

NEURAL DATA ANALYSIS

**ALEXANDER ECKER, PHILIPP BERENS,
MATTHIAS BETHGE**

**COMPUTATIONAL VISION AND
NEUROSCIENCE GROUP**

STIMULUS

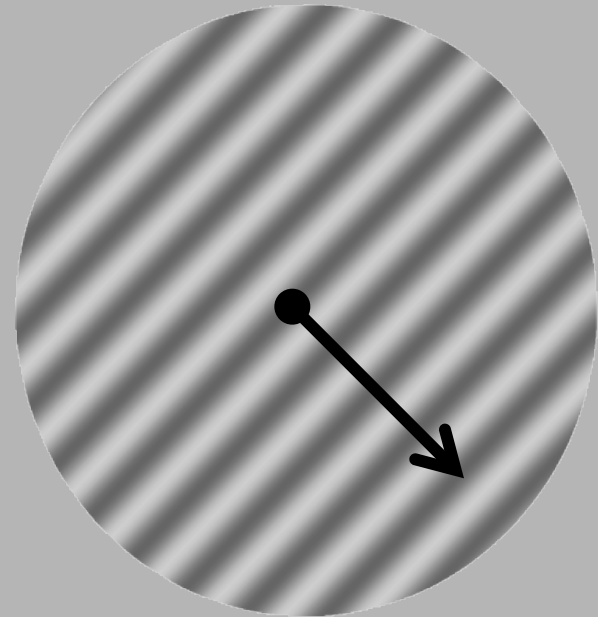
Drifting gratings presented in trials

Parameters:

- 2 sec per trial
- 16 directions of motion
- Diameter: 2 deg
- Eccentricity: ~2-3 deg
- Speed: 3.4 cycles / sec
- Spatial frequency: 3 cycles / deg

Recordings:

