

# Emergence of bimodal cell population responses from the interplay between analog single-cell signaling and protein expression noise

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Birtwistle *et al.* *BMC Systems Biology* 2012, **6**:109  
<http://www.biomedcentral.com/1752-0509/6/109>

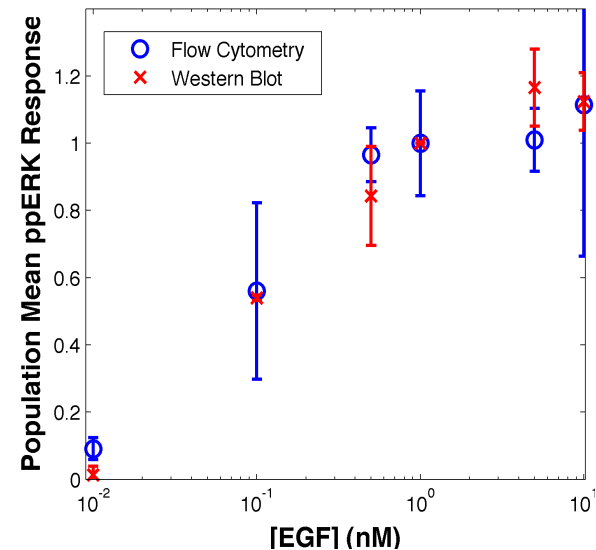
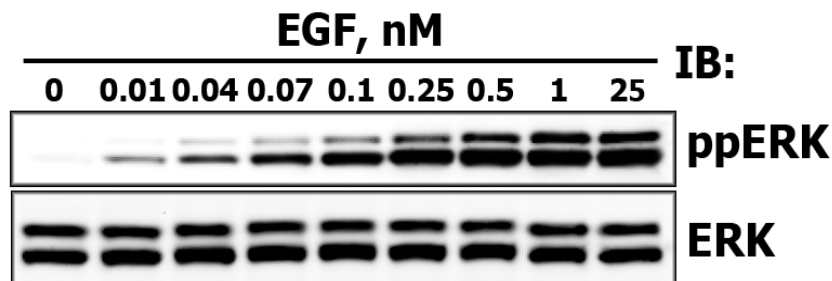
Presented by Sanjana Gupta

# Background

- Individual cells need to be able to convert noisy, analog signals into clear, yes-or-no cell fate decisions.
- Because of heterogeneity in protein abundance, population average measurements are not sufficient for investigating “all or nothing” responses.
- Single cell measurement techniques capable of capturing the dynamics of digital signal transduction are needed.

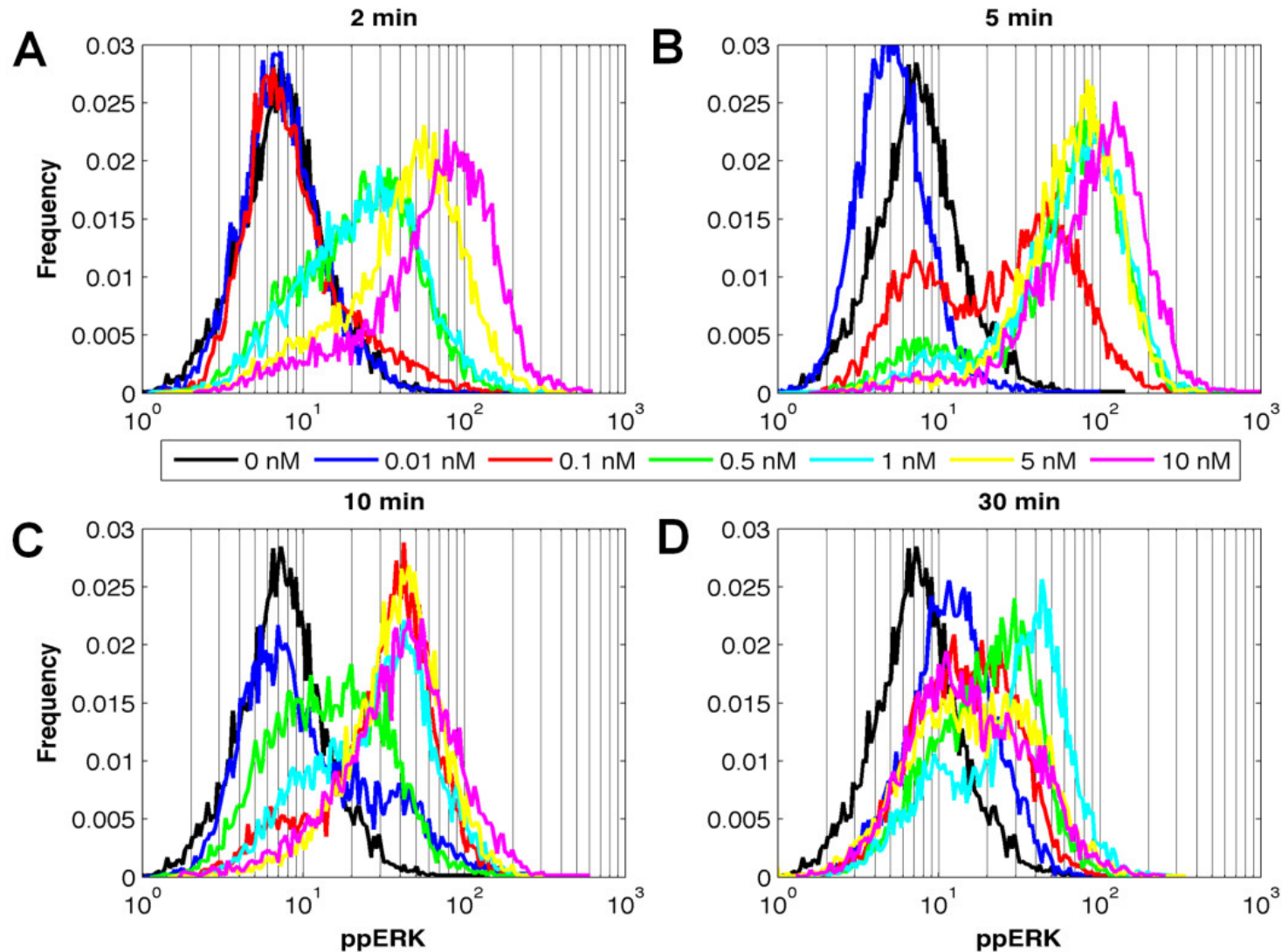
# Analyses of ERK responses to EGF in individual cells and populations

- Flow cytometry-based phosphorylation assay to determine the kinetics and dose response of ERK activation by EGF in HEK293 cells.



