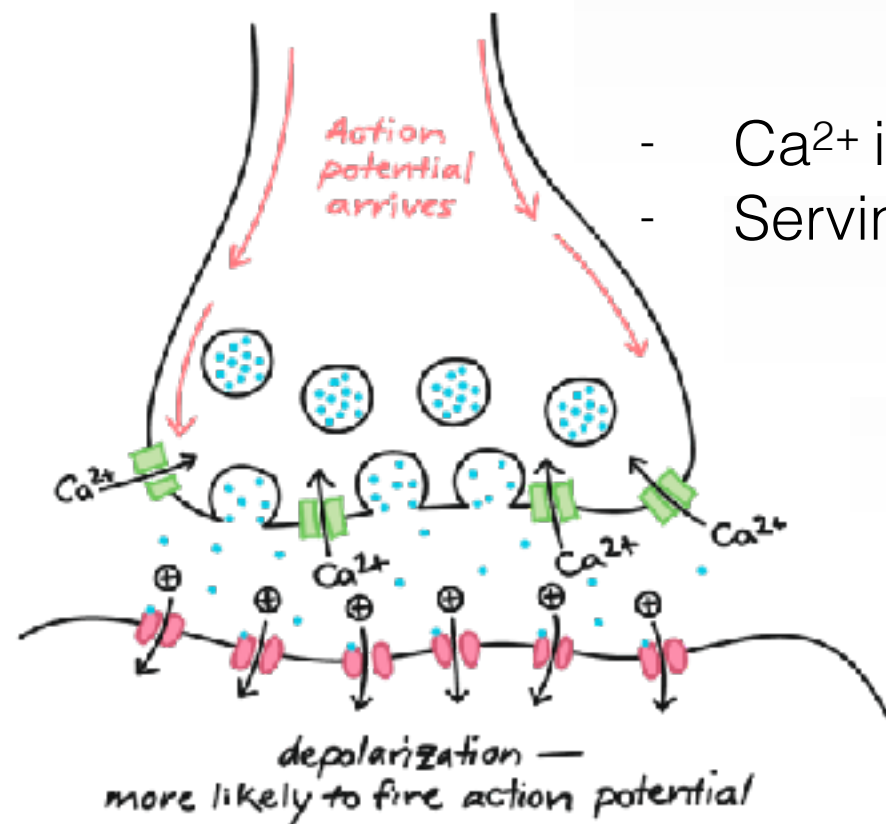
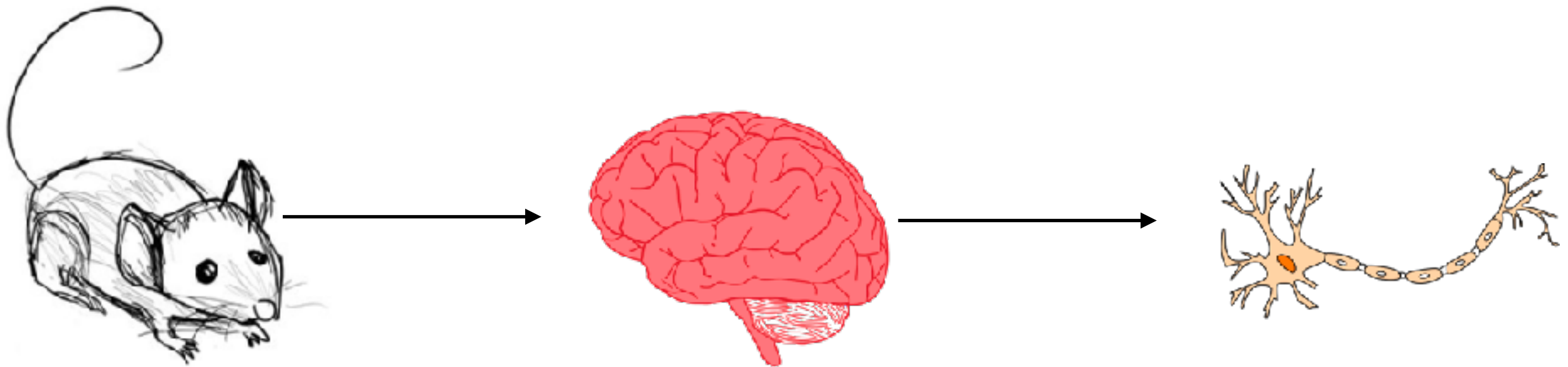


# Exact spike train inference via $\ell_0$ optimization<sup>1</sup>

Author: S. Jewell & D. Witten (AoAS 2018)