- Anesthetized monkey V1



STRATEGIES FOR ANALYZING DATA

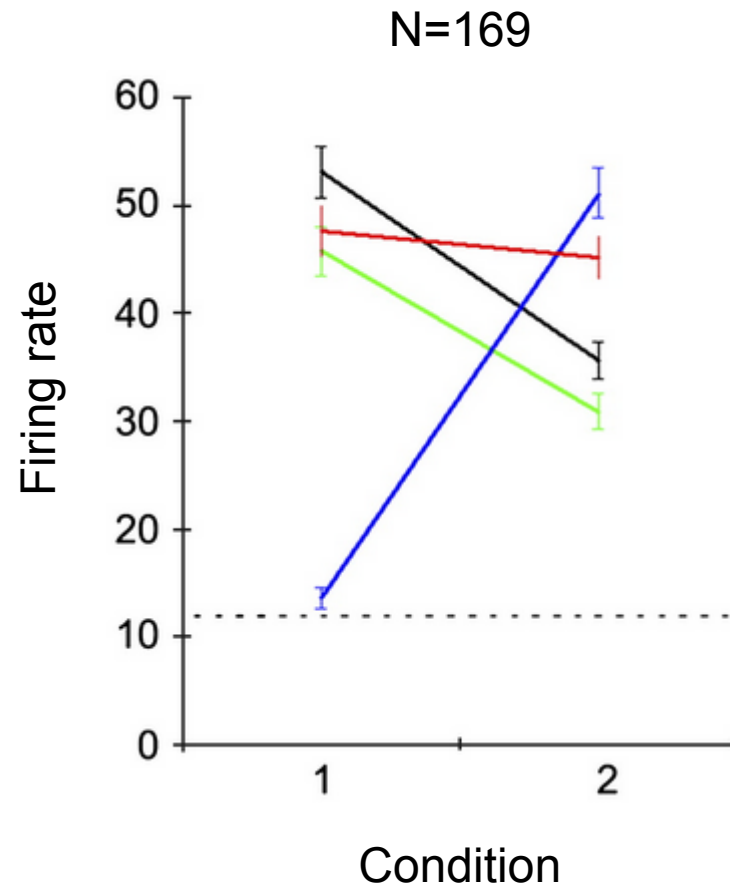
Look at raw data

Visualize spikes

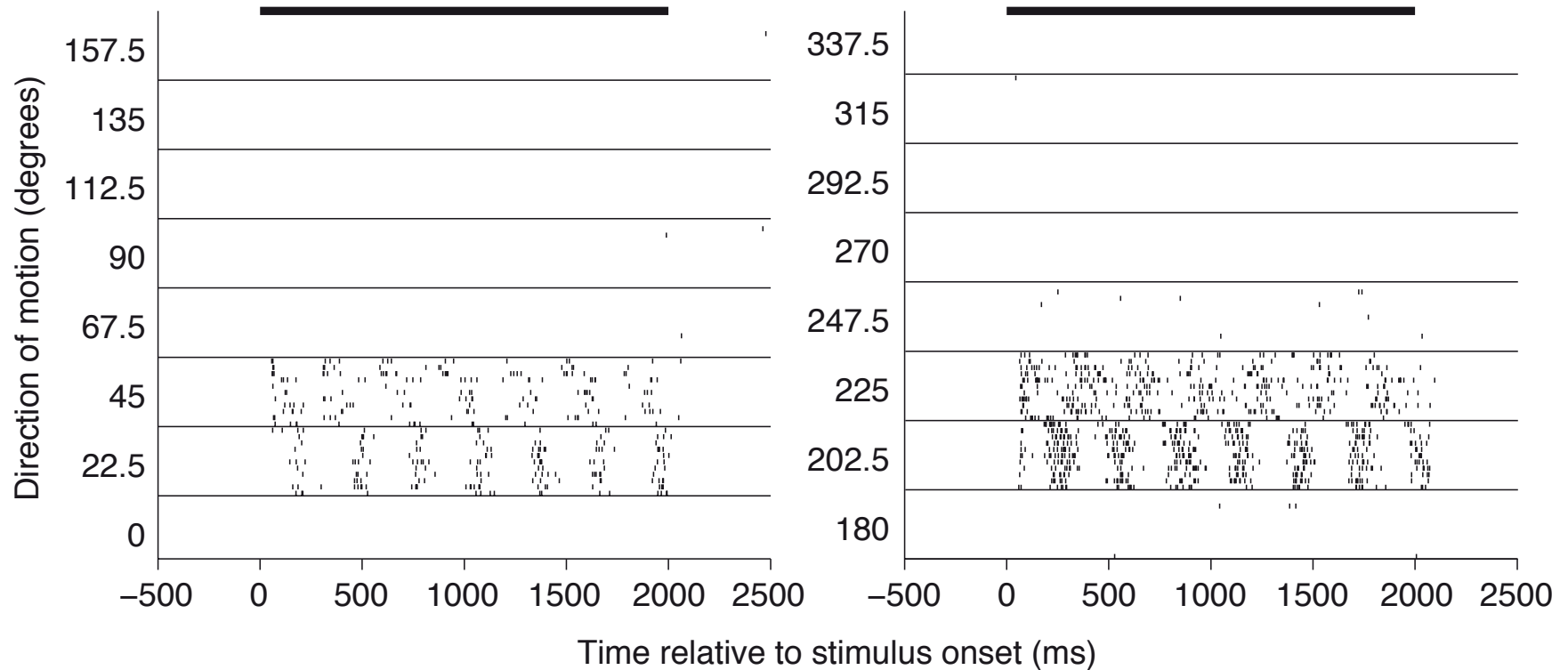
Do not look at

- Fitted tuning functions
- Adaptation indices
- Population averages
- Population histograms

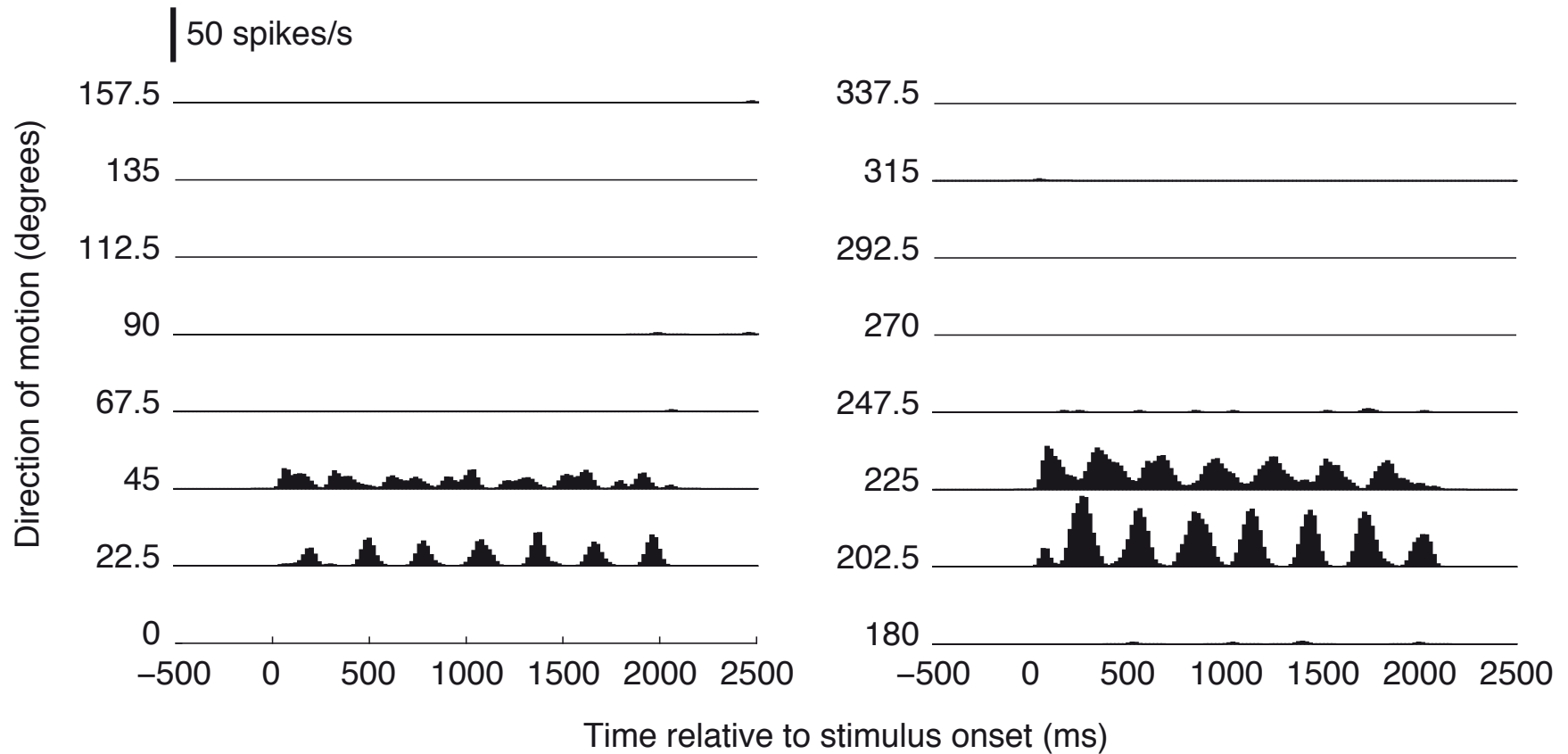
until you spent considerable time looking at raw data and examples