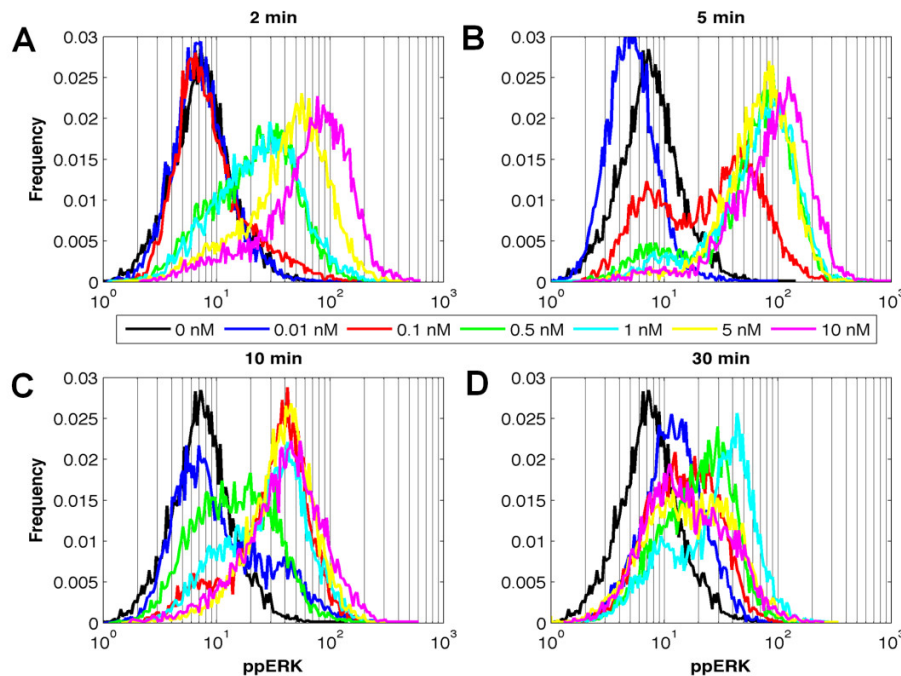
# How do individual cells contribute to this collective population response?

Figure 1, Birtwistle et al.



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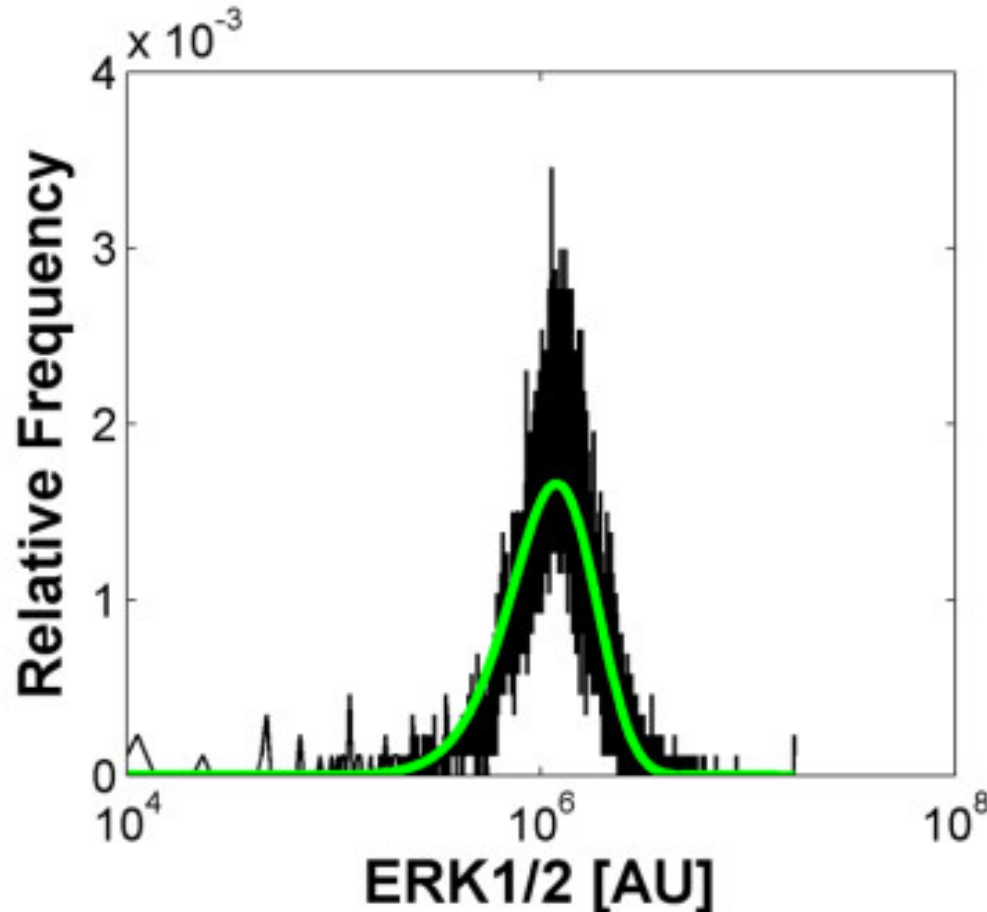
Figure 1, Birtwistle et al.



- Increase in mean values of ppERK was dose-dependent: **Analog signaling**
- A fraction of cells contain ppERK levels similar to those of the basal state.
- The height of this shoulder decreases with increasing EGF dose, but the position remains unchanged.
- **Digital on/off behaviour**

