

Statistical methods for understanding neural codes:

Multineuronal spike coding in primate retina

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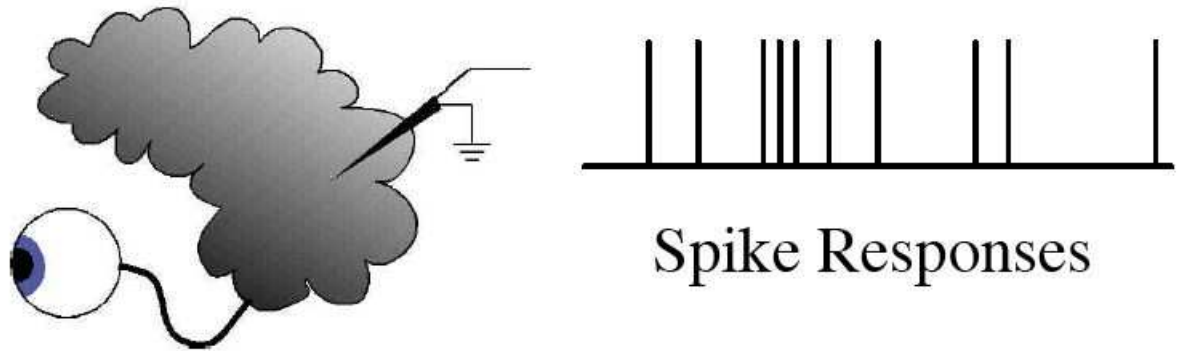
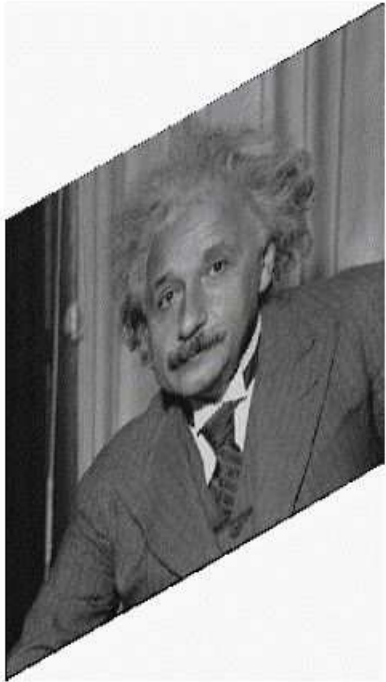
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— with J. Pillow (Gatsby), E. Simoncelli (NYU), E.J. Chichilnisky, J. Shlens (Salk), T. Toyoizumi (Columbia).