SPIKE RASTERS

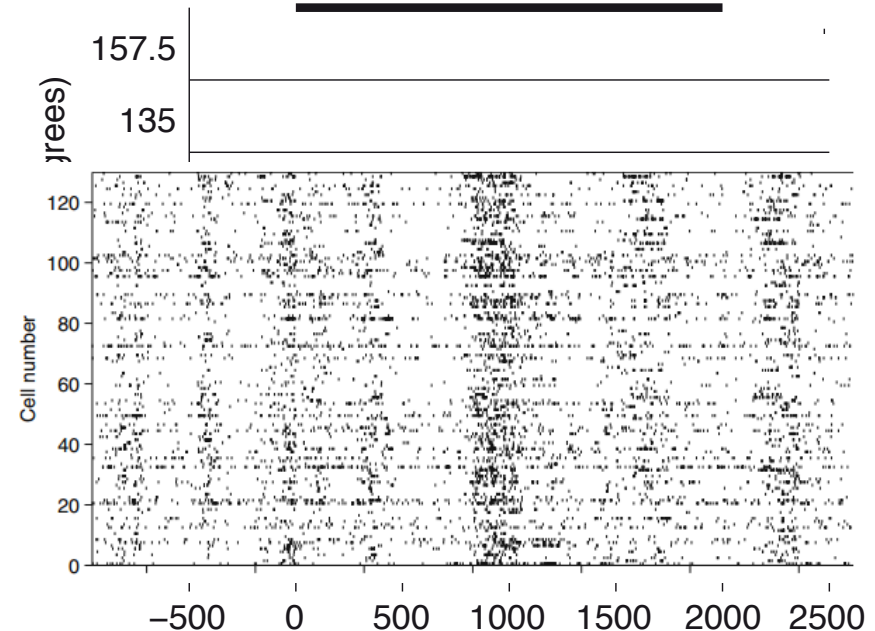


PERI-STIMULUS TIME HISTOGRAM



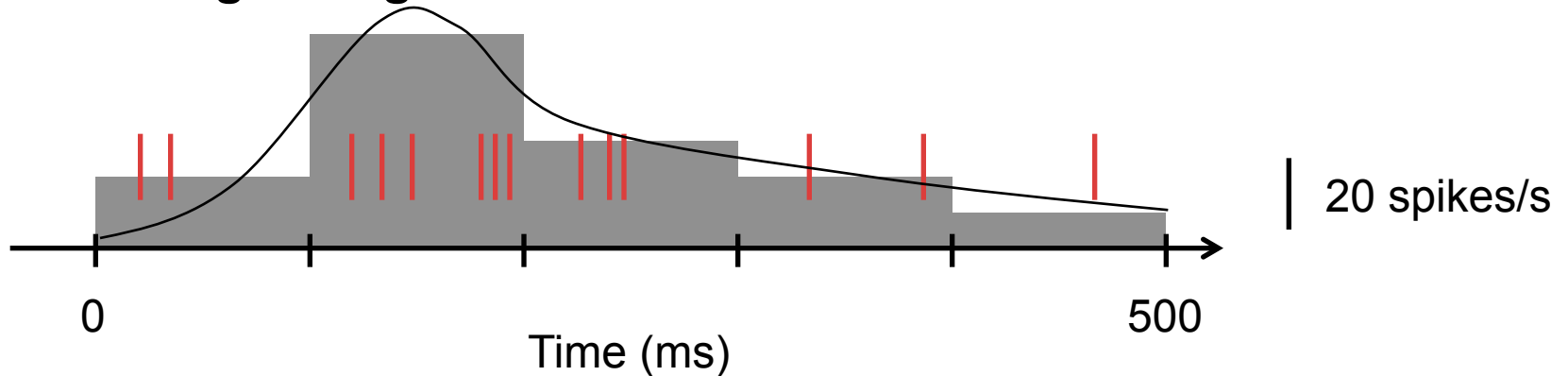
PSTH & SPIKE RASTERS

- Repeated measurements of spike trains with a reference time to align on
 - Stimulus
 - Onset of Movement/Saccade
- Event-related single neuron rate dynamics
- Population raster plot
 - population dynamics
- Align on LFP phase?

