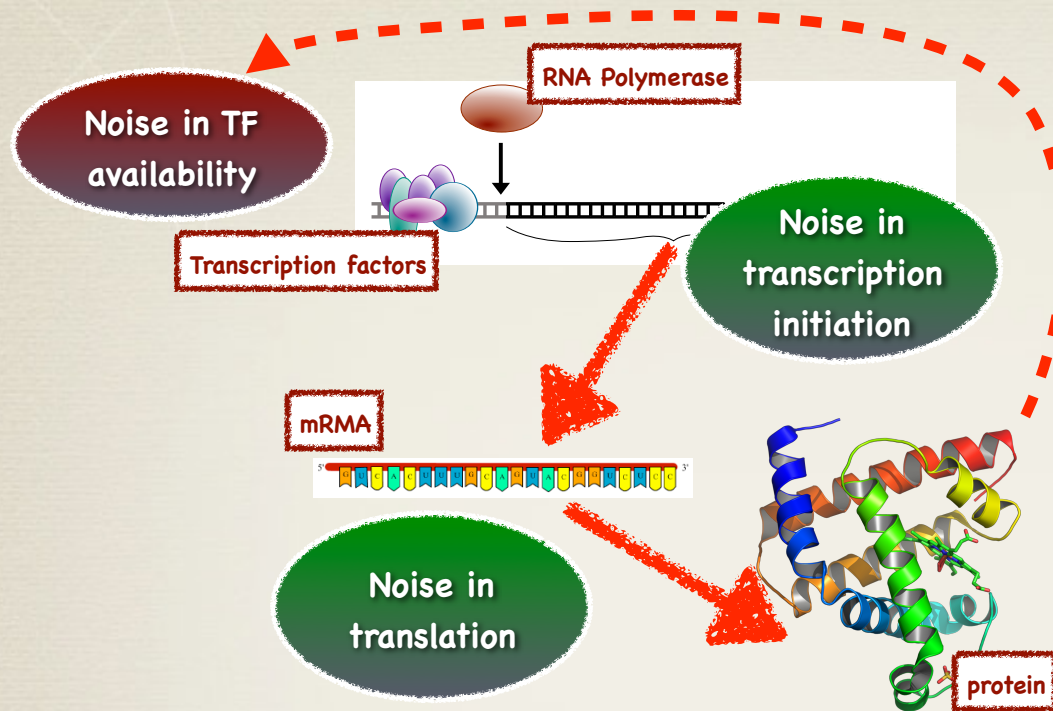


# Transcriptome-wide noise controls lineage choice in mammalian progenitor cells

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Martin Hemberg,  
Mauricio Barahona,  
Donald E. Ingber  
Sui Huang

*Nature* 453, 2008.

# Background on cellular noise



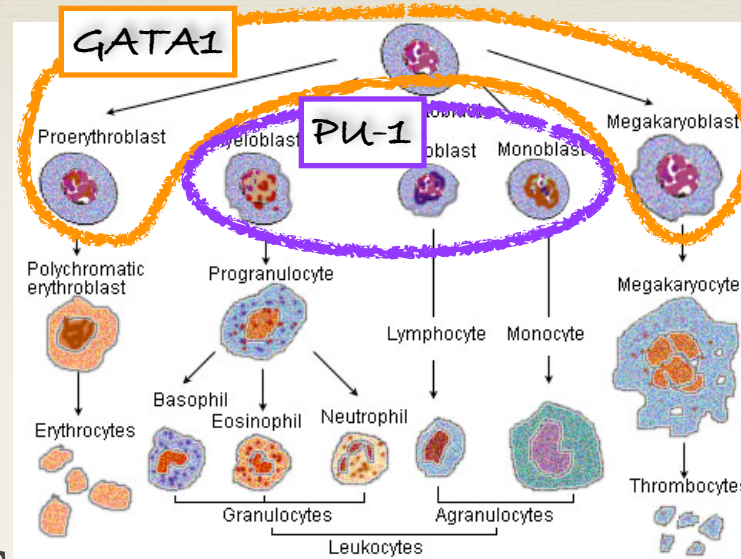
- \* in Eukariotic systems: dominated by **bursty transcription**
- \* phenotypic cell to cell variability
- \* beautifully measured, quantified, modelled in *E. coli* and *Yeast*

- \* **stable** phenotypic variants with the same genome (memory)
  - \* in human cells: some proteins fluctuate between the low and high range on the timescale of 2 cell cycles!
  - \* proteins in the same pathway **fluctuate in correlated fashion**
- \* **adaptive advantage** upon environmental challenge

# Problems in differentiation

## Instructive

- \* external signals **impose** a genetic program
- \* **master regulators** of lineage
- \* tissue specific promoters
  
- \* complicated interplay between many regulators
- \* directing differentiation is very difficult... (<50% efficiency)