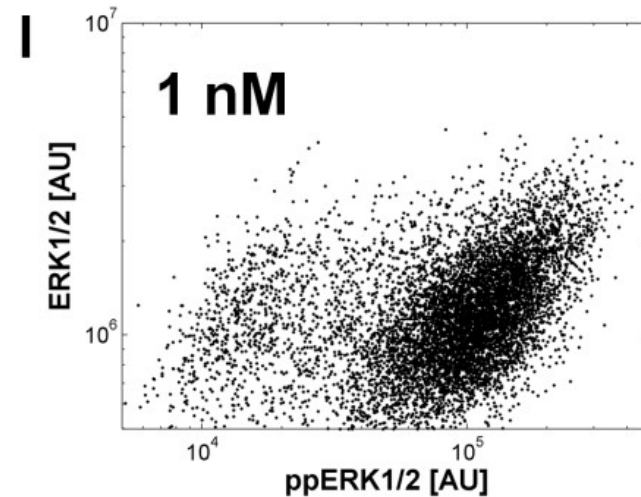
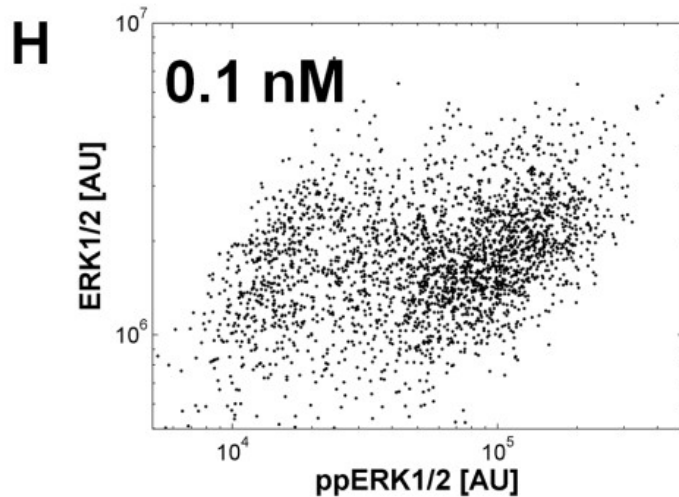
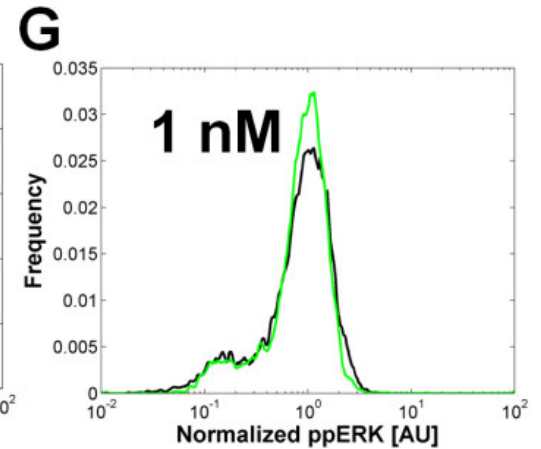
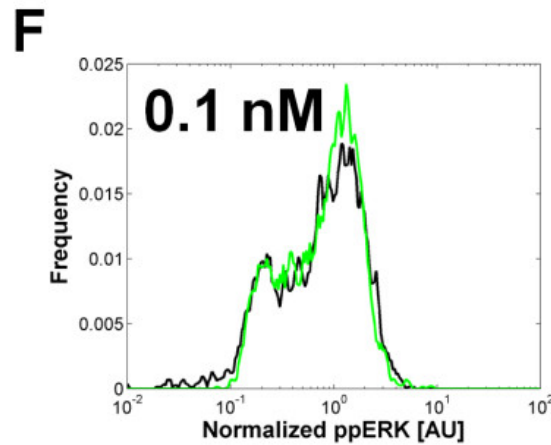
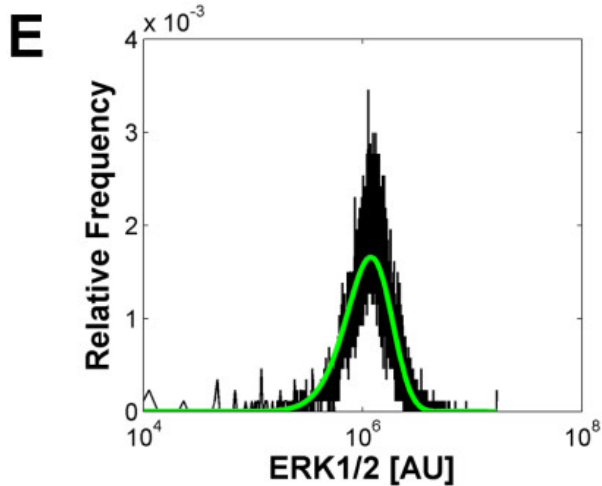
# How does cell to cell variability in total Erk abundance affect the ppERK response?

**E**



- Substantial cell to cell variability in total ERK levels.
- Data is well approximated by a gamma distribution.

# How does cell to cell variability in total Erk abundance affect the ppERK response?

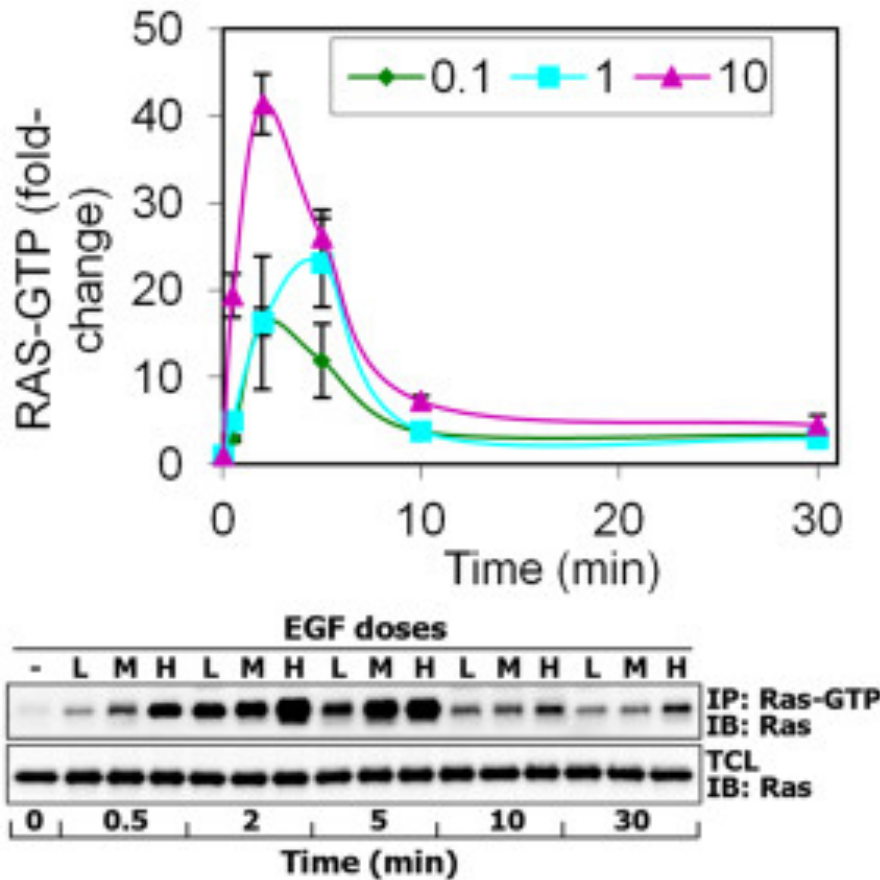


**“Although cell to cell variability in ERK abundance contributes to ppERK response variability, it does not control bimodality, raising the question of what other factors contribute to the observed bimodality”**