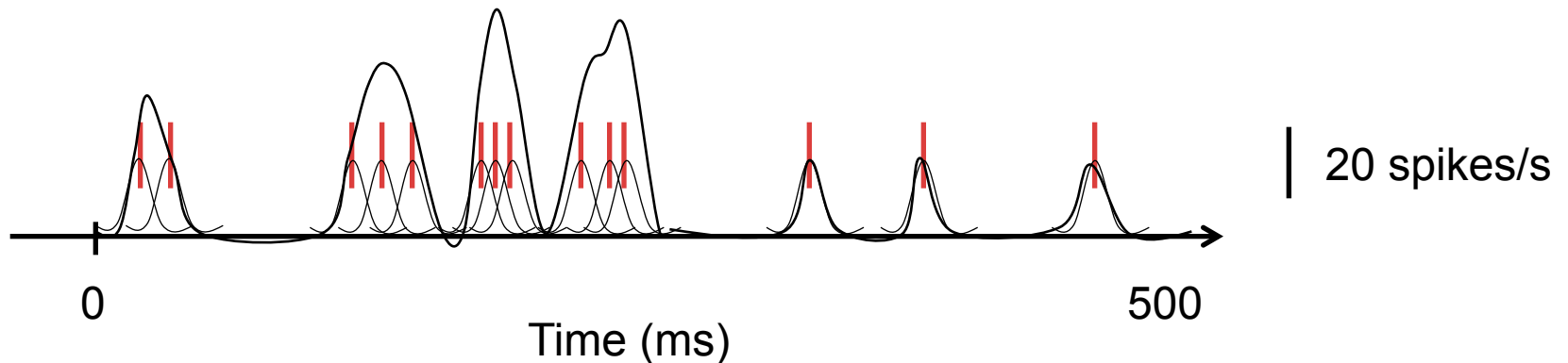


PSTH/SDF – HOW TO

1. Average firing rate in a bin



2. Directly estimate spike density



WINDOW SIZE

Optimal bin width / Gaussian window size
(Shimazaki & Shinimoto 2007)

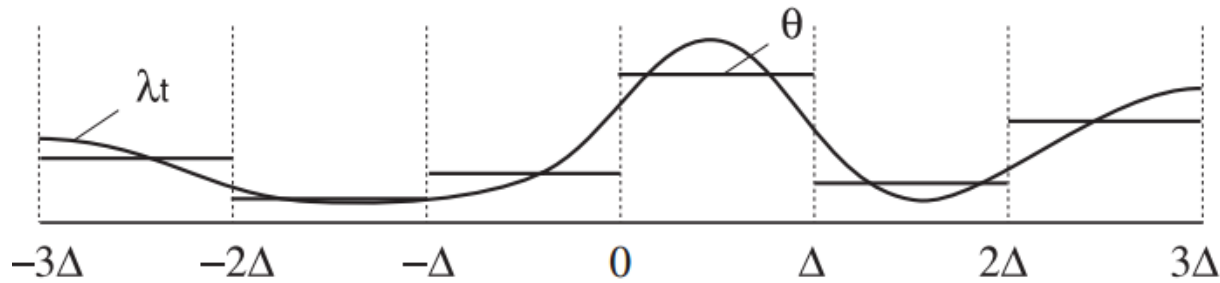
Gaussian process (Cunningham, Yu, Shenoy 2007)

BARS (DiMatteo et al. 2001)

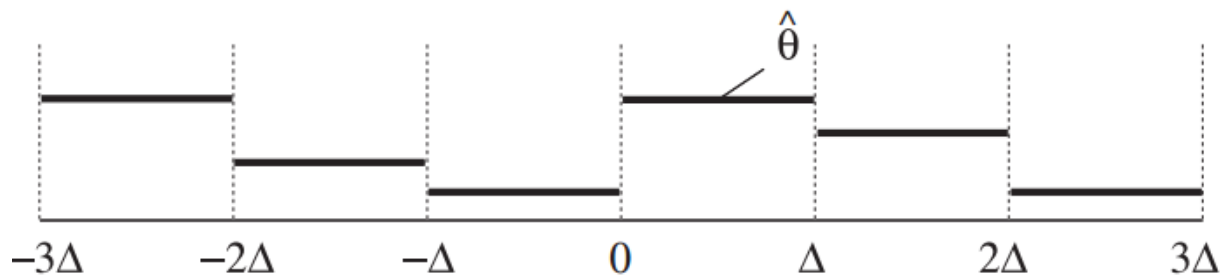
Bayesian binning (Endres et al. 2008)