Presenter: Yiqun Chen

# Calcium imaging data: what is it and why do we care?



- $\text{Ca}^{2+}$  is essential for **information transmission** in the brain;
- Serving as a good **proxy of activity**

# Calcium imaging data: what is it and why do we care?

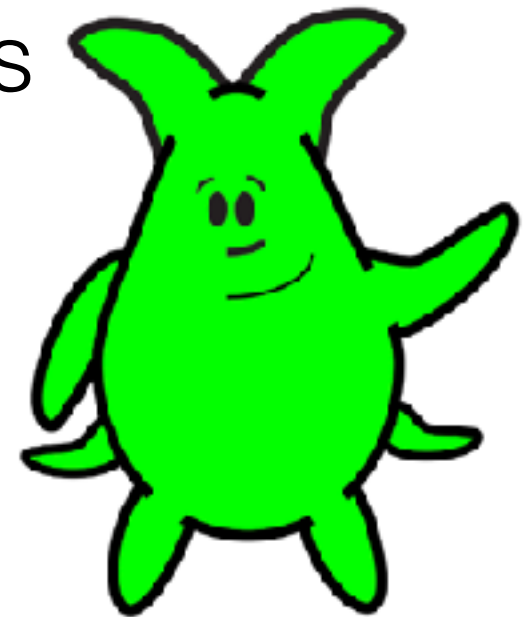
We genetically modify mice genome such that the neurons will be **fluorescent**..

