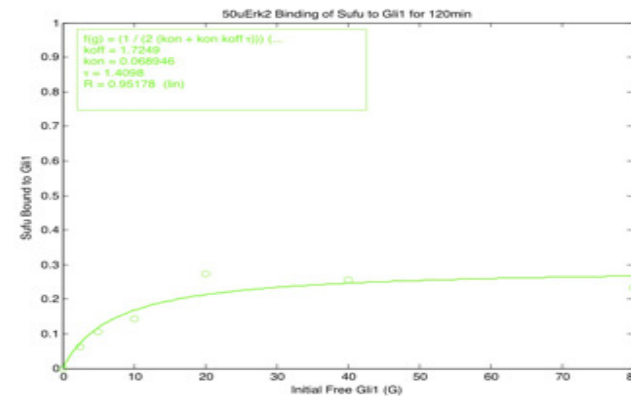
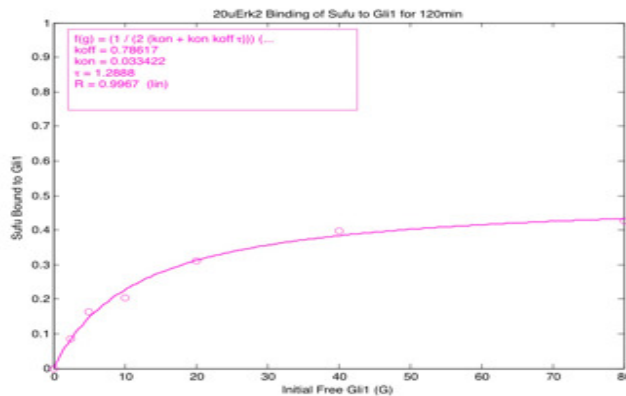
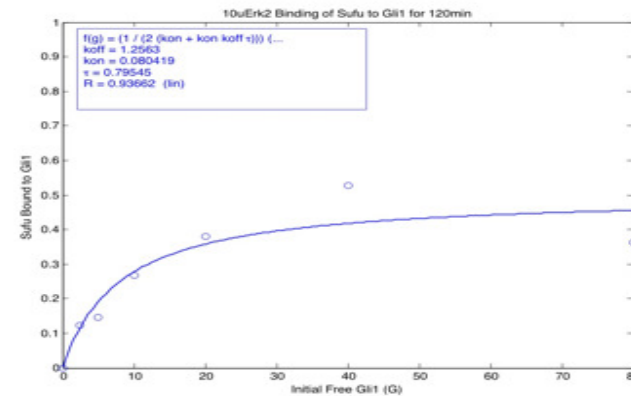
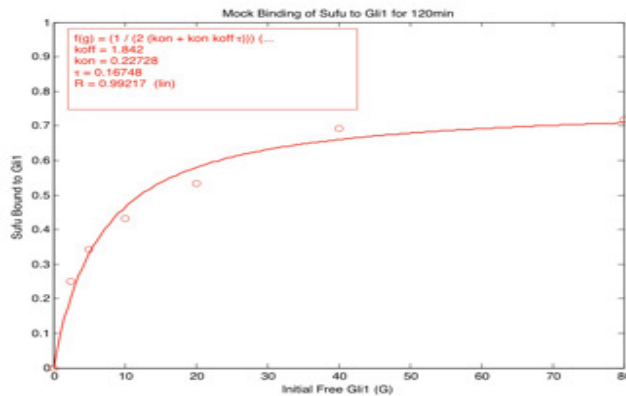
Steady-State Dynamics



- Non-linear system allows for variable saturation levels. Increasing τ lowers total saturation level bound Sufu. Similar dynamics occur for k_{off} . Increasing k_{on} results in saturation at lower levels total Gli1
- Saturation occurs at $\frac{S_{total}}{1+k_{off} \tau}$