Reviewed by Cunningham et al. 2009

BIN SIZE SELECTION



$$\theta = \frac{1}{\Delta} \int_0^{\Delta} \lambda_t dt$$

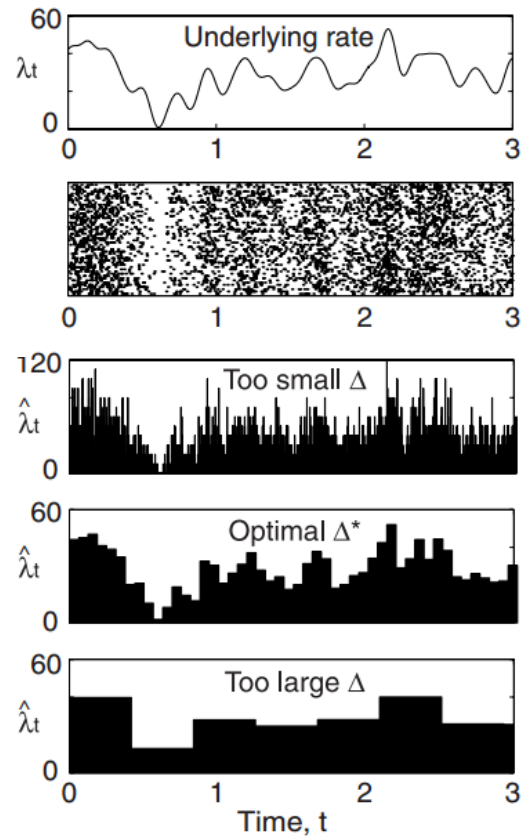
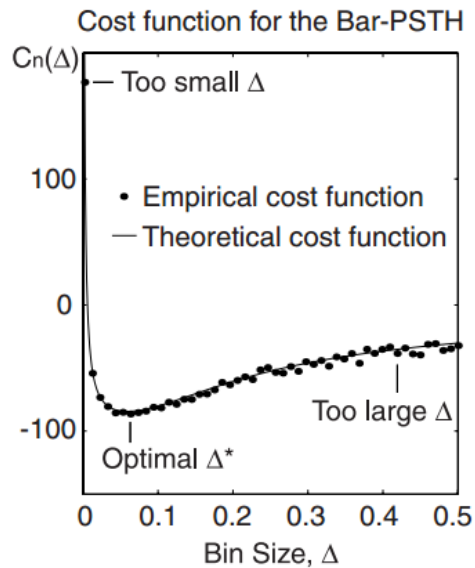


$$\hat{\theta} = \frac{k}{n\Delta}$$

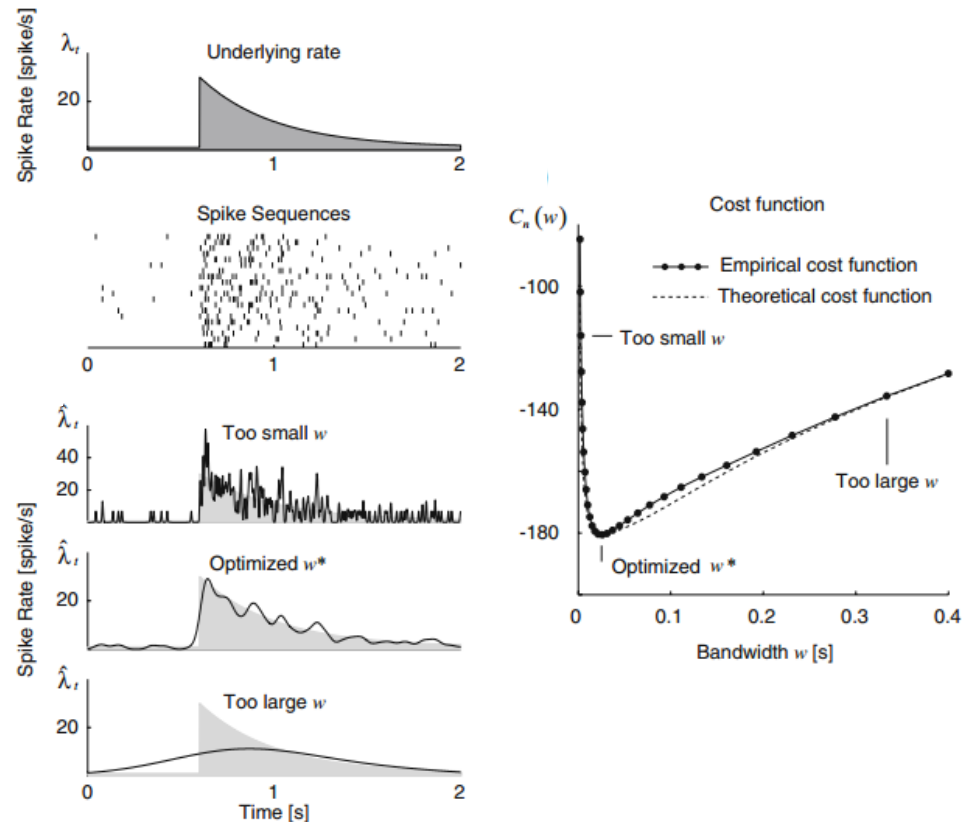
BIN SIZE SELECTION

- Divide the observation period T into N bins of width Δ .
- Count the number of spikes k_i from all n trials that fall into i -th bin
- Compute $\bar{k} = \frac{1}{N} \sum_i^N k_i$ and $v = \frac{1}{N} \sum_i^N (k_i - \bar{k})^2$
- Compute $C_n(\Delta) = \frac{2\bar{k}-v}{(n\Delta)^2}$
- Search for Δ^* that minimizes $C_n(\Delta)$