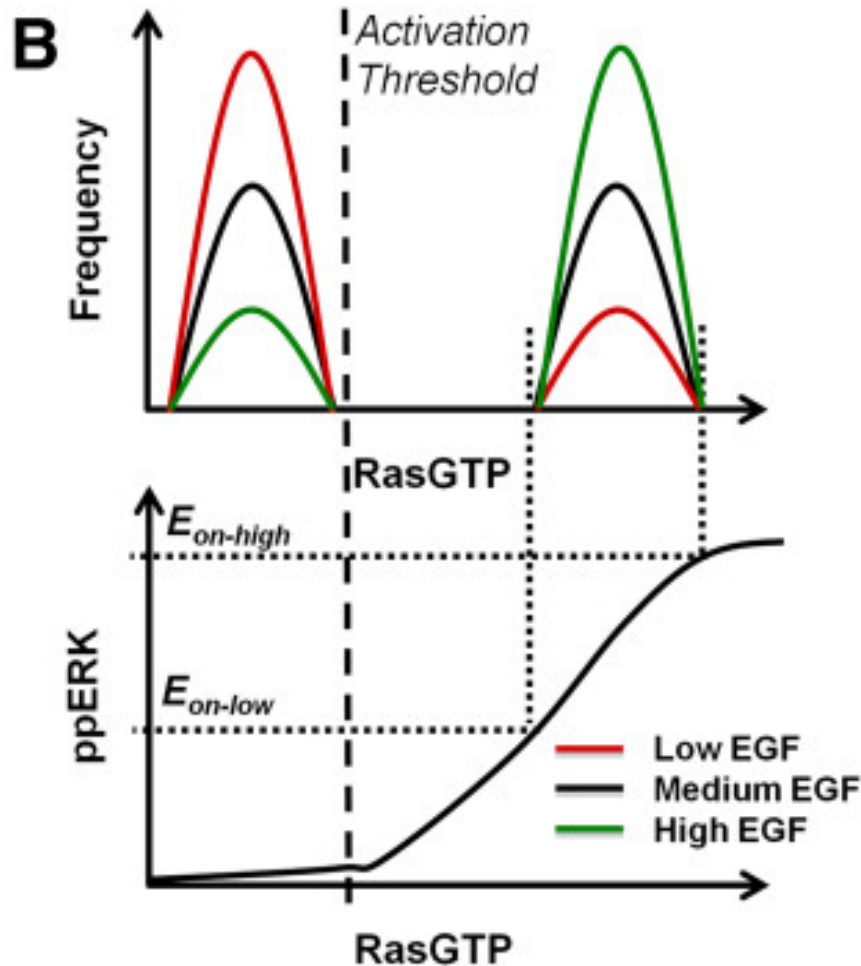
# Stochastic dynamic modeling explanation of the data

- EGF activates RAS, which activates ERK
- The population average dose and dynamic responses of GTP-bound active RAS were assayed via pull-down and Western blotting and then quantified.

**A**

*“A recent study suggested that in T lymphocytes, a positive feedback between RasGTP and its activator guanine exchange factor Son of Sevenless(SOS) leads to bistability and hysteresis in Ras activation”*

# Can bistability in RasGTP explain the bimodal ppERK populations?

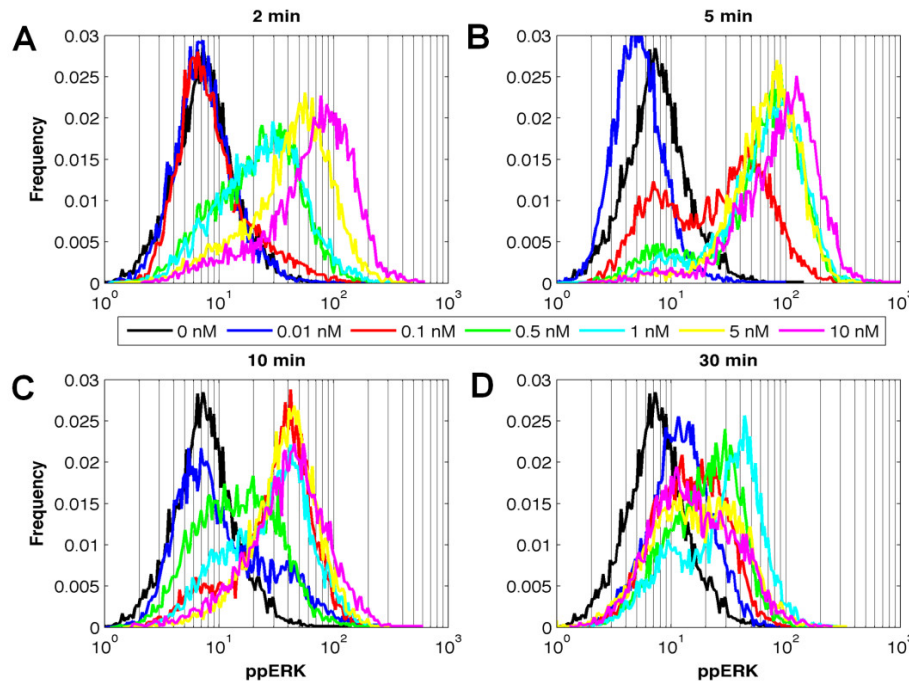


Stimulation by EGF would affect the relative fraction of cells in the two populations, but not their means

A defined high mean RasGTP population would induce a defined high mean ppERK population.

# Can bistability in RasGTP explain the bimodal ppERK populations?

Figure 1, Birtwistle et al.



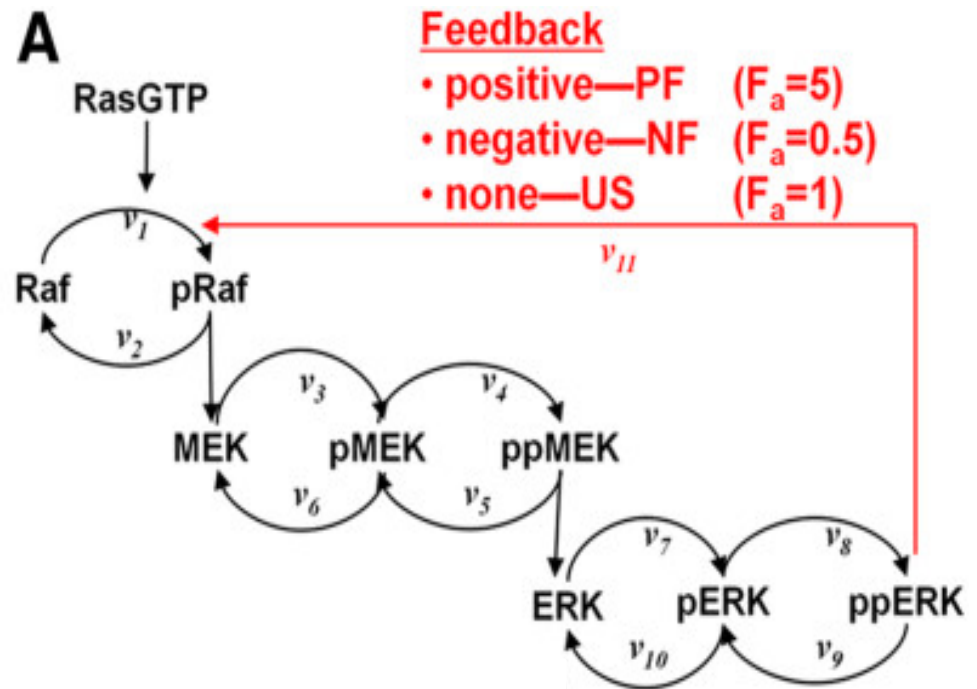
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# Examining different configurations of the MAPK/ERK cascade

Distributions of active ERK display bimodal/shouldering behavior with increasing EGF dose

The “ERK-on” population mean increases with increasing EGF dose at early time points, but decreases with time at constant EGF dose.