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The neural code



Input-output relationship between

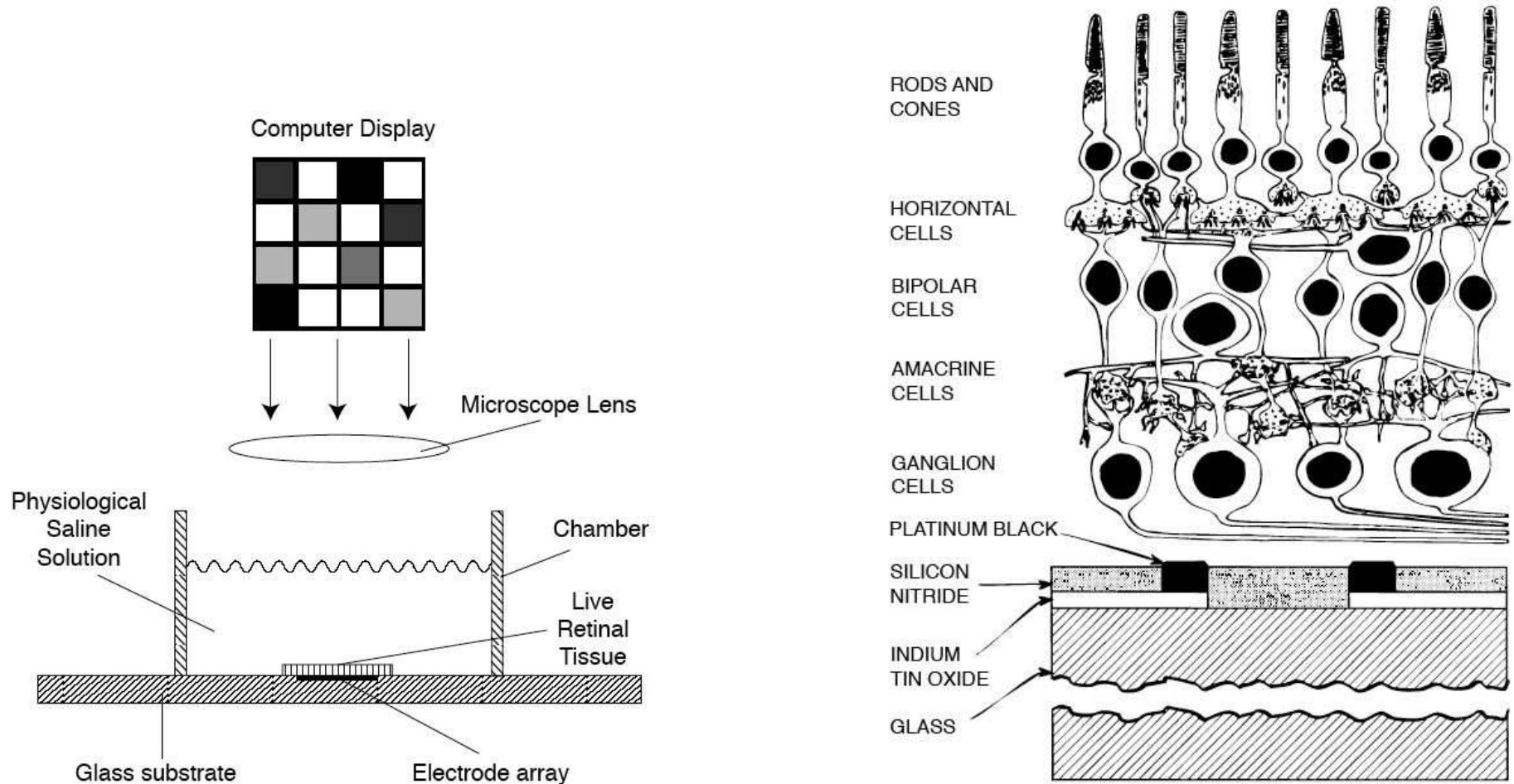
- External observables x (sensory stimuli, motor responses...)
- Neural variables y (spike trains, population activity...)