BIN SIZE SELECTION



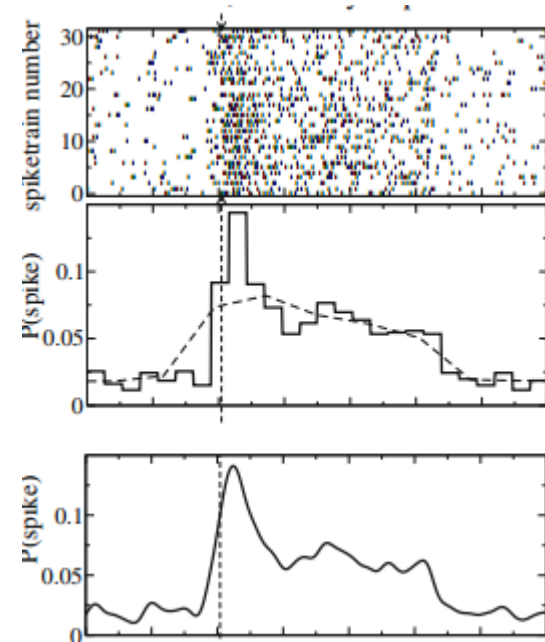
OPTIMAL KERNEL WIDTH



$$n^2 C_n(\omega) = \frac{N}{\omega} + \frac{2}{\omega} \sum_{i < j} \left[\exp\left(-\frac{(t_i - t_j)^2}{4\omega^2}\right) - 2\sqrt{2} \exp\left(-\frac{(t_i - t_j)^2}{2\omega^2}\right) \right]$$

PROBLEMS WITH PSTH/SDF

- **Problems with PSTH approaches**
 - Sharp transients
 - Many bins where rate is constant
- **Problems with SDFs**
 - Sharp transients are blurred

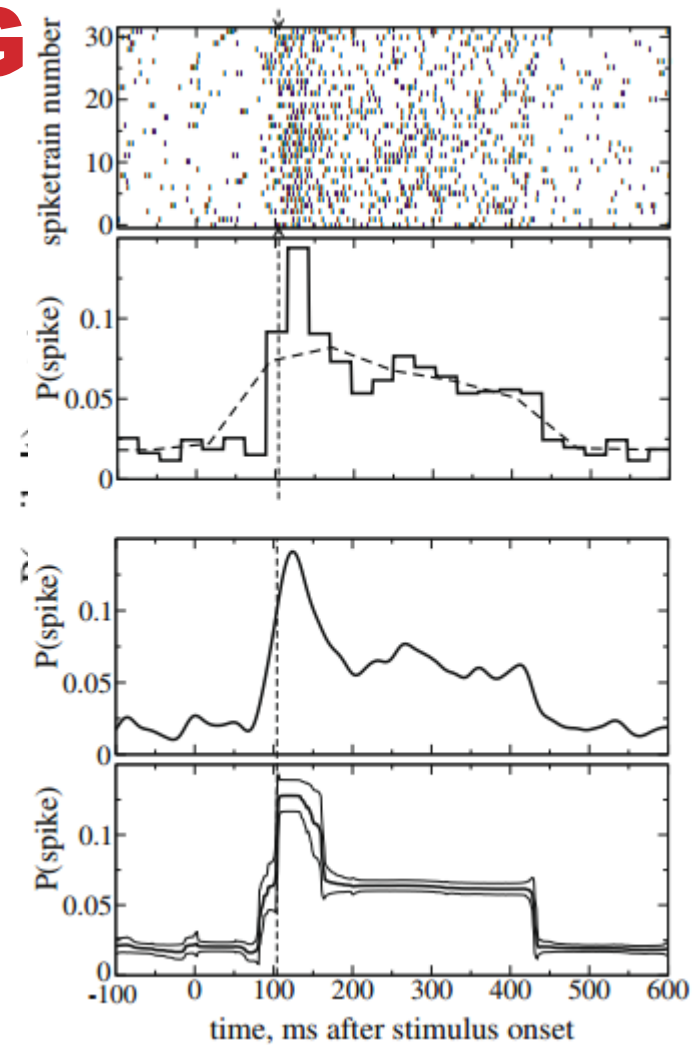


BAYESIAN BINNING

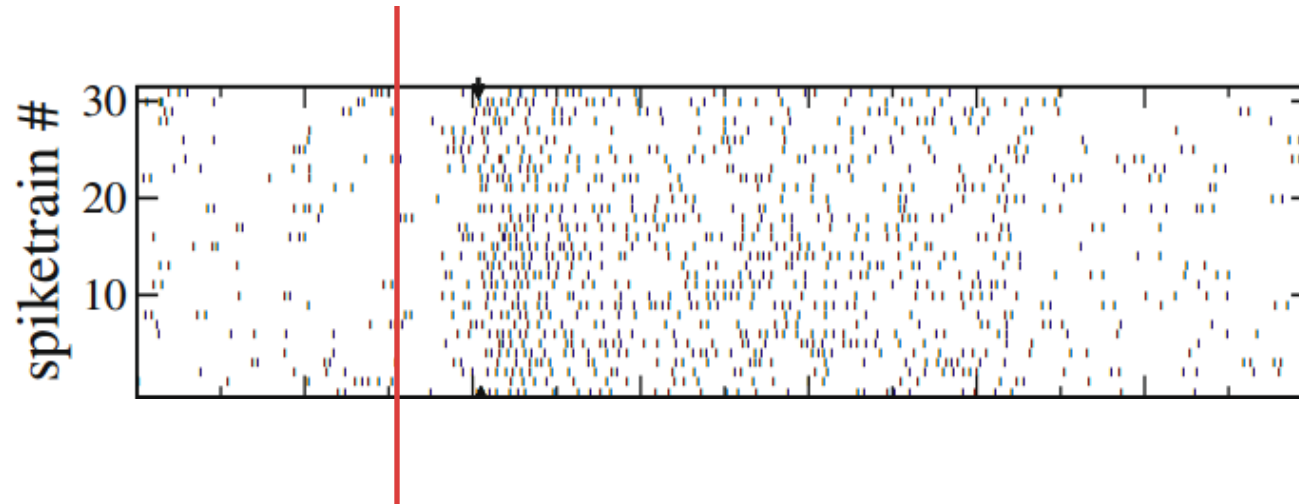
- Model PSTH as sequence of intervals with constant firing probability