When neuron 1 is active/spikes

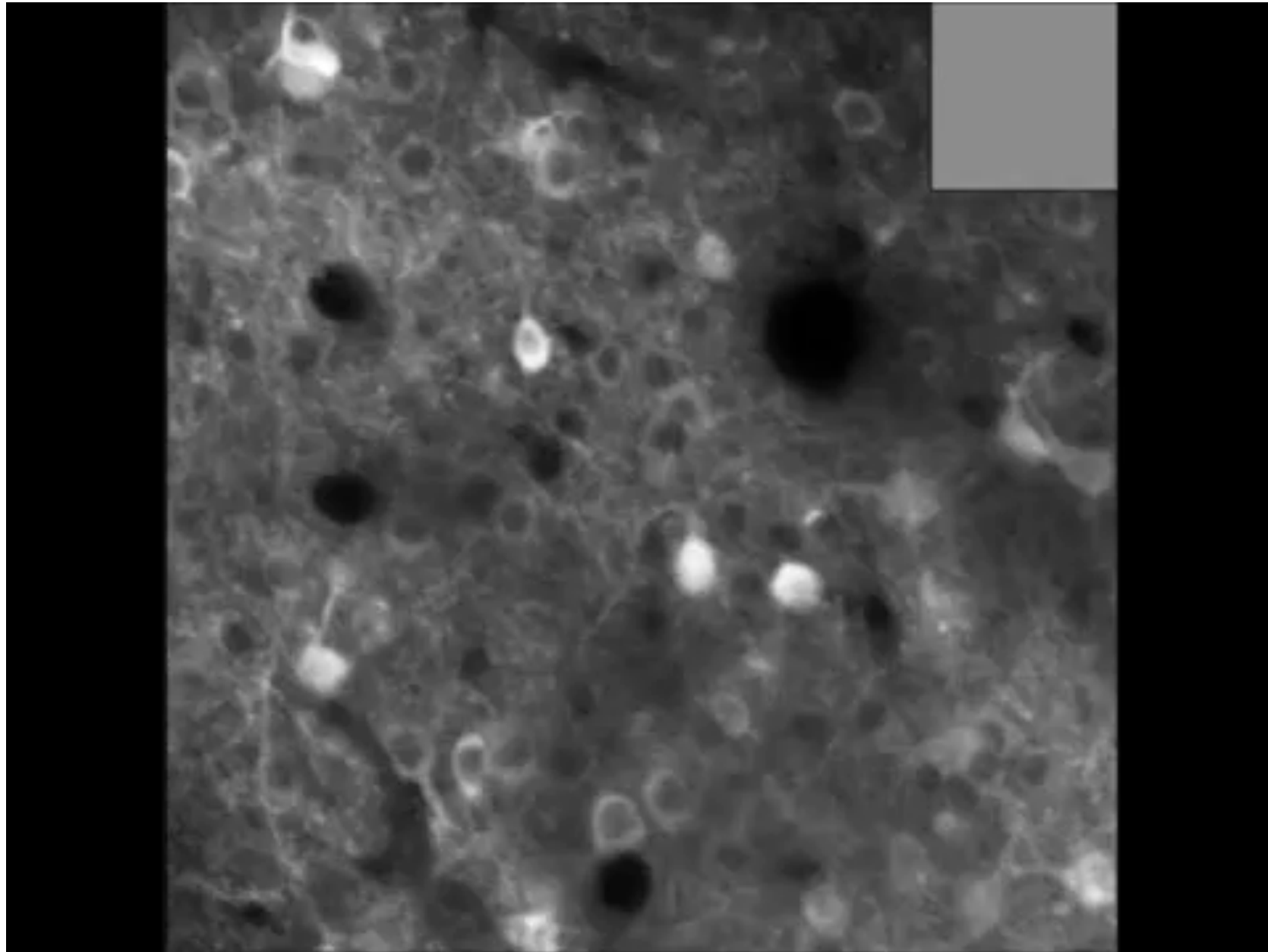


Neuron 1

$\text{Ca}^{2+}$   $\text{Ca}^{2+}$   $\text{Ca}^{2+}$   
 $\text{Ca}^{2+}$   $\text{Ca}^{2+}$   $\text{Ca}^{2+}$   
 $\text{Ca}^{2+}$   $\text{Ca}^{2+}$   $\text{Ca}^{2+}$