Probabilistic formulation: $p(y|x)$

Retinal ganglion neuronal data

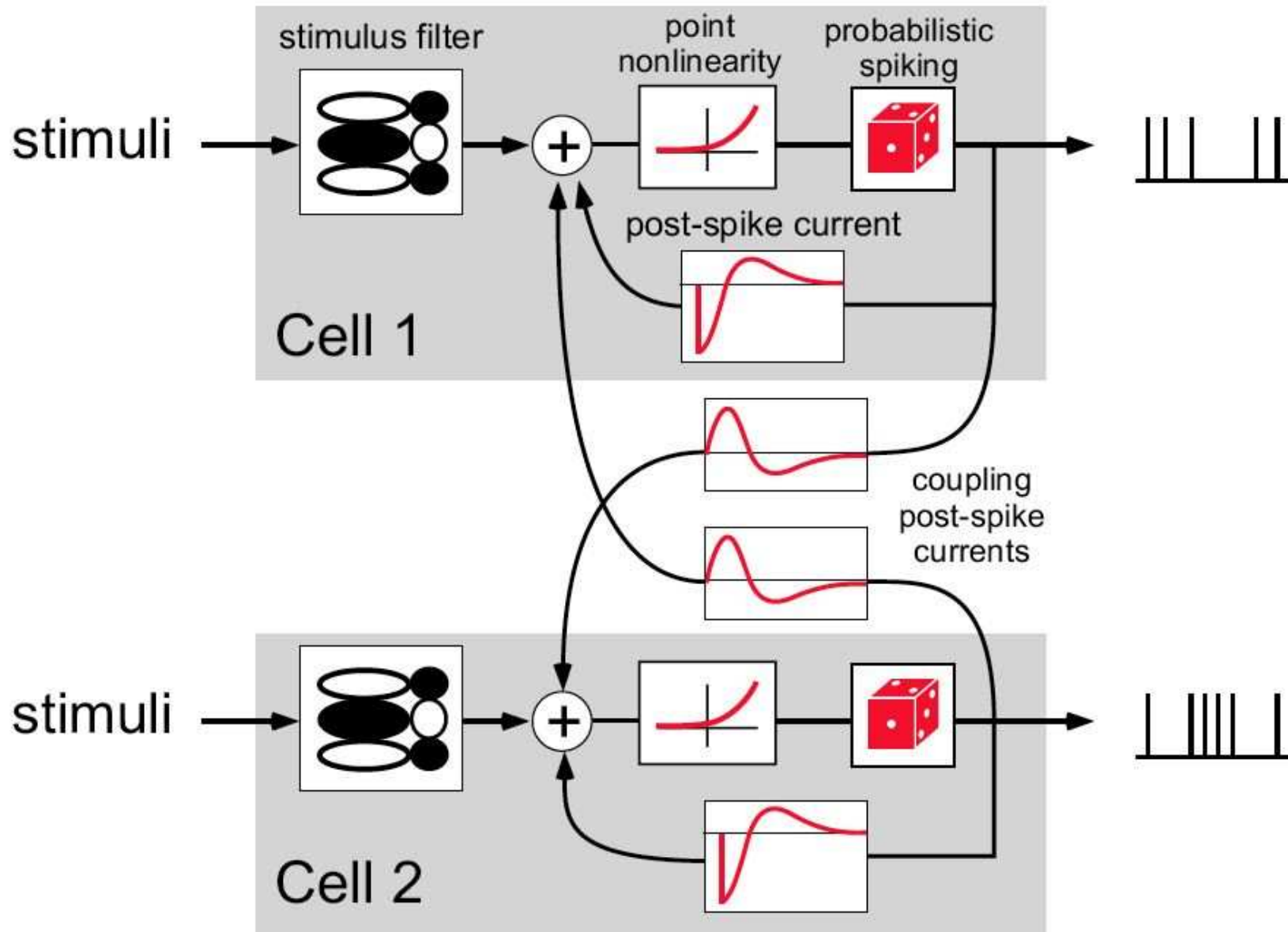
Preparation: dissociated salamander and macaque retina

— extracellularly-recorded responses of populations of RGCs



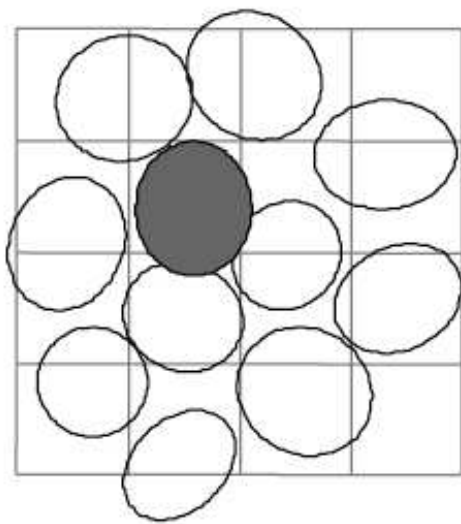
Stimulus: random spatiotemporal “flicker” visual stimuli

Multineuronal generalized linear model

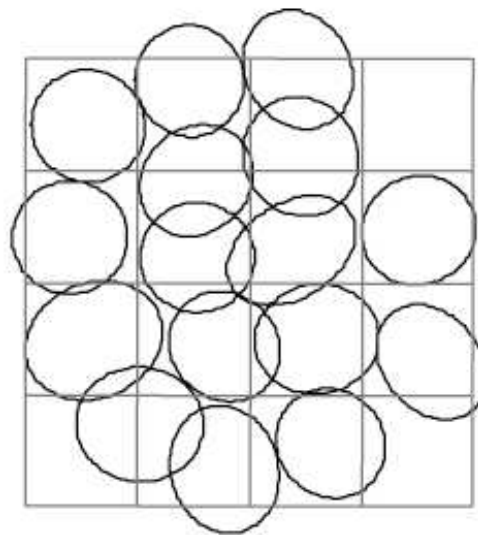


— Fit by L1-penalized max. likelihood (concave optimization)