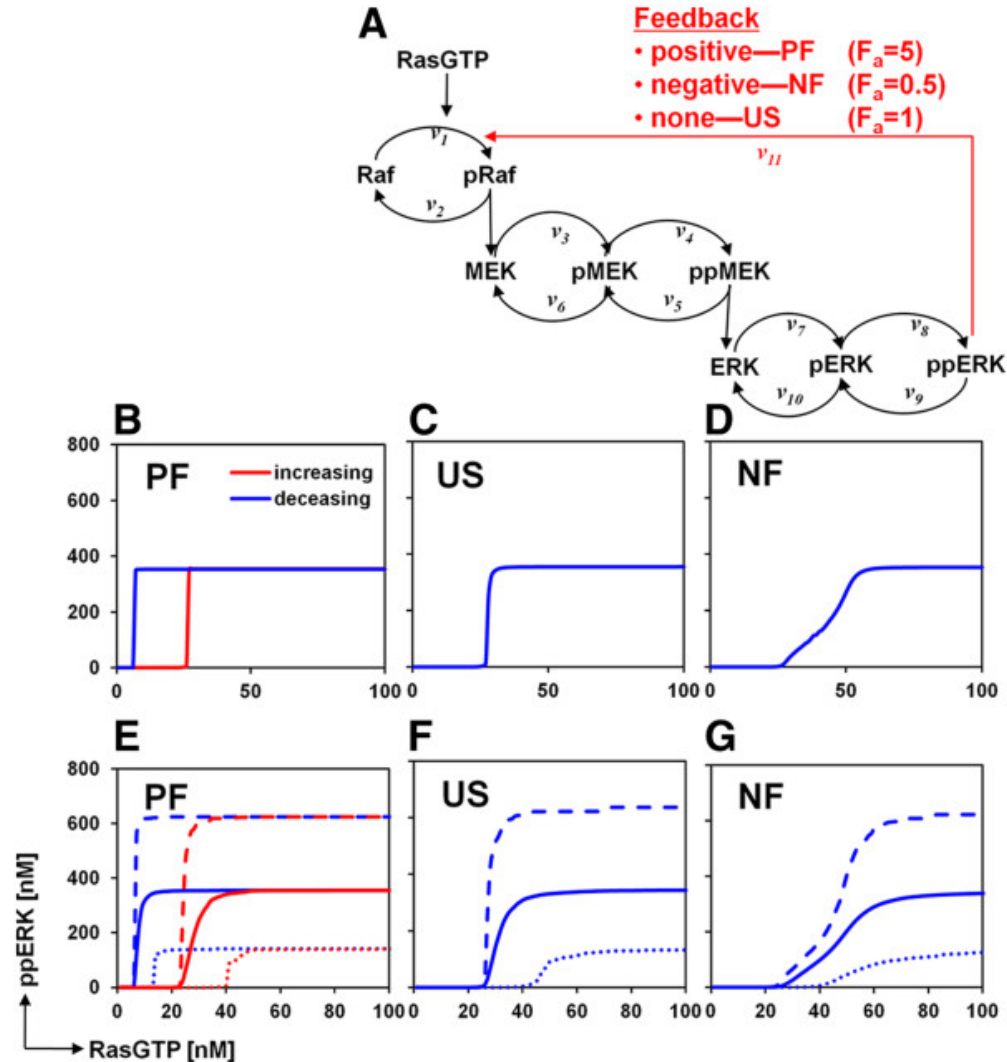


**Table 1 Kinetic description of the ERK signaling cascade**

N	Reaction	Rate	Kinetic constant*
1	MAP3K → pMAP3K	$V_1 = \frac{k_1^{cat} \cdot [Ras-GTP] \cdot [MAP3K]/K_{m1}}{(1+[MAP3K]/K_{m1})} \cdot g(F_a)$	$k_1^{cat} = 0.2; K_{m1} = 50$
2	pMAP3K → MAP3K	$V_2 = \frac{V_{max2} \cdot [pMAP3K]/K_{m2}}{(1+[pMAP3K]/K_{m2})}$	$V_{max2} = 5; K_{m2} = 50$
3	MAP2K → pMAP2K	$V_3 = \frac{k_3^{cat} \cdot [pMAP3K] \cdot [MAP2K]/K_{m3}}{(1+[MAP2K]/K_{m3}+[pMAP2K]/K_{m4})}$	$k_3^{cat} = 1; K_{m3} = 130$
4	pMAP2K → ppMAP2K	$V_4 = \frac{k_4^{cat} \cdot [pMAP3K] \cdot [pMAP2K]/K_{m4}}{(1+[MAP2K]/K_{m3}+[pMAP2K]/K_{m4})}$	$k_4^{cat} = 5; K_{m4} = 50$
5	ppMAP2K → pMAP2K	$V_5 = \frac{V_{max5} \cdot [ppMAP2K]/K_{m5}}{(1+[ppMAP2K]/K_{m5}+[pMAP2K]/K_{m6}+[MAP2K]/K_{i1})}$	$V_{max5} = 250; K_{m5} = 100$
6	pMAP2K → MAP2K	$V_6 = \frac{V_{max6} \cdot [pMAP2K]/K_{m6}}{(1+[ppMAP2K]/K_{m5}+[pMAP2K]/K_{m6}+[MAP2K]/K_{i1})}$	$V_{max6} = 250; K_{m6} = 100; K_{i1} = 80$
7	MAPK → pMAPK	$V_7 = \frac{k_7^{cat} \cdot [ppMAP2K] \cdot [MAPK]/K_{m7}}{(1+[MAPK]/K_{m7}+[pMAPK]/K_{m8})}$	$k_7^{cat} = 1; K_{m7} = 50$
8	pMAPK → ppMAPK	$v_8 = \frac{k_8^{cat} \cdot [ppMAP2K] \cdot [pMAPK]/K_{m8}}{(1+[MAPK]/K_{m7}+[pMAPK]/K_{m8})}$	$k_8^{cat} = 20; K_{m8} = 50$
9	ppMAPK → pMAPK	$V_9 = \frac{V_{max9} \cdot [ppMAPK]/K_{m9}}{(1+[ppMAPK]/K_{m9}+[pMAPK]/K_{m10}+[MAPK]/K_{i2})}$	$V_{max9} = 380; K_{m9} = 10$
10	pMAPK → MAPK	$V_{10} = \frac{V_{max10} \cdot [pMAPK]/K_{m10}}{(1+[ppMAPK]/K_{m9}+[pMAPK]/K_{m10}+[MAPK]/K_{i2})}$	$V_{max10} = 50; K_{m10} = 18; K_{i2} = 100$
11	Feedback	$g(F_a) = \frac{(1+F_a \cdot ([ppMAPK]/K_a)^2)}{(1+([ppMAPK]/K_a)^2)}$	$K_a = 100; F_a = 5; 1; 0.5(\text{PF}; \text{US}; \text{NF})$

\* Maximal rates, Michaelis and catalytic constants are expressed in [nM/s], [nM], and [s<sup>-1</sup>], respectively. Total protein concentrations are [MAPK3]<sub>total</sub> = 200nM, [MAPK2]<sub>total</sub> = 200nM, and [MAPK]<sub>total</sub> = 360nM.