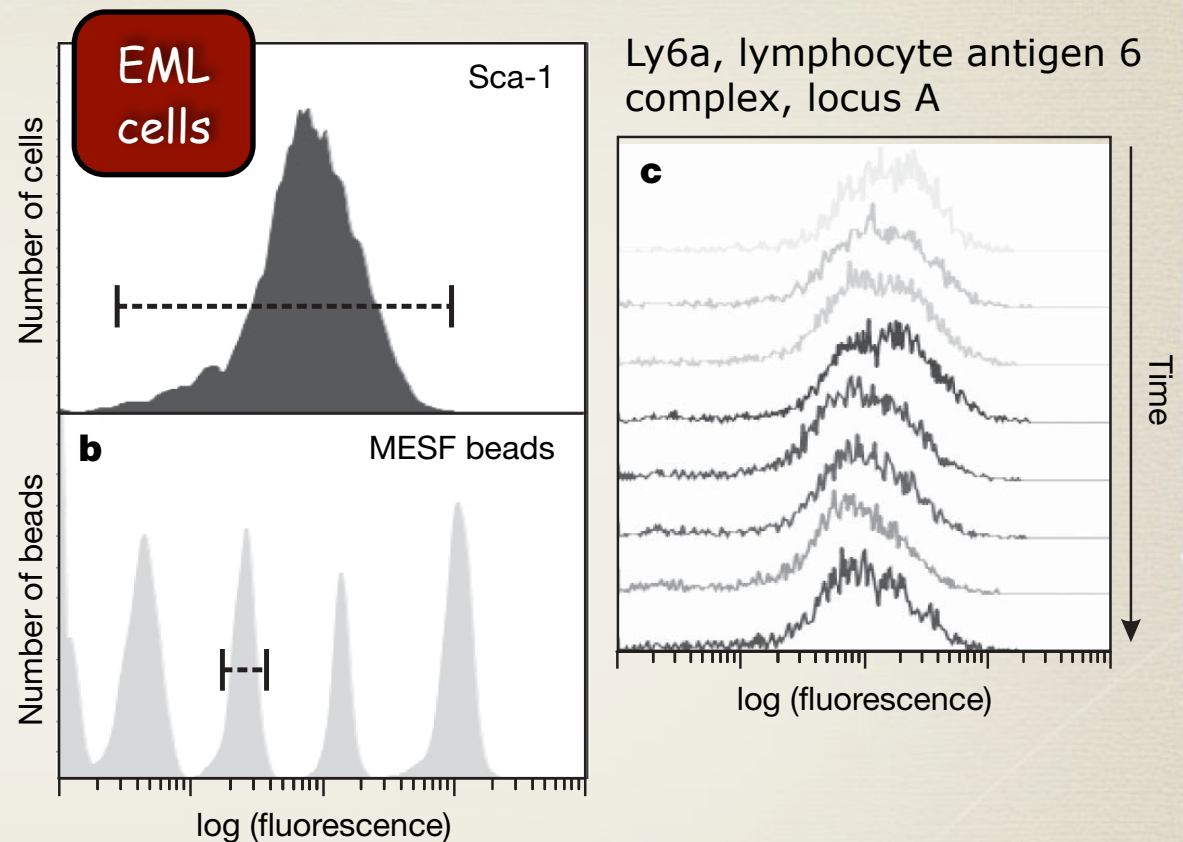
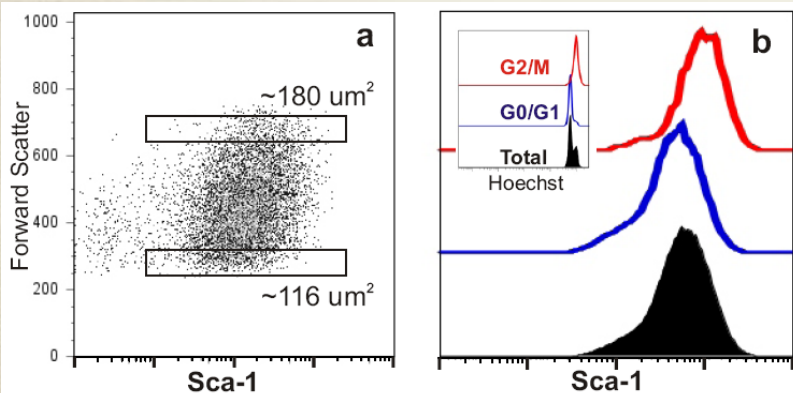
promiscuous expression of many lineage markers

## Selective

- \* spontaneously differentiated **subpopulations** “selected” by external factors
  
- \* mouse embryonic stem cells: remove stem cell medium => spontaneous differentiation to many lineages
- \* single cell plating of hematopoietic stem cells => macrophage, erythrocyte, platelet

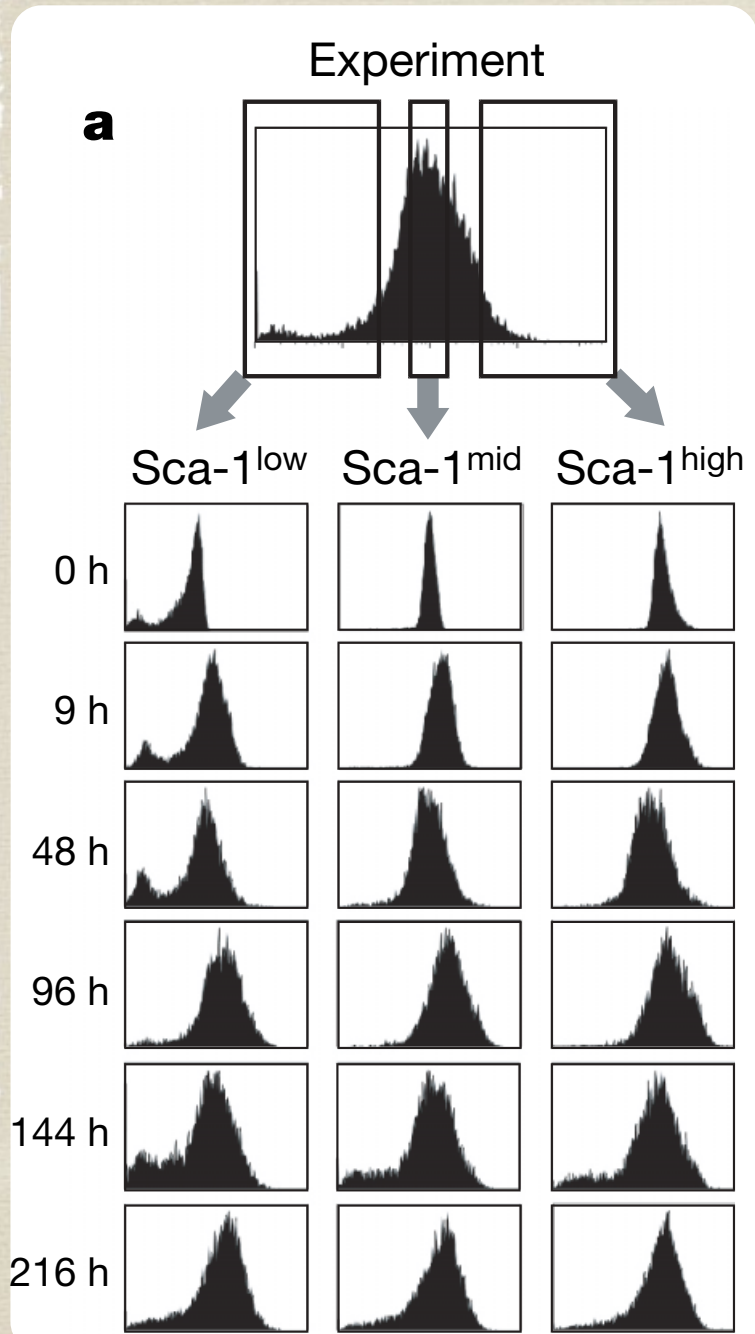
# Does cellular noise play a role in differentiation?