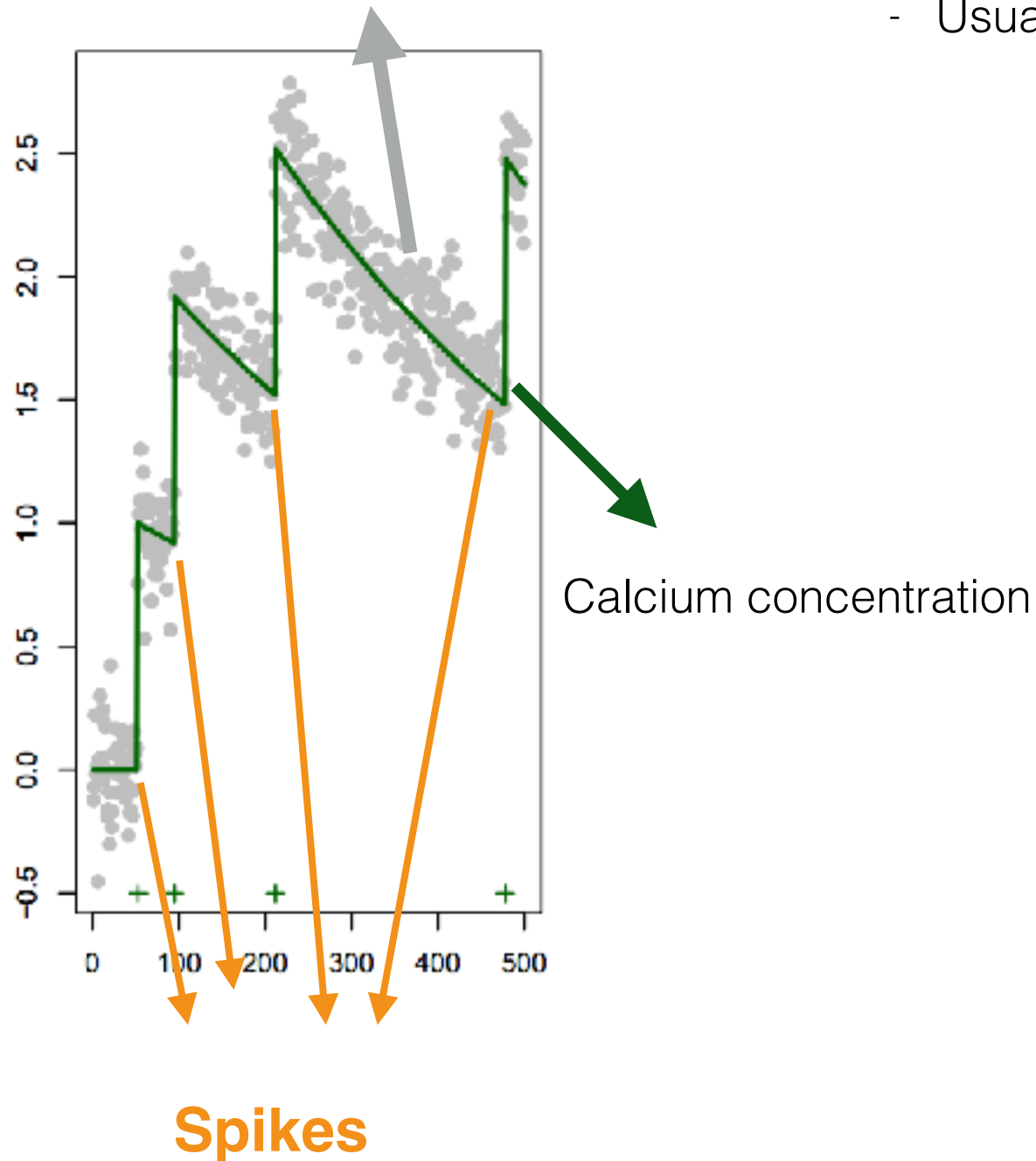
Neuron 1



Calcium imaging from Mouse V1  
- Courtesy of the Sur Lab

# A model for neuron spiking

Discrete-time measurements



- Usually decays **exponentially**

$$C_t = \gamma C_{t-1} + s_t$$

$$F_t = C_t + \epsilon_t$$

Unless a **spike** drives it up

Observed fluorescence is a **noisy** version of the Ca<sup>2+</sup> concentration

# Optimization problem for neuron spiking

$$c_t = \gamma c_{t-1} + s_t$$

$$F_t = c_t + \epsilon_t$$

Non-negative tuning parameter

$\ell_0$  penalty!

$$\text{minimize}_{c_1, \dots, c_T} \sum_{t=1}^T (F_t - c_t)^2 + \lambda \cdot \sum_{t=2}^T 1\{c_t \neq \gamma c_{t-1}\}$$

Fit noisy fluorescence

But we shouldn't fit a ton of spikes

Ideally we wanna solve

$$\text{minimize}_{c_1, \dots, c_T} \sum_{t=1}^T (F_t - c_t)^2 + \lambda \cdot \sum_{t=2}^T 1\{c_t \neq \gamma c_{t-1}\}$$

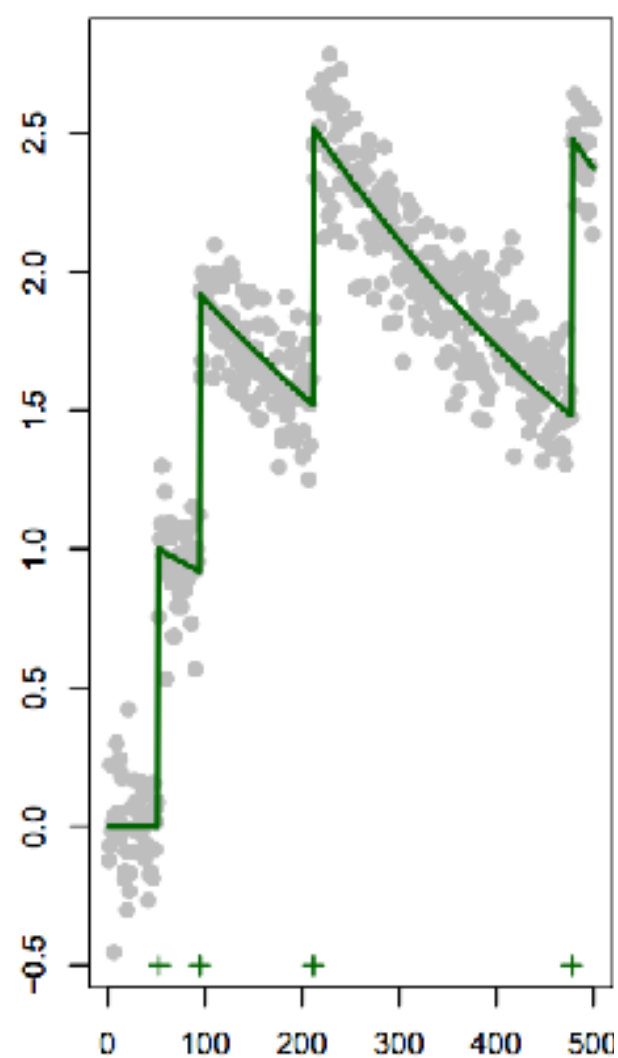
But *naively* this takes  $O(2^T)$  operations

So let's instead try the following...