# Steady state analysis

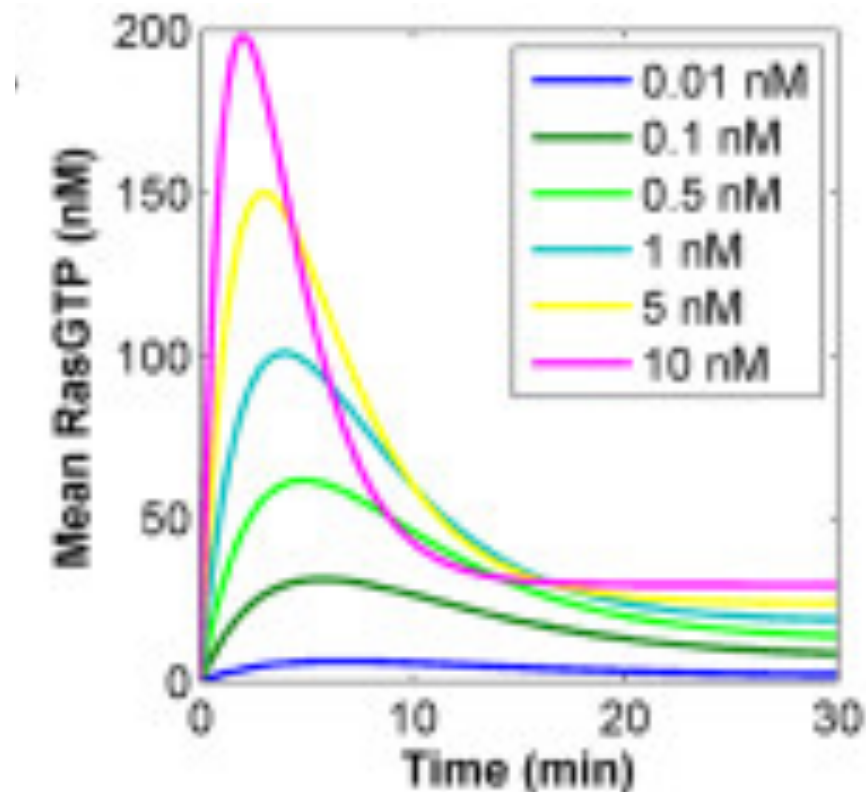


Deterministic  
simulations

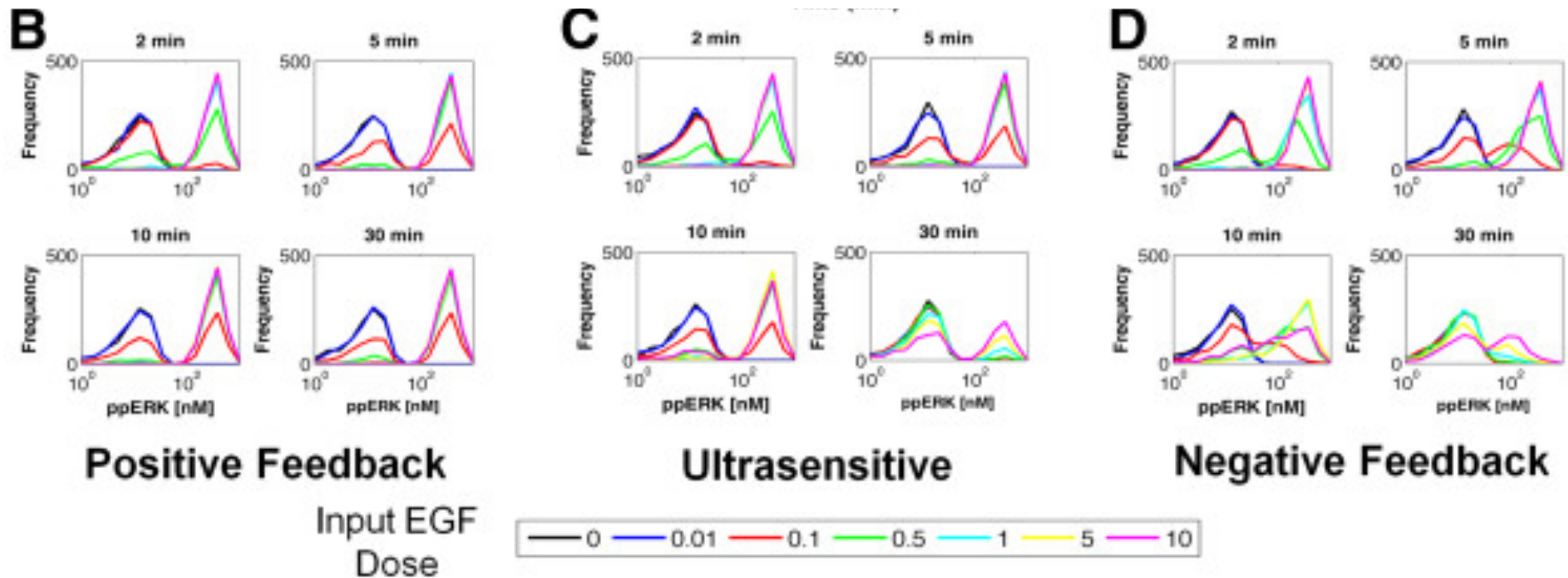
Stochastic  
simulations

# Analysis of transient responses

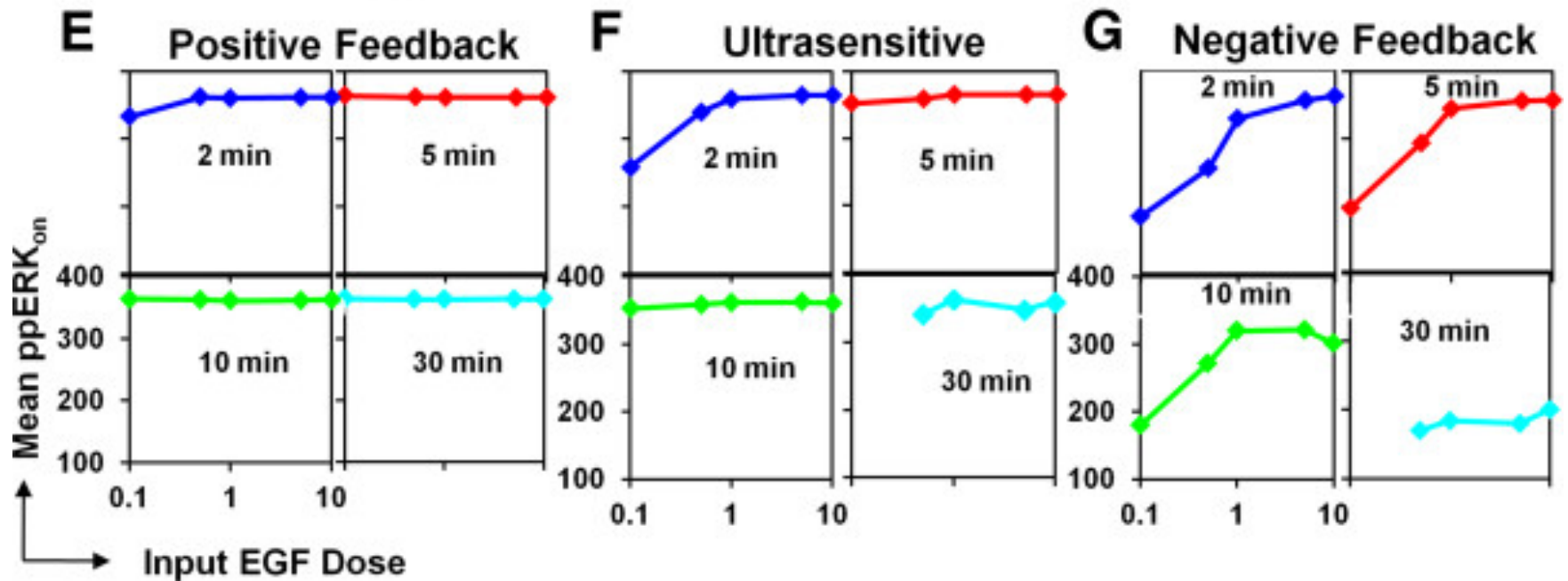
- Unimodal RasGTP distribution assumed
- Peak RasGTP values sampled from a gamma distribution