- \* Lets find this noise...
- \* immunofluorescence  
flow cytometry of **cell surface Sca-1 protein**
- \* clonal population
- \* **1000-fold range!**
- \* stable over time
- \* much larger than measurement noise
- \* **NOT** just cell size or cycle

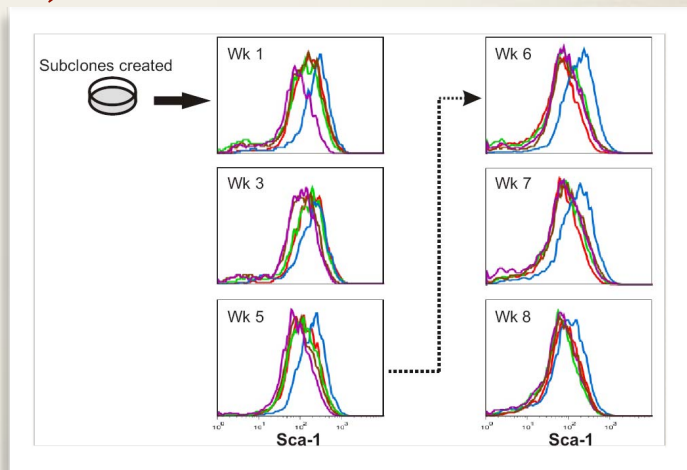


**Figure 1 | Robust clonal heterogeneity.** **a, b**, Heterogeneity among clonal cells in Sca-1 protein expression, detected by immunofluorescence flow cytometry (**a**), was significantly larger than the resolution limit of flow cytometry approximated by measurement of reference fluorescent MESF<sup>24</sup> beads (**b**). The dashed lines show the difference in spread of the distributions as explained in the text. **c**, Stability of clonal heterogeneity in Sca-1 over three weeks.

# How does the heterogeneity arise?



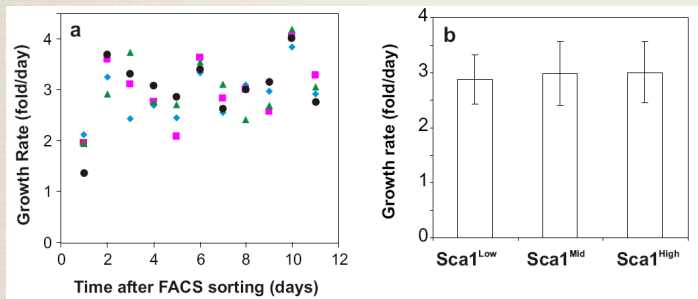
- \* Lets FACS sort the subpopulations!
- \* Low, Med, High Sca-1
- \* culture the fractions
- \* **slow restoration of full heterogeneity (> 12 cell doublings)!**
- \* **also works with single-cell clones (very slowly...):**



**Figure 2 | Restoration of heterogeneity from sorted cell fractions.** **a**, Clonal cells with the highest (Sca-1<sup>high</sup>), middle (Sca-1<sup>mid</sup>) and lowest (Sca-1<sup>low</sup>) 15% Sca-1 expression independently re-established the parental extent of clonal heterogeneity after 216 h in separate culture. As an example, each cell in the Sca-1<sup>high</sup> experiment was theoretically partitioned into one of two GMM subpopulations (blue and red, right).

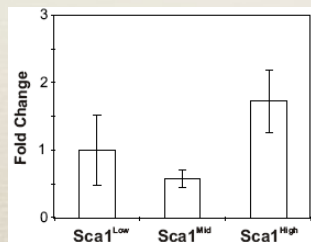
# What drives the heterogeneity?

- \* A few things to exclude:
  - \* NO differential growth of Sca-1 subpopulations



- \* mutations are too slow (9 days, 12 cell divisions)
- \* widening of distribution too fast for uneven partitioning of Sca-1 protein during cell division

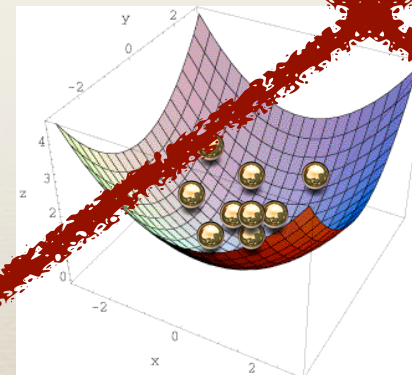
- \* Not much difference in transcription



- \* What governs Sca-1 expression?
  - \* circuitry not known
  - \* explicit modeling unfeasible
- \* Phenomenological approach
  - \* find **class of stochastic processes** that can explain the data