ON
cells



OFF
cells

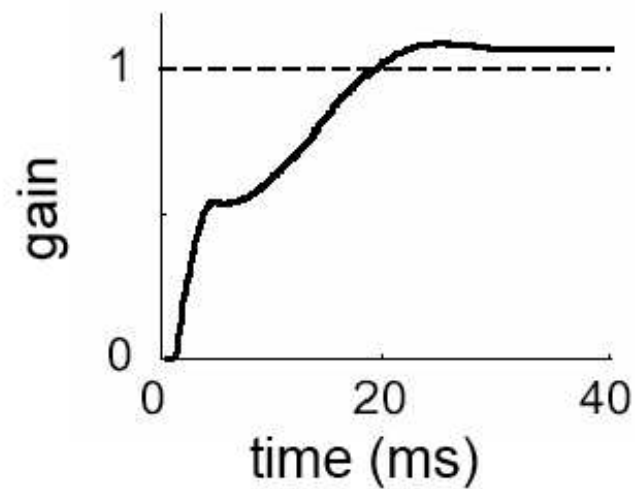
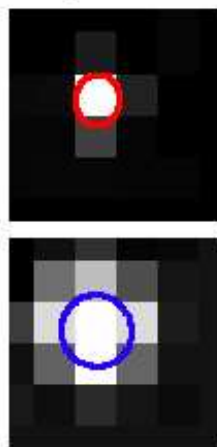
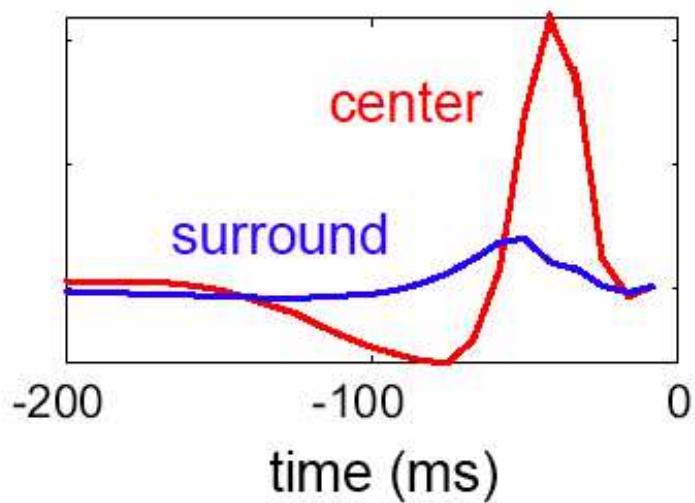