$$P(\vec{z}^i | \{f_m\}, \{k_m\}, M) = \prod_{m=0}^M f_m^{s(\vec{z}^i, m)} (1 - f_m)^{g(\vec{z}^i, m)}$$

- Bayesian inference for model parameters by computing posterior
- Dynamic programming
- Code online: <http://mloss.org/software/view/67/>



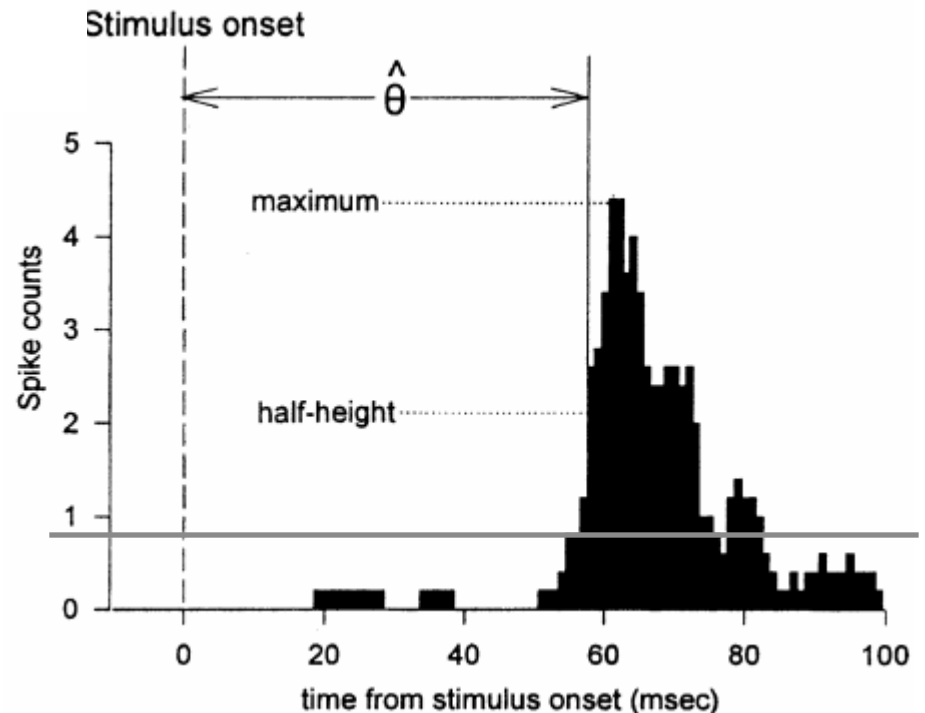
LATENCY DETECTION



„Latency is where is the signal starts“

LATENCY DETECTION

- Non-parametric estimation as half height of peak firing rate
- Threshold criterion above spontaneous firing



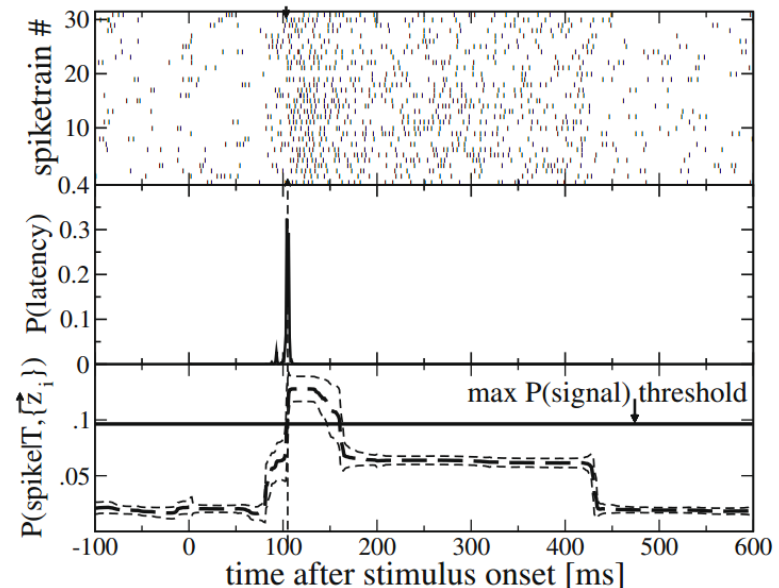
LATENCY DETECTION

- Change point detection

$$s(t) = \text{Poisson}(\lambda_1), t \in [0, \theta)$$

$$s(t) = \text{Poisson}(\lambda_2), t \in [\theta, \kappa)$$

- Bayesian binning



SIMPLE OR COMPLEX CELL?