~~I. Mean-reverting process (Ornstein-Uhlenbeck process)~~

- ~~\* noisy relaxation process towards an equilibrium~~



~~Gaussian distribution~~

# What drives the heterogeneity?

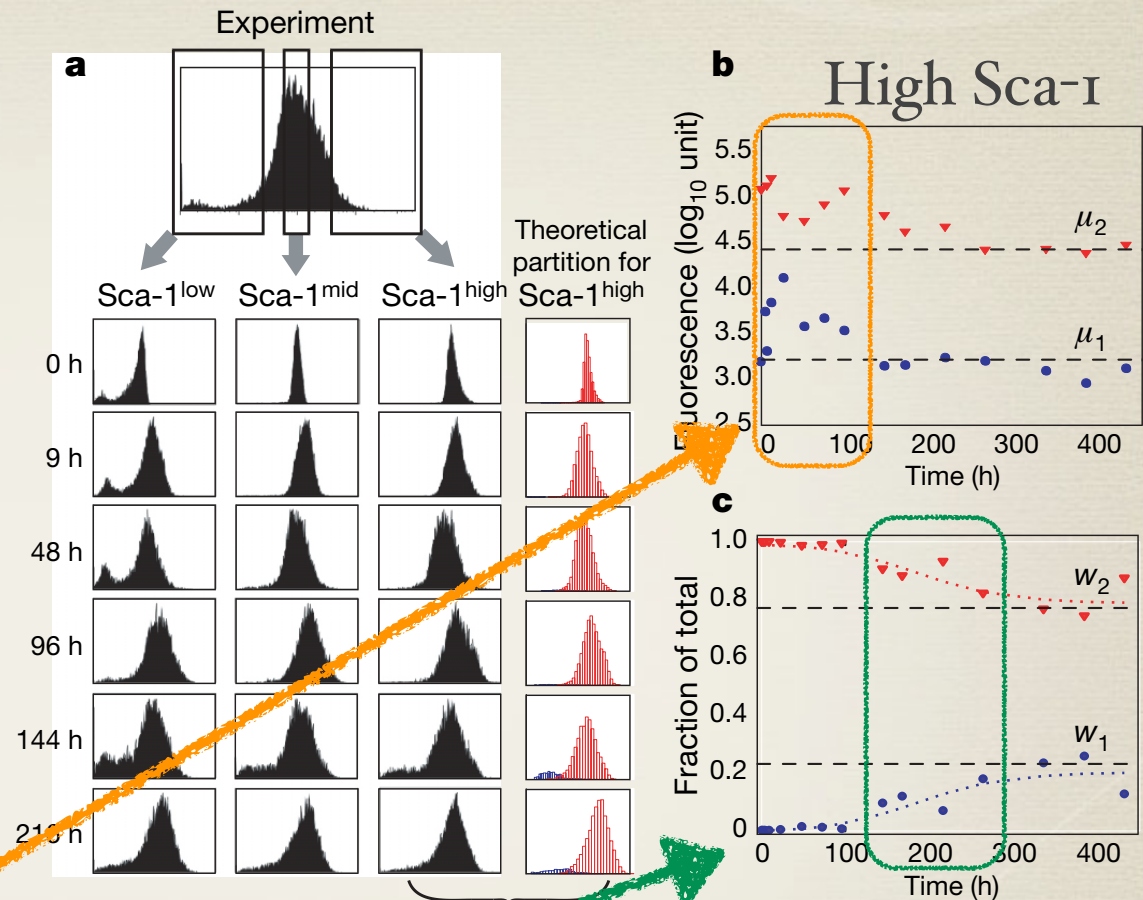
\* How about this one?



## 2. Gaussian mixing model

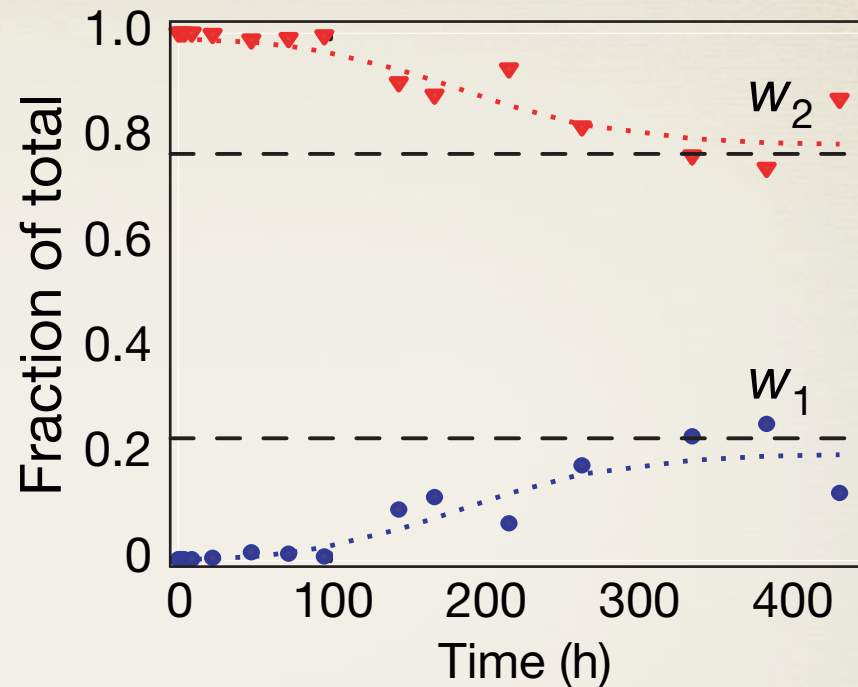
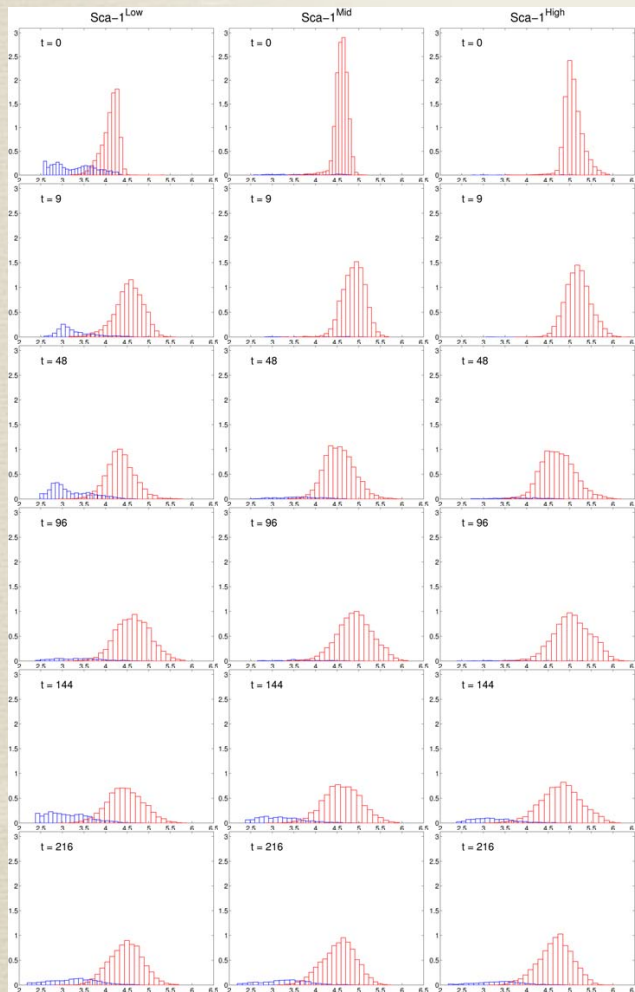
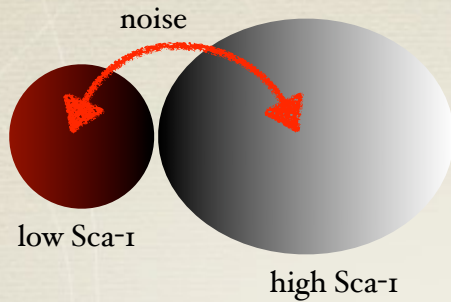
- \* rugged landscape
- \* multiple meta-stable states
- \* relaxation within basins  
(sub-populations!)
- \* stochastic transition  
between states