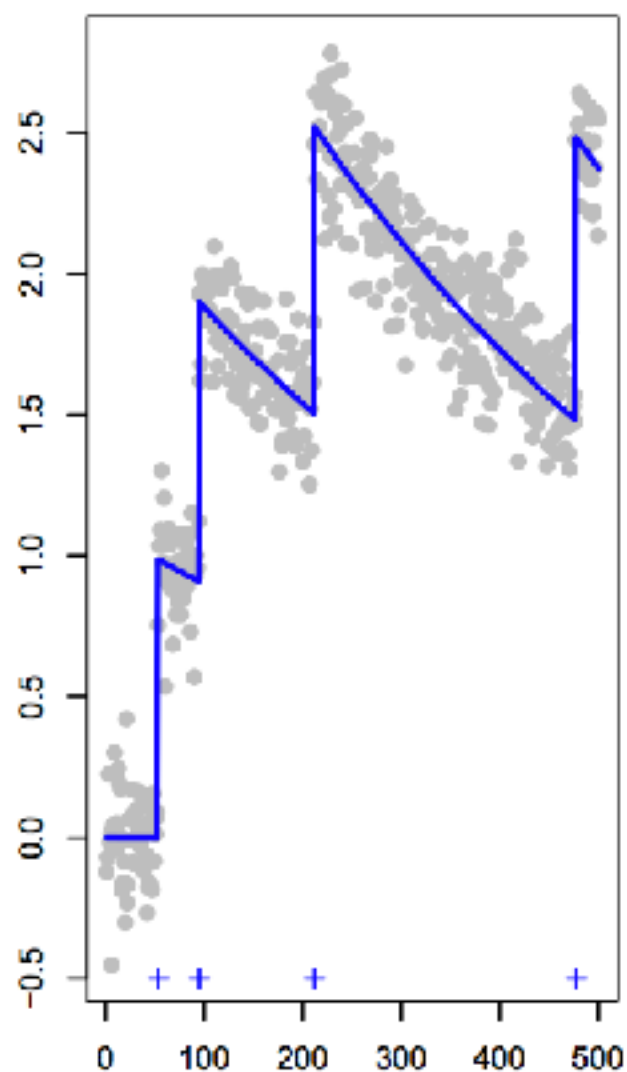
$$\text{minimize}_{c_1, \dots, c_T; s_2, \dots, s_T} \left\{ \frac{1}{2} \sum_{t=1}^T (y_t - c_t)^2 + \lambda \sum_{t=2}^T |c_t - \gamma c_{t-1}| \right\}$$

This now is computationally **tractable!**

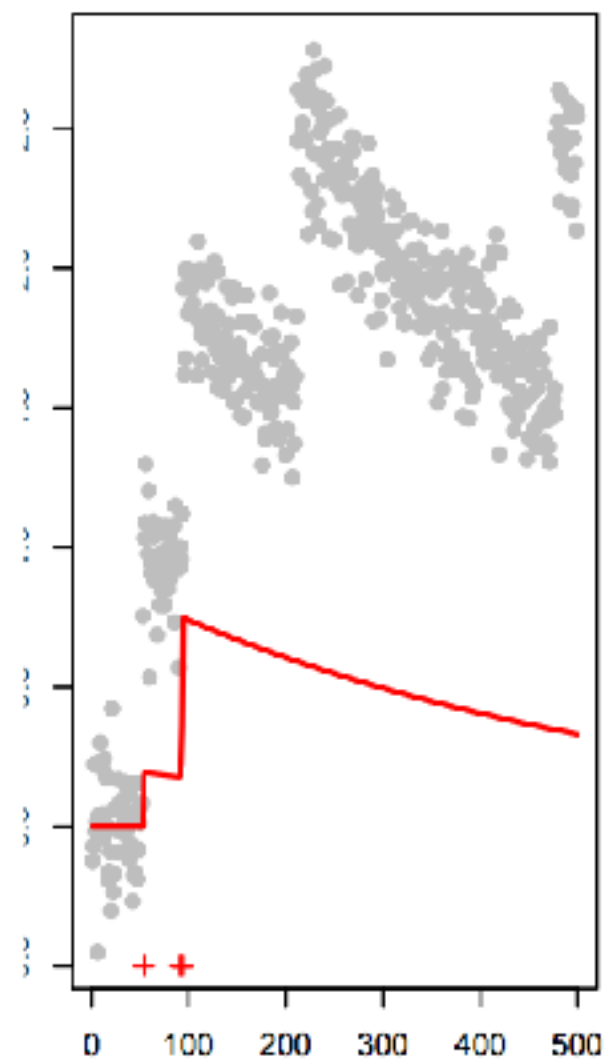
But is it good?