**All three topologies exhibit time and dose-dependent bimodality or “shouldering”**

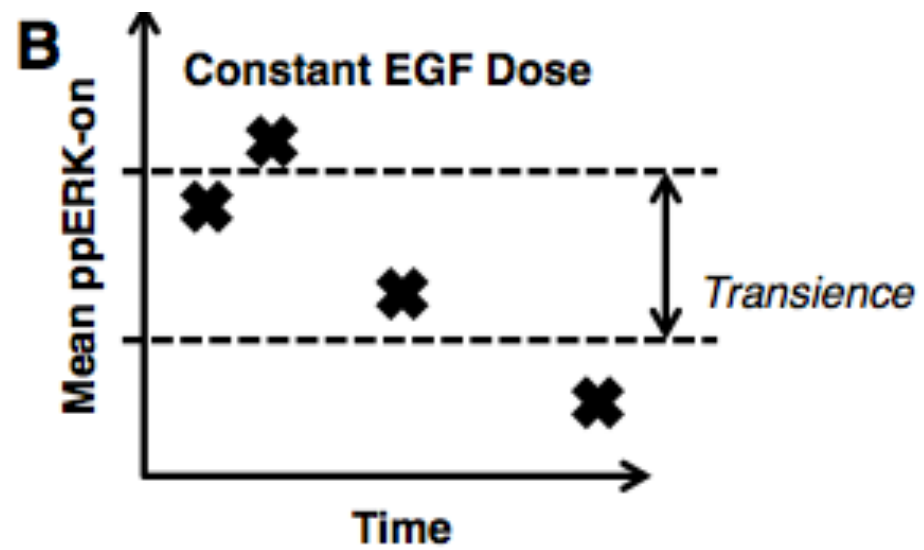
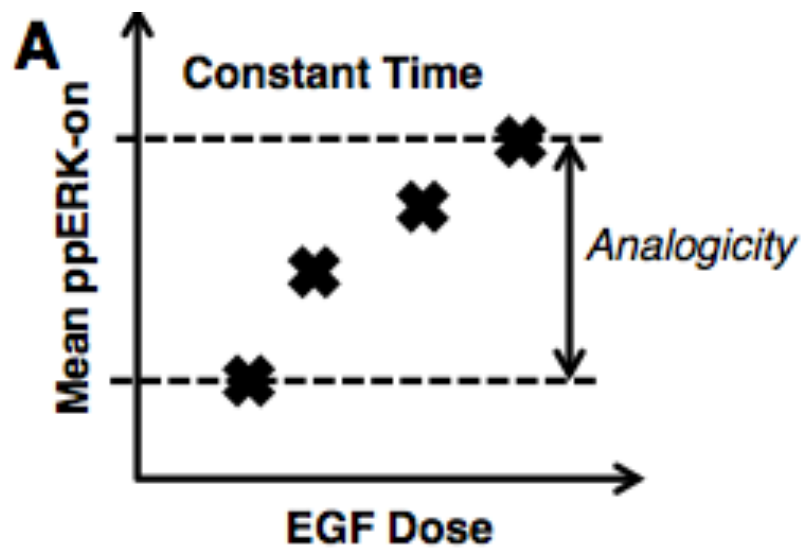


**Only the negative feedback model reproduce proper behavior of the ERK-on population mean:**

- Mean increases as a function of dose at short times
- Decreases as a function of time at a particular EGF dose