Overlap of 2 Gaussians!



**Figure 2 | Restoration of heterogeneity from sorted cell fractions.** **a**, Clonal cells with the highest (Sca-1<sup>high</sup>), middle (Sca-1<sup>mid</sup>) and lowest (Sca-1<sup>low</sup>) 15% Sca-1 expression independently re-established the parental extent of clonal heterogeneity after 216 h in separate culture. As an example, each cell in the Sca-1<sup>high</sup> experiment was theoretically partitioned into one of two GMM subpopulations (blue and red, right). **b**, **c**, The temporal evolution of the means  $\mu_{1,2}$  (**b**) and weights  $w_{1,2}$  (**c**) for the Sca-1<sup>high</sup> GMM subpopulations 1 and 2. The evolution of the weights was fitted to a sigmoidal function (**c**, dotted curves). Black dashed lines, equilibrium values for  $\mu_i$  and  $w_i$ .

# Hidden surprise in Supplementary Data!



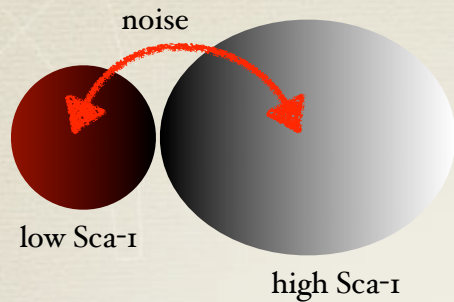
\* does the fitted line mean anything?

Best fit: relative population size AFFECTS transition rates!