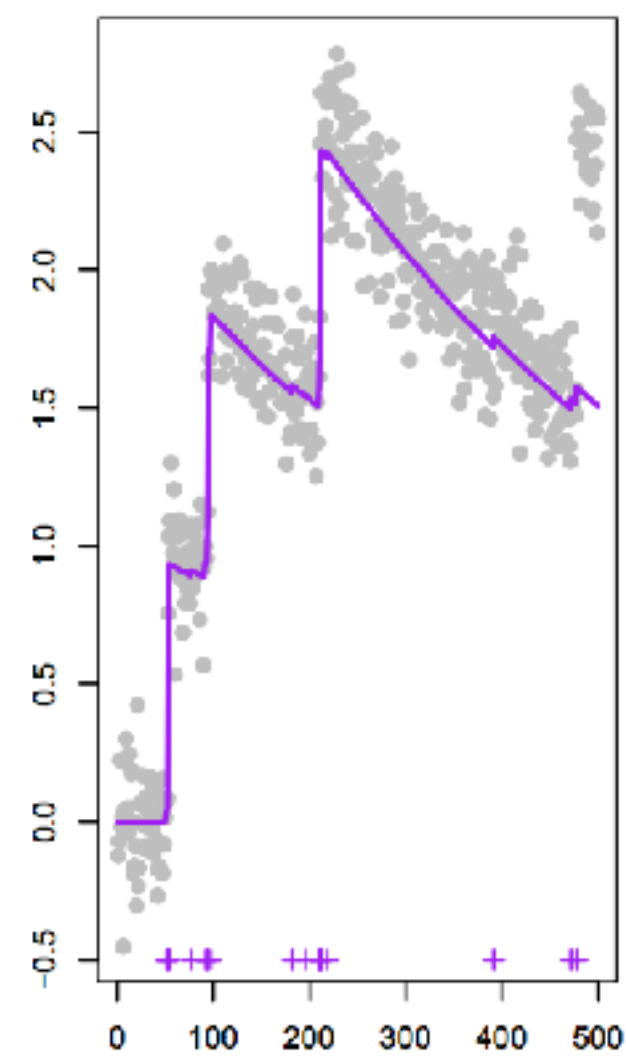
True data  
generating mechanism



$\ell_0$  with small  $\lambda$



$\ell_1$  with large  $\lambda$

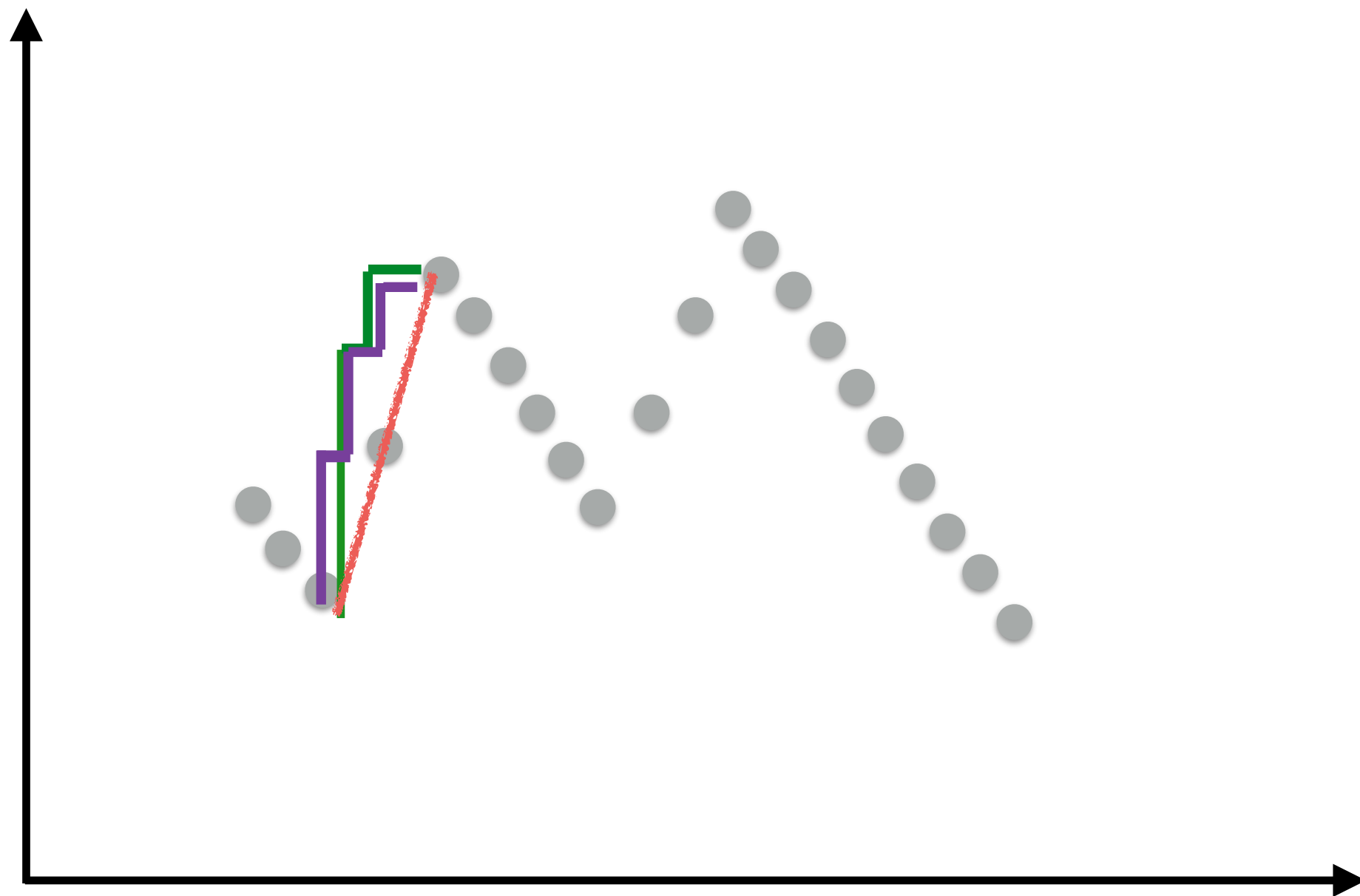


$\ell_1$  with small  $\lambda$

:( Fast computation seems to come with a cost!!