stimulus filter

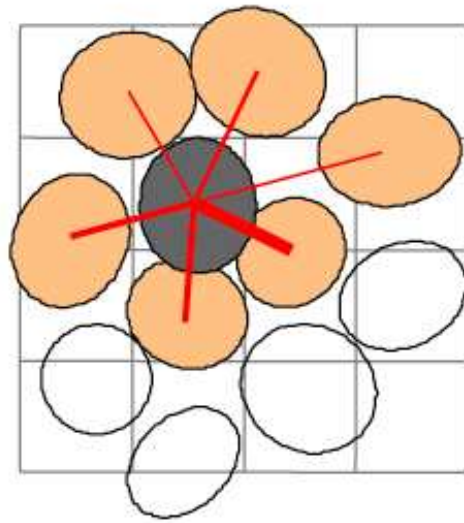
post-spike filter

temporal

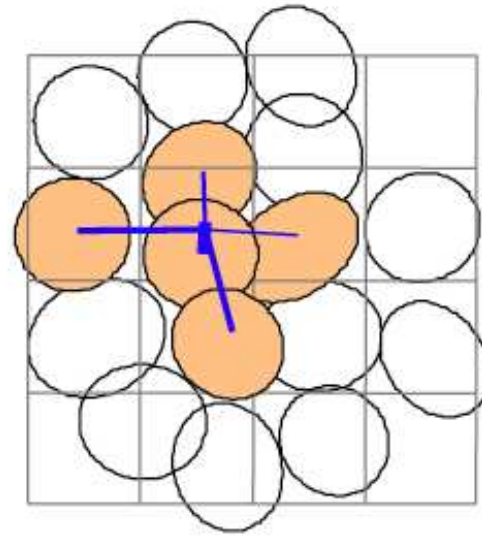
spatial



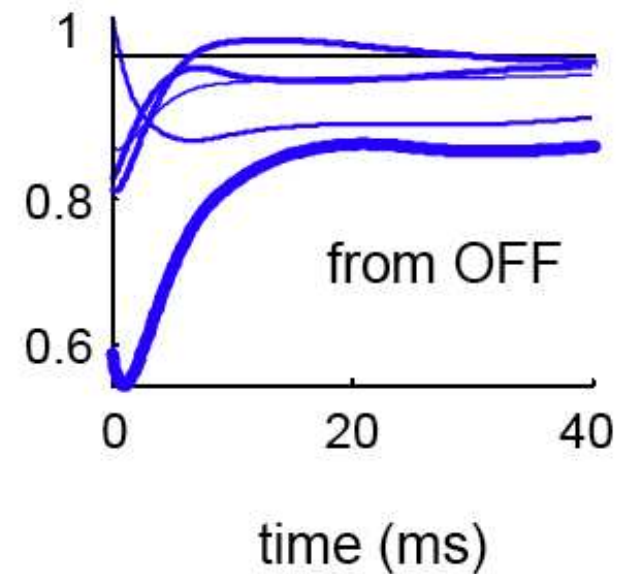
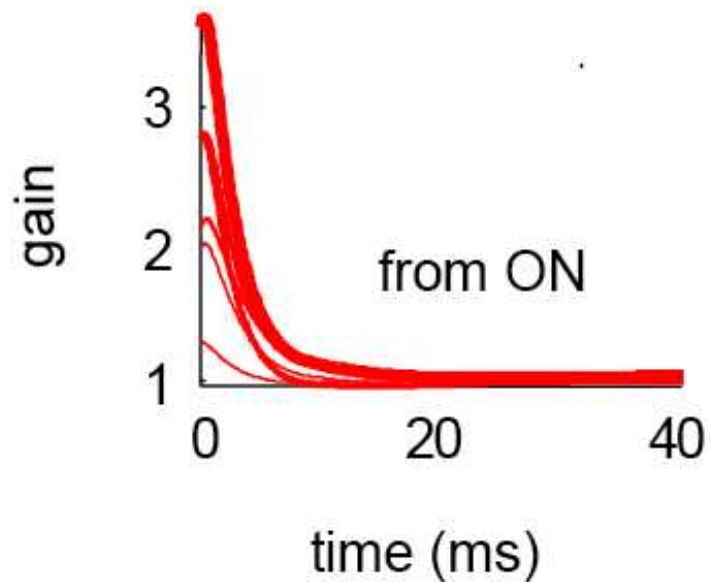
ON
cells