Autocrine signaling

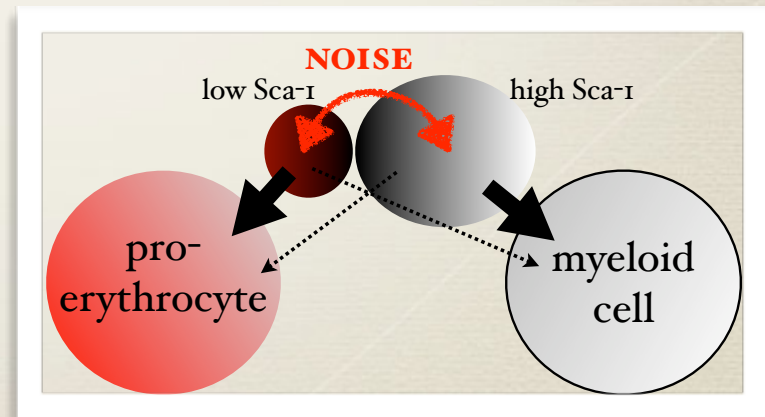
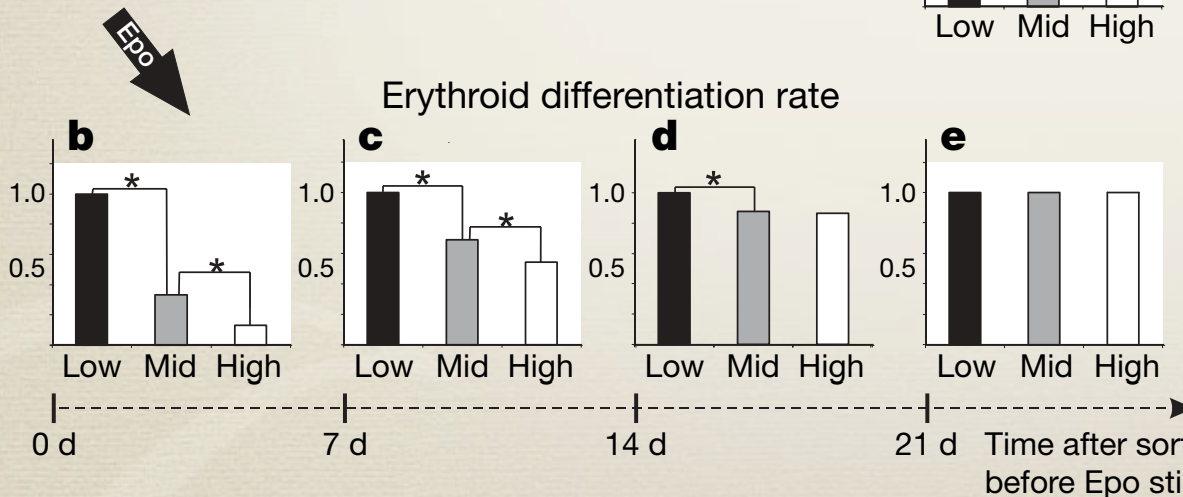
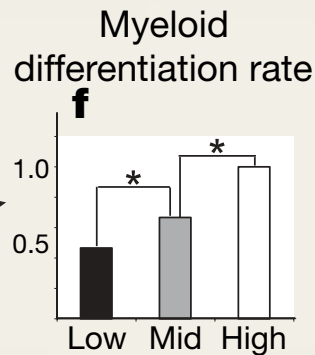
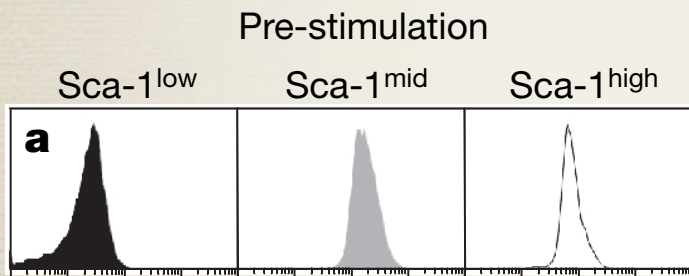
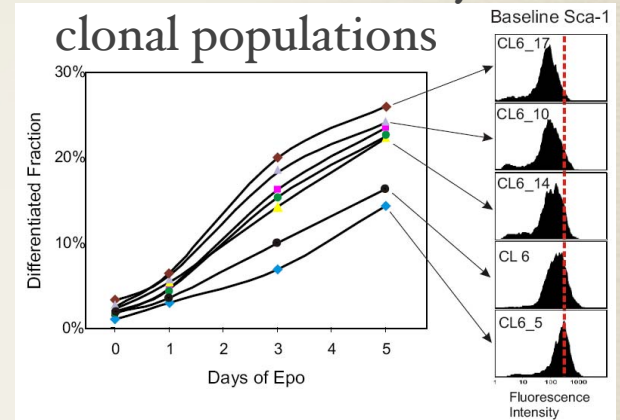


# A cell population with two states!



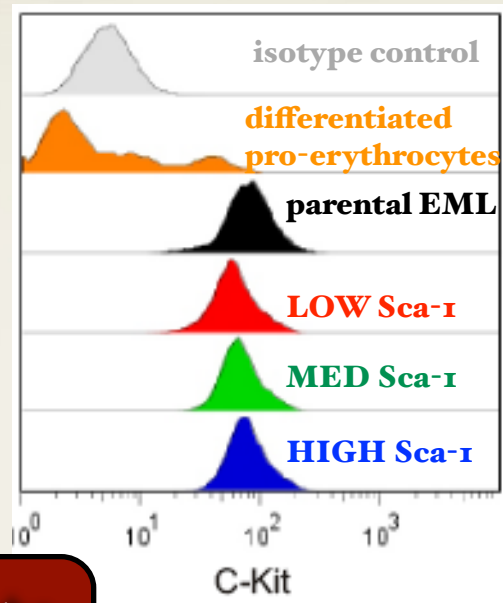
- \* Is it biologically relevant?
- \* how about differentiation potential?

\* Similar in secondary clonal populations



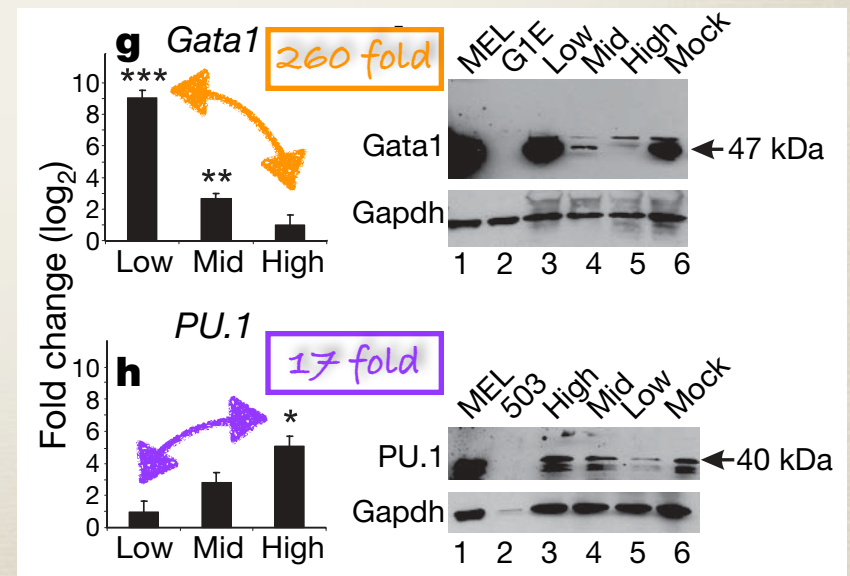
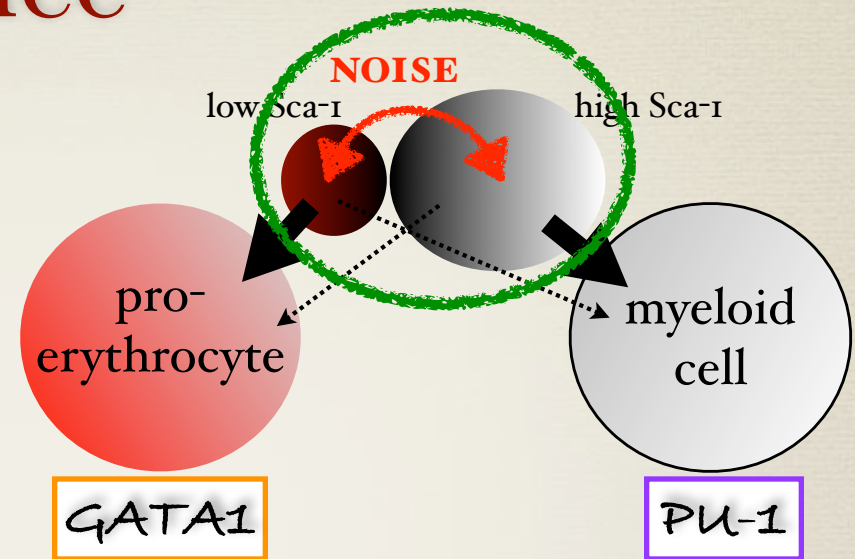
# Persistent but reversible lineage preference