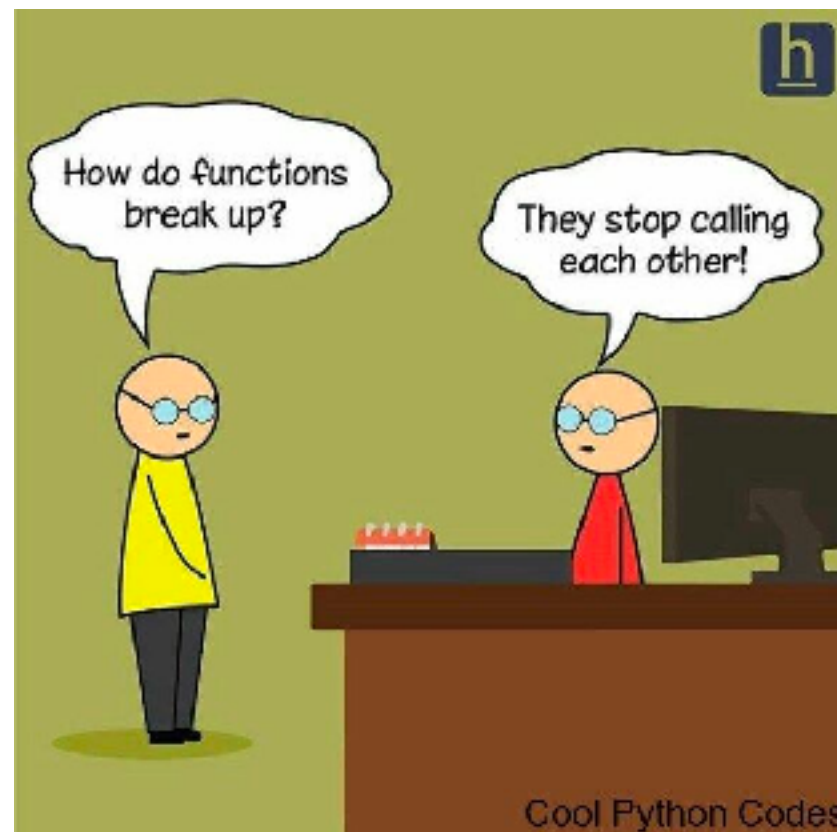
$\ell_0$  versus  $\ell_1$



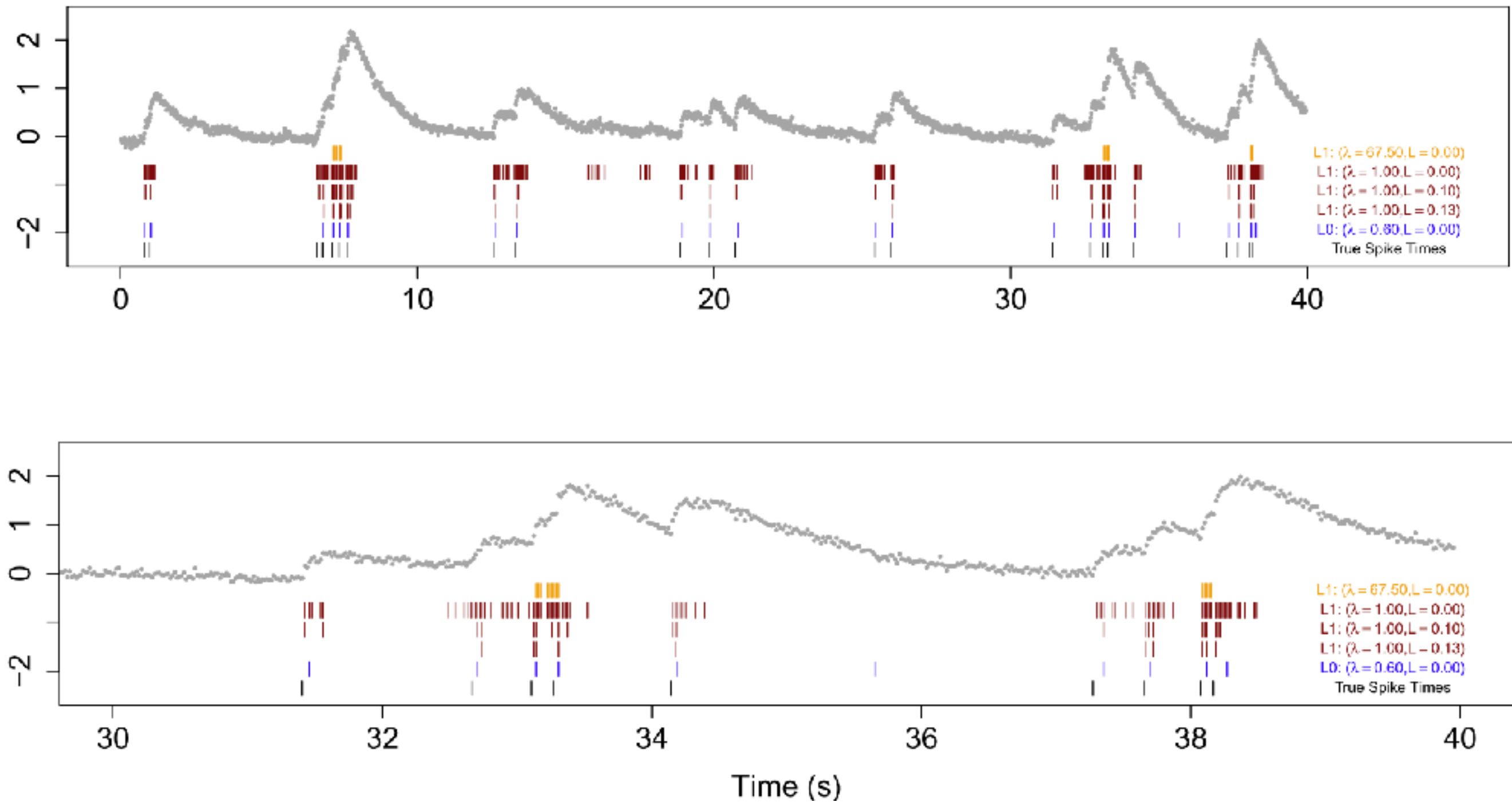
# Contribution of this paper

$$\text{minimize}_{c_1, \dots, c_T} \sum_{t=1}^T (F_t - c_t)^2 + \lambda \cdot \sum_{t=2}^T 1\{c_t \neq \gamma c_{t-1}\}$$

Can be re-casted and solved **efficiently** in  **$O(T^2)$** !  
using  
**dynamic programming** (aka smart recursion)



# Application of the proposed algorithm to data in Chen et al. (2013)



# Reference:

Chen, T.-W., et al. (2013). Ultra-sensitive fluorescent proteins for imaging neuronal activity. *Nature*

Friedrich, J., Zhou, P., & Paninski, L. (2017). Fast online deconvolution of calcium imaging data. *PLoS Computational Biology*, 13(3), e1005423.

Jewell, S., & Witten, D. (2018). EXACT SPIKE TRAIN INFERENCE VIA  $\ell_0$  OPTIMIZATION. *The Annals of Applied Statistics*, 12(4), 2457–2482.

Vogelstein, J. T., et al. (2010). Fast nonnegative deconvolution for spike train inference from population calcium imaging. *Journal of Neurophysiology*, 104(6), 3691–3704.

Thank you for your attention!  
Any questions?