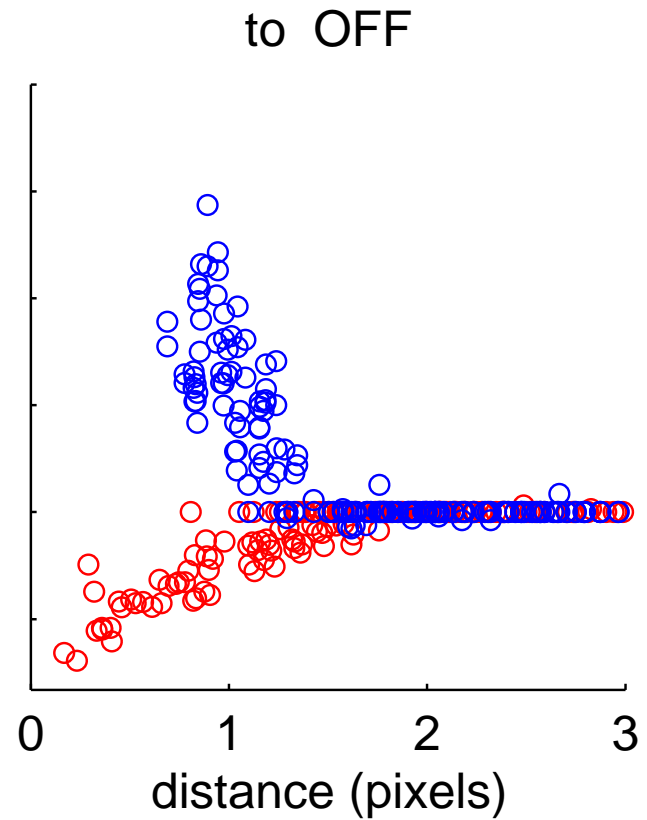
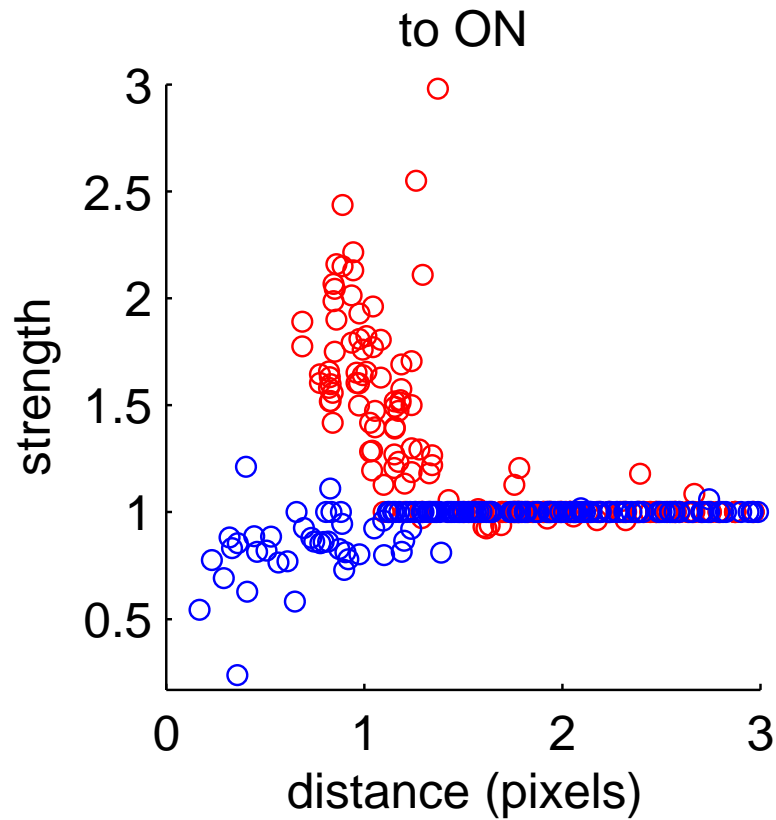
OFF
cells