- \* The low Sca-1 subpopulation is not pre-committed!
- \* preference lost in 3 weeks (distributions are no longer distinguishable at day 7!)

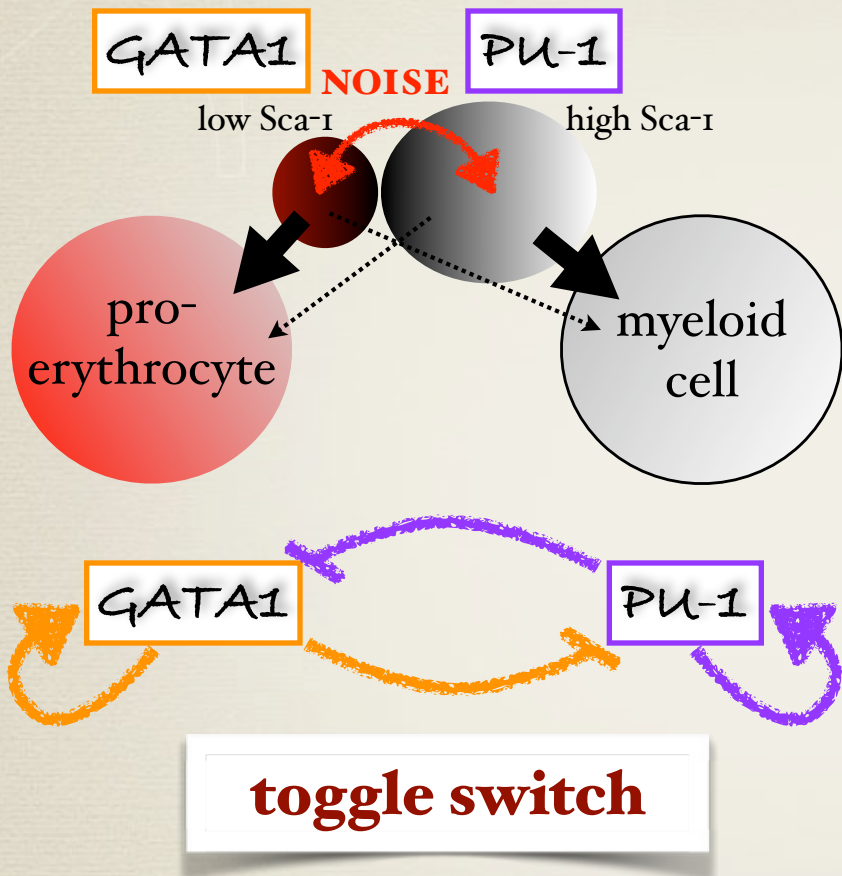


more than Sca-1 to the difference!

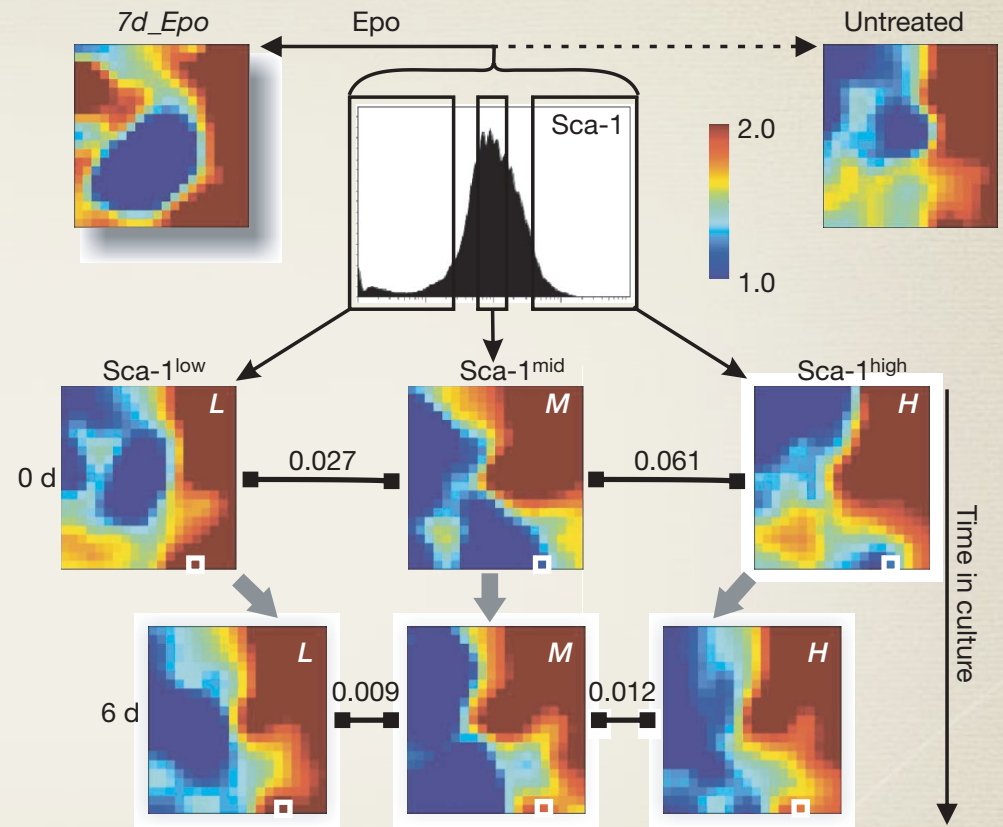
- \* how about **master regulators** of lineage?