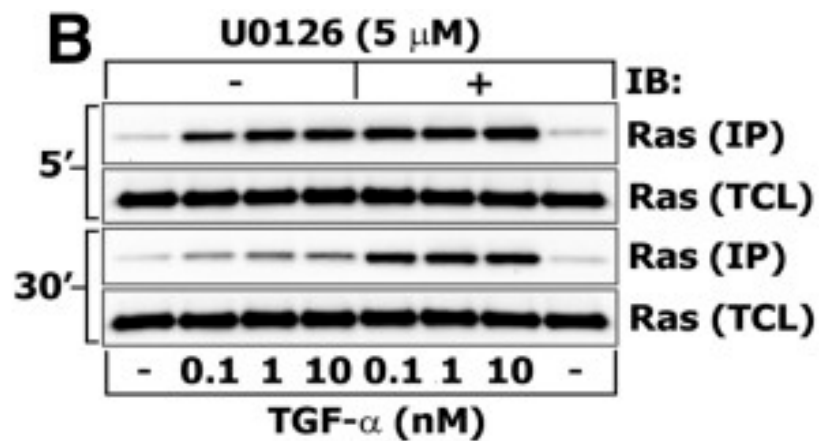
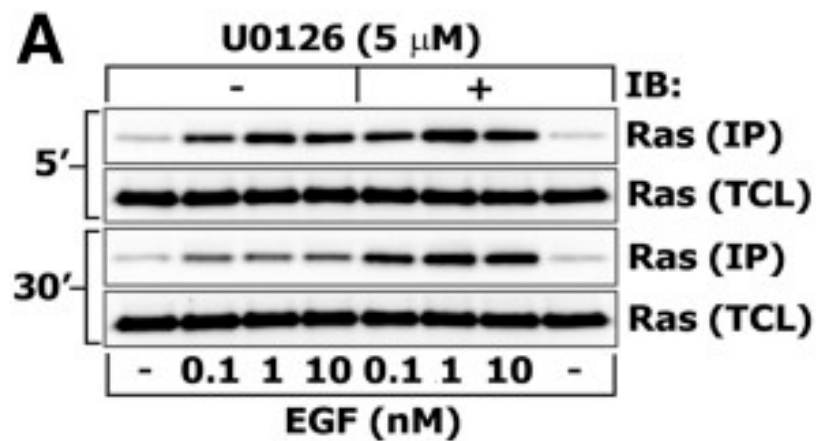
# 500 parameter sets, Latin hypercube sampling: Simulations were analyzed for three features

- **Analogicity:**
  - Erk on population = cells with ppERK > 200nM
  - Calculate the mean ppERK in the ERK-on populations if the population has > 10 cells
  - Analogicity of a time point = max ERK on population mean – min ERK on population mean
  - Total analogicity = analogicity(2min) + analogicity(5min)
- **Transience:** Transience of an individual EGF dose is the mean of the ERK-on population at 2 and 5 minutes minus that at 10 and 30 minutes. Sum over all doses.
- **Bimodality:** Hartigan's Dip Test



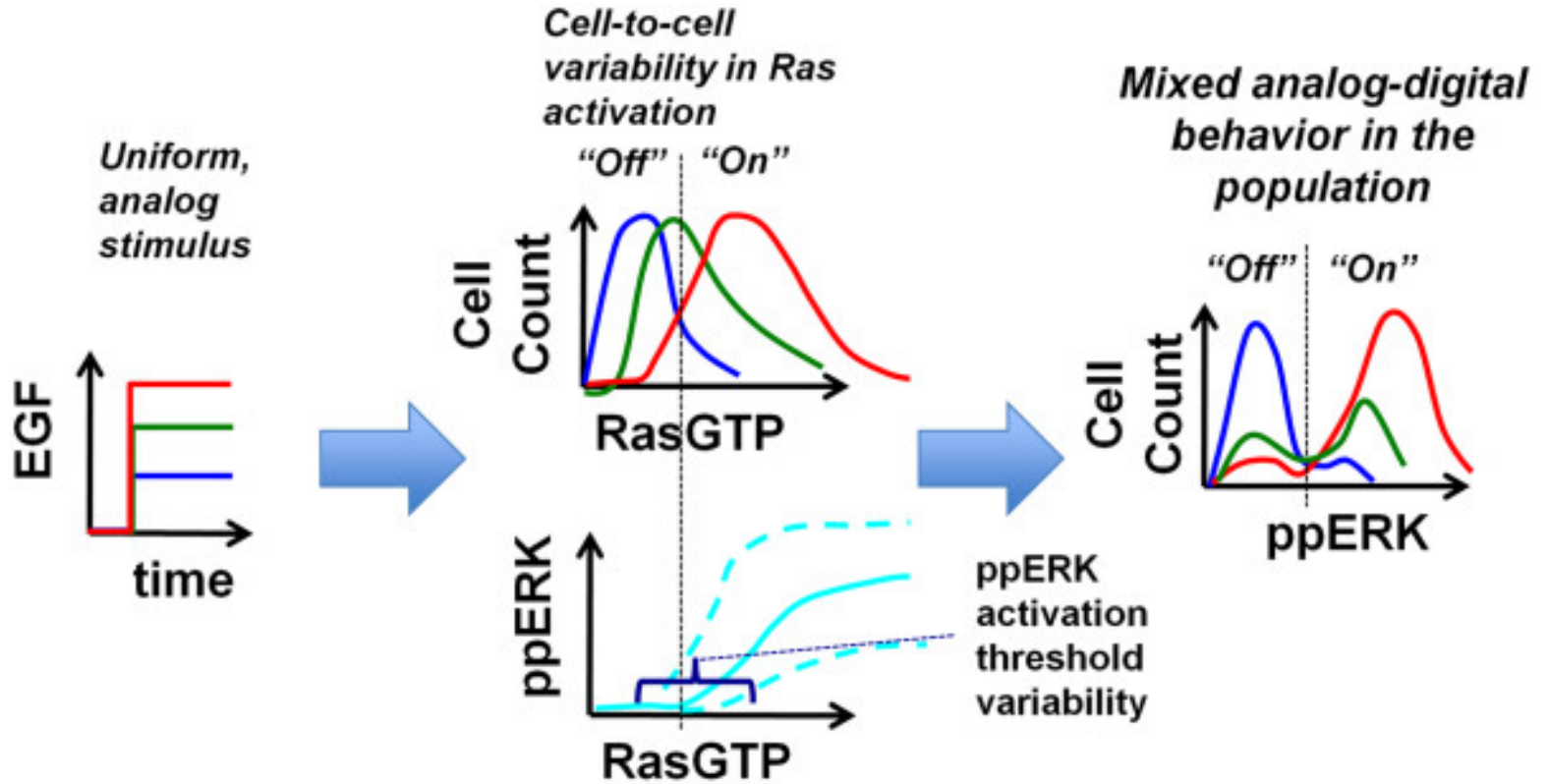
# Test of the negative feedback prediction

- If the feedback were negative, blocking ERK activity should increase the activation of upstream elements such as RasGTP.
- Measure the dynamic and dose response of RasGTP with and without the MEK inhibitor U0126.



# Summary and conclusions

- Flow cytometry to measure EGF-induced single cell ERK activation responses in HEK293 cells
- Bimodal response distributions in cell populations
- An ERK cascade signaling model incorporating negative feedback and a graded, analog single cell dose response is shown to be consistent with observed population responses
- This analog to digital response conversion is suggested to be due to protein abundance variability
- **Thus bimodal distributions can arise from the interplay between protein expression noise and negative feedback-mediated analog single-cell responses.**



*Smooth but variable threshold-linear dose response in single cells controlled by negative feedback and protein expression noise*



Thank you