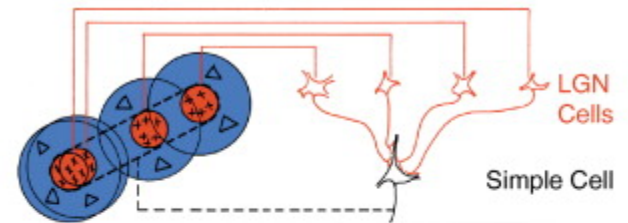
Simple cells:

- Selective for spatial frequency and orientation
- Modulated by phase of a grating

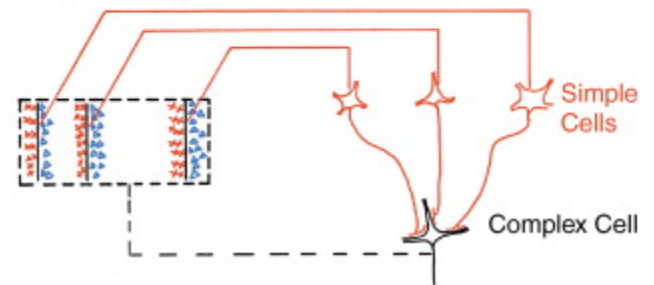
Complex cells:

- Selective for spatial frequency and orientation
- Invariant to phase of the grating

Circuit Building a Simple Cell from LGN Cells



Building a Complex Cell from Simple Cells



STIMULUS

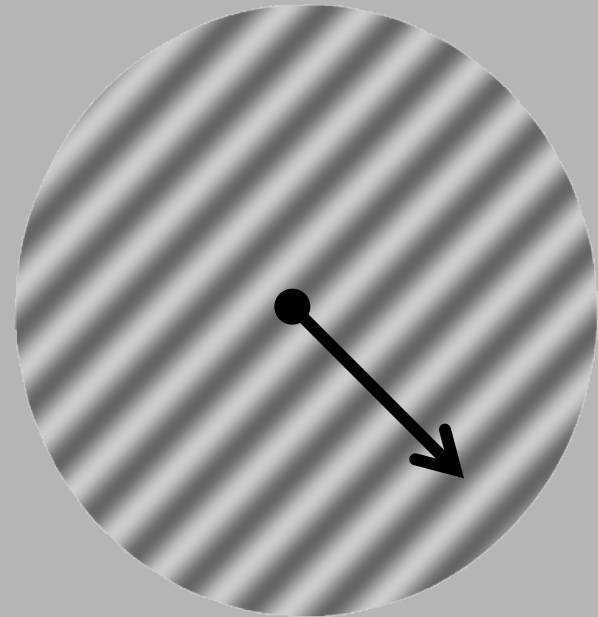
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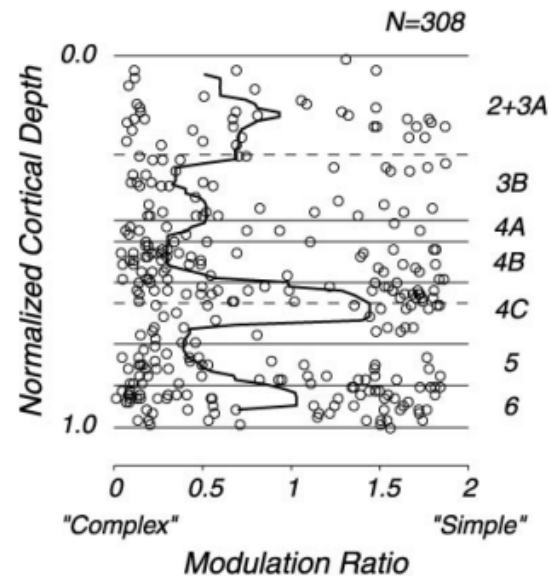
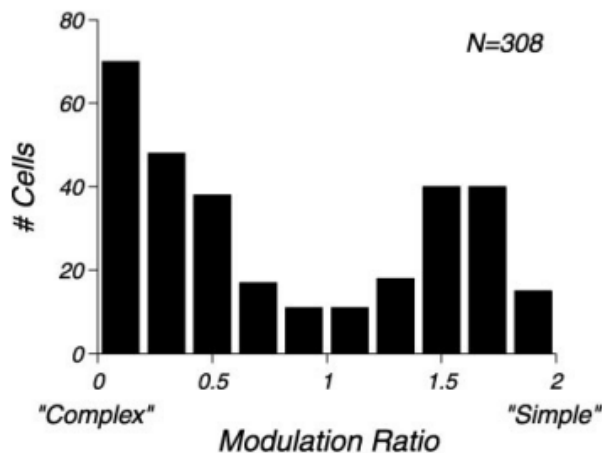
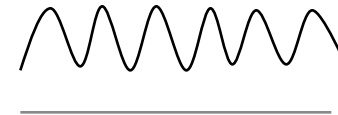
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LINEARITY INDEX

Compare oscillation at stimulus frequency to DC component

$$LI = \frac{F_1}{F_0}$$



SIMPLE OR COMPLEX?