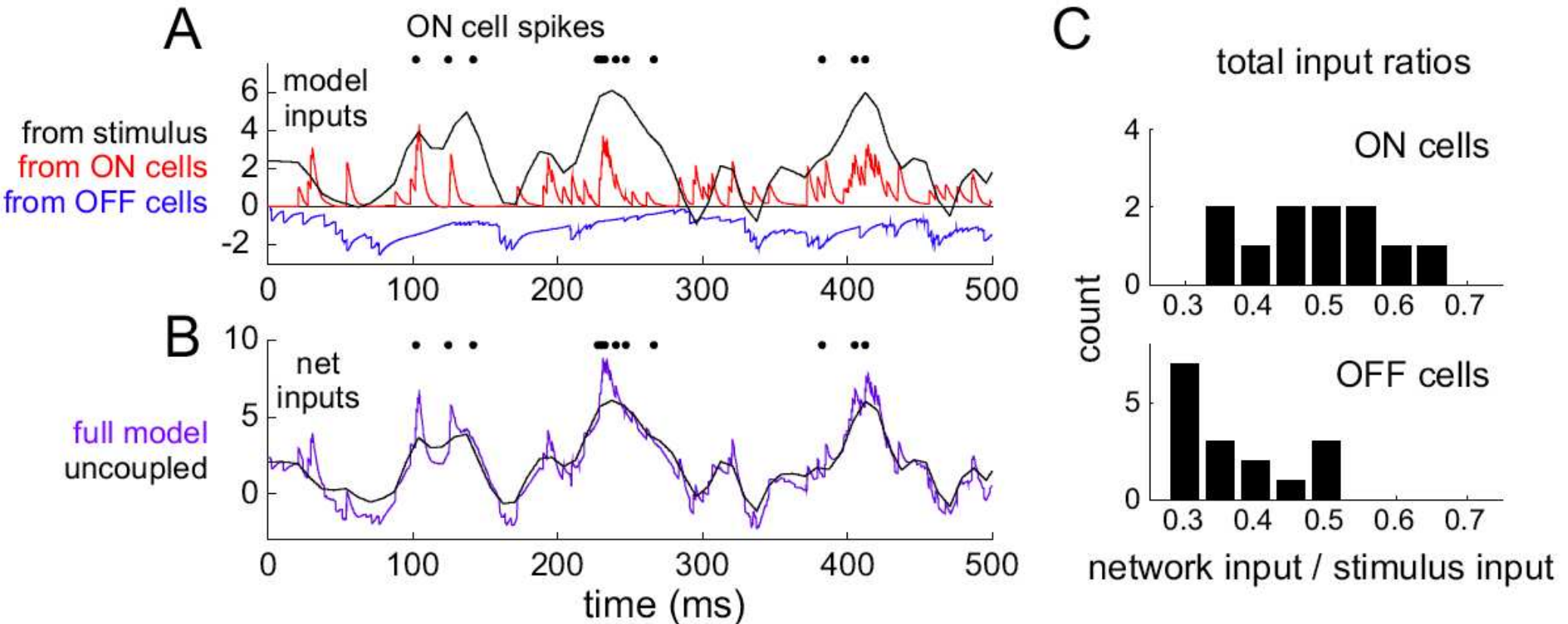
coupling filters



Nearest-neighbor connectivity



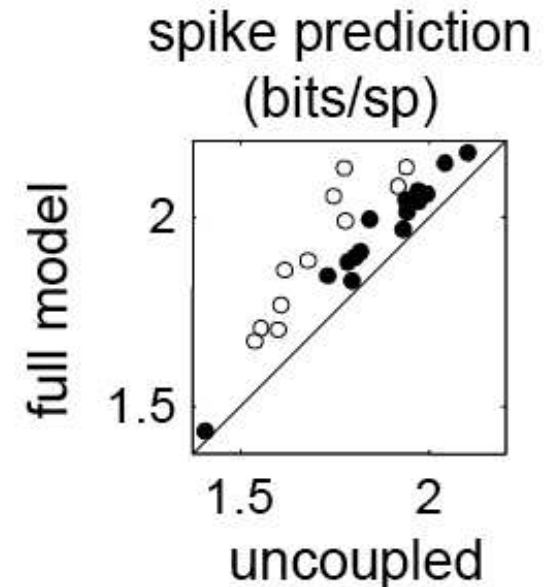
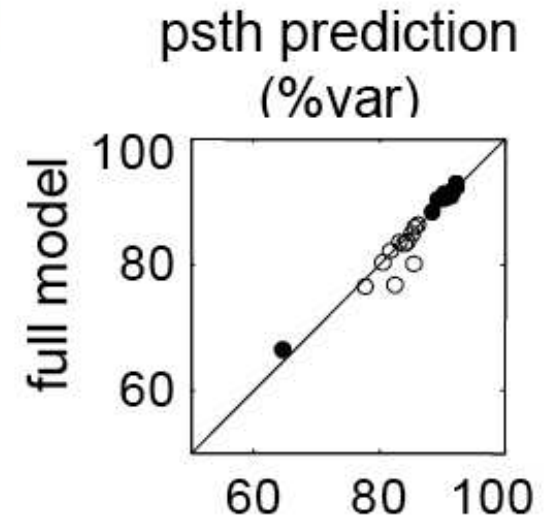