Curve fitting: Non-Mass Action

Percent bound Sufu



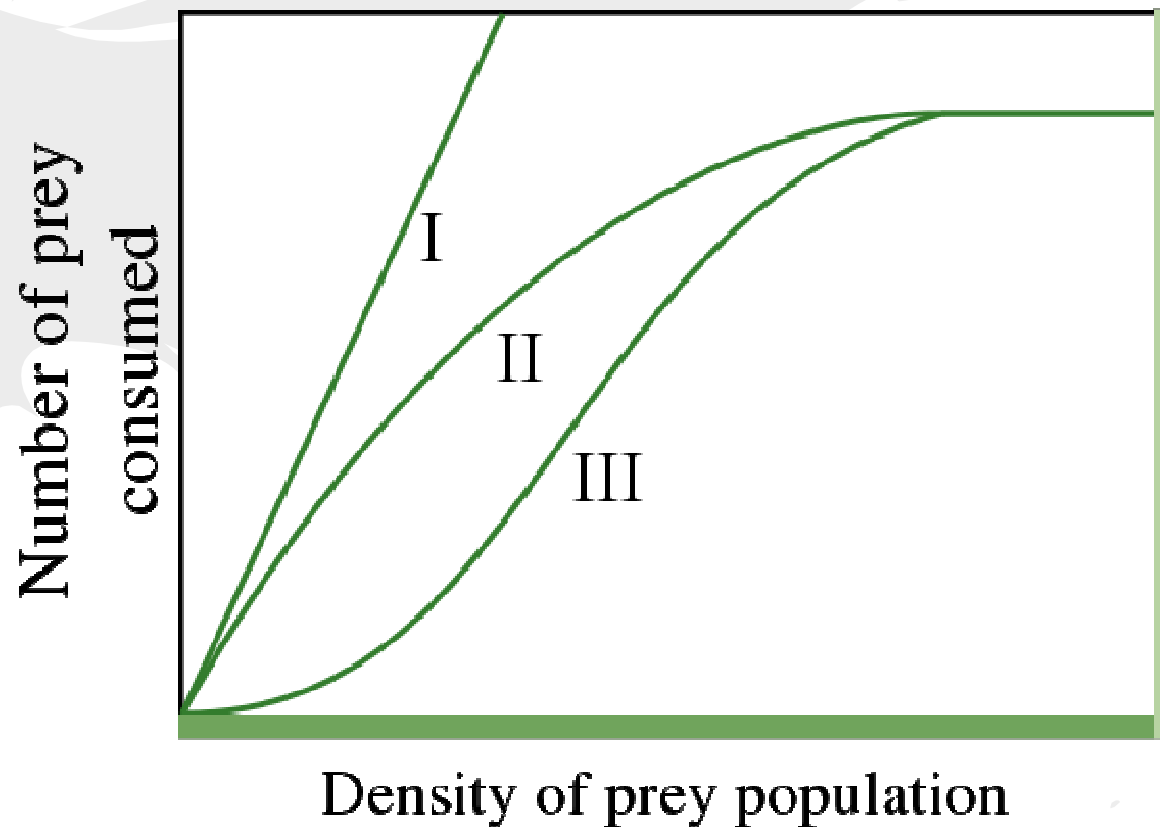
($R > 0.94$
for all)

Concentration Gli1 total

$$\frac{G_t k_{on} + k_{off} + k_{on} S_t + G_t k_{on} k_{off} \tau + \sqrt{4 G_t k_{on} S_t (-k_{on} - k_{on} k_{off} \tau) + (G_t k_{on} + k_{off} + k_{on} S_t + G_t k_{on} k_{off} \tau)^2}}{2(k_{on} + k_{on} k_{off} \tau)}$$

Validity of Non-Linear Model

- Holling Type-II functional response
 - Mass-Action is Holling Type-I



Validity of Non-Linear Model

- Binding time = handling time
- Sufu can only bind and unbind at a certain maximal rate
 - binding is not instantaneous
- Introduce time-delay to model this

Δt : some time interval

y : # binding per sufu molecule in time interval Δt

$$r = \frac{y}{\Delta t} S$$

Non Mass-Action

τ : binding time per sufu molecule

$$y = k_{on} G (\Delta t - y \tau)$$

...

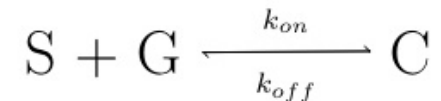
$$y = \frac{k_{on} G}{1 + k_{on} G \tau} \Delta t$$

$$r = \frac{1}{\tau} \frac{G}{\frac{1}{k_{on} \tau} + G} S$$

$$= \frac{1}{\tau} \frac{G}{\frac{1}{k_{on} \tau} + G} S$$

$$= -k_{on} \frac{G}{1 + k_{on} \tau G} S$$

Binding Rate Comparison



Mass-action binding rate:

$$r = k_{on}SG \longrightarrow G' = -k_{on}SG + k_{off}C$$

Proposed non-linear binding rate:

$$r = k_{on} \frac{G}{1 + k_{on}\tau G} S \longrightarrow G' = -k_{on} \frac{G}{1 + k_{on}\tau G} S + k_{off}C$$

Conclusions

- Phosphorylation of Gli1 by Erk2 lowers total Sufu-Gli1 binding
- Biochemical interactions of phosphorylated Gli1-Sufu are more complex than is accounted for by mass action
- Modeling of Sufu-Gli1 binding based on Holling Type-II rate dynamics allows for variable saturation levels at steady-state, consistent with biological data
- Total Gli1-Sufu bound saturation is $\frac{S_{total}}{1+k_{off}\tau}$
- The interaction of Gli1-Sufu is not instantaneous and phosphorylation of Gli1 by Erk2 increases time delay, τ
- Cross-disciplinary studies can bring new insight, even when within a field

Future Direction

- Gli1 is a multisite protein – we believe this is likely important for Erk2/Sufu binding
 - Possible phosphorylation of given sites affects subsequent binding to Sufu
- Multisite dynamics experiments are currently in process [Bardwell Lab]
 - We would like to extend our analysis to a multisite phosphorylation model of Gli1 with Erk2 and Sufu to incorporate this data.
 - In particular, we hope to be able to fit this data to a concerted, redundant activation (CRA) model recently developed by G. Enciso. [submitted June 11, 2013]

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- The Mathematical Biosciences Institute

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References

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J.H.P. Dawes, M.O. Souza, A derivation of Holling's type I, II and III functional responses in predator–prey systems, Journal of Theoretical Biology, Volume 327, 21 June 2013, Pages 11-22, ISSN 0022-5193,
<http://dx.doi.org/10.1016/j.jtbi.2013.02.017>.

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