# Lineage-specific markers drive a broad expression program!



\* these two master regulators do not DECIDE lineage!!!



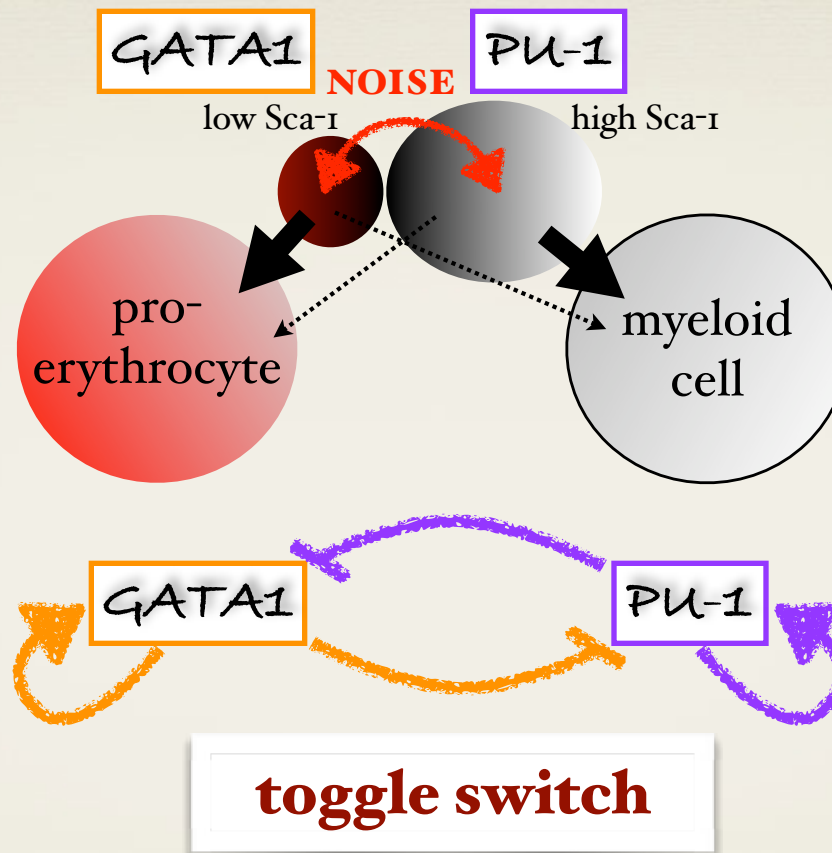
**Figure 4 | Clonal heterogeneity of Sca-1 expression reflects transcriptome-wide noise.** Self-organizing maps of global gene expression for a subset of 2,997 genes in Sca-1<sup>low</sup> (L), Sca-1<sup>mid</sup> (M), Sca-1<sup>high</sup> (H) differentiating cells and control samples. The same master regulator value of gene expression is shown. Dissimilarity distance symbols. The Gata1-containing pixel is boxed in white.

Large, transcriptome-wide differences!  
 >3,900 genes L <-> H

# Instructive AND Selective

## Instructive

- \* terminal differentiation only happens upon stimulation
- \* master regulator expression is part of the commitment process



## Selective

- \* noise can switch cells between functional states with different differentiation potentials
- \* master regulators do not always determine lineage choice

Noise drives preference but not pre-commitment of lineage

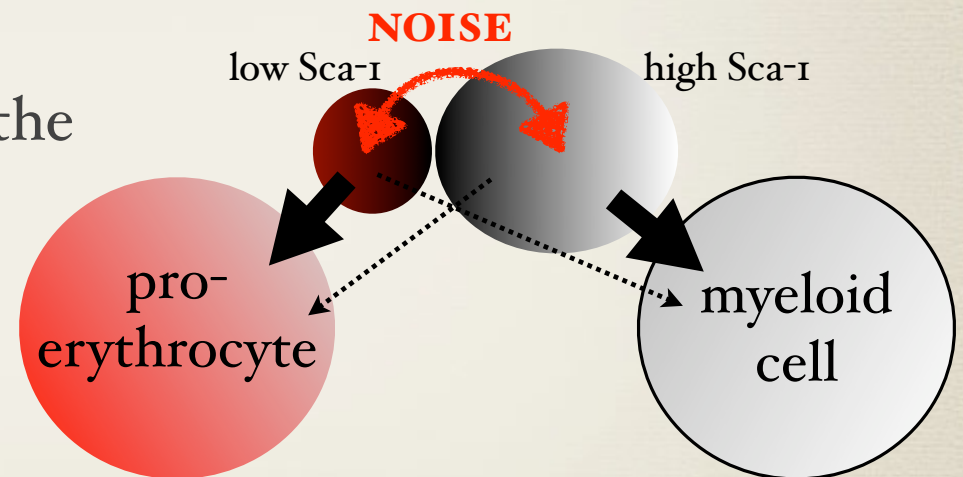
# Outlook

## \* Strengths:

- \* I admit I am biased: it's beautiful (inspiring is a more accepted way of putting it...)
- \* it really asks us to keep in mind the complexity and non-linear nature of the regulatory network
- \* supports the idea of cell states as stable attractors

## \* Weaknesses:

- \* ??? (I have no expertise to judge the experimental techniques)



## \* More coming ...

- \* Siu Huang's lab: working on switching cancer cell lines back to normal
- \* Jim Collins's lab (bioengineer, leader in cellular noise control) with first author Hanna Chang: noise-assisted embryonic stem cell differentiation

Thank you!