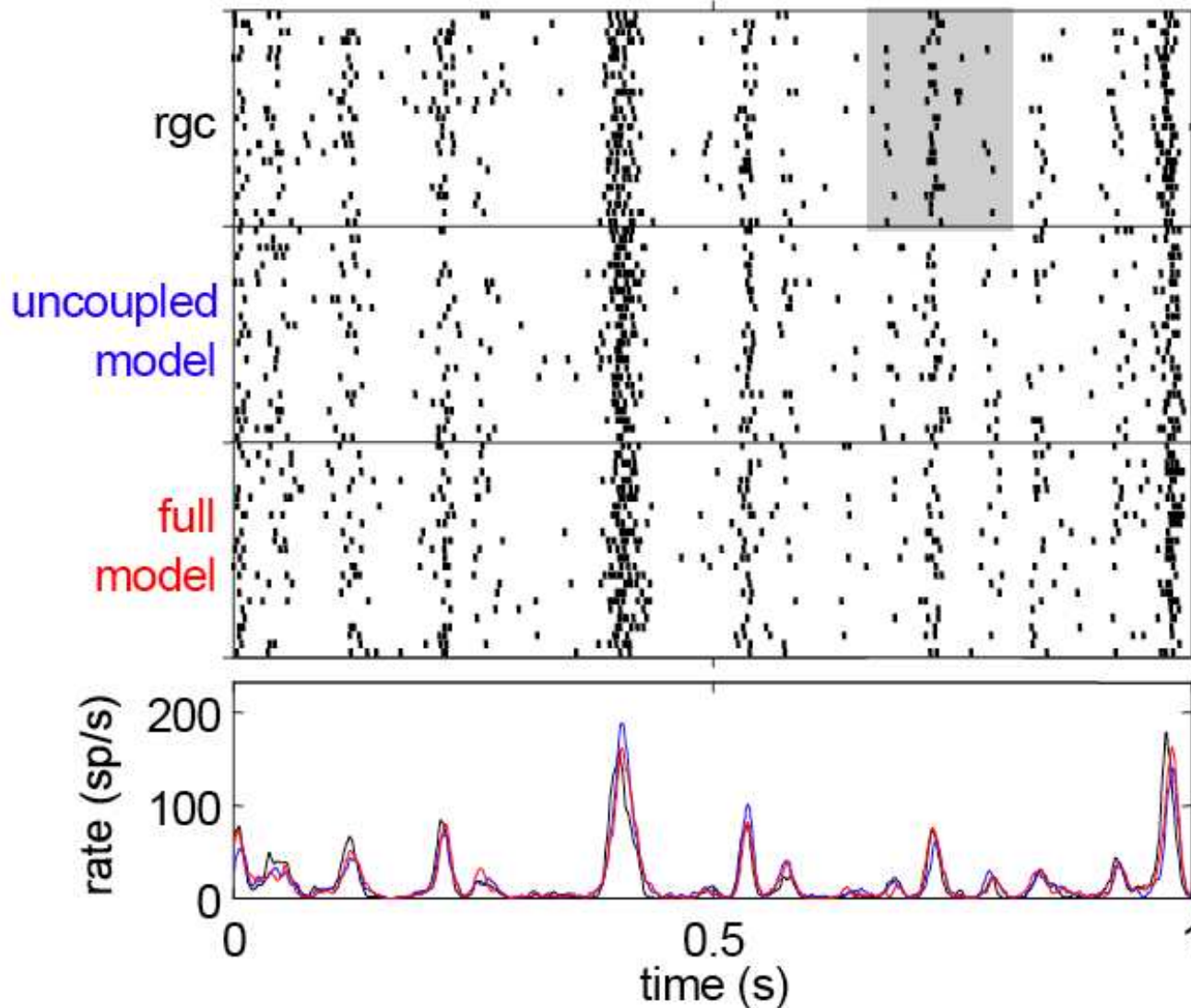
Network vs. stimulus drive



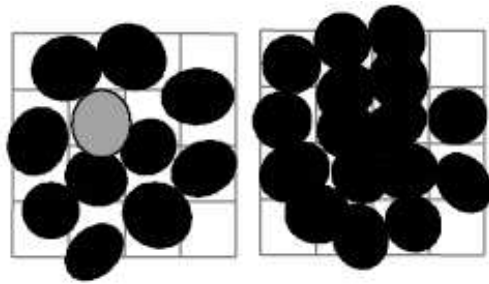
— Network effects are $\approx 50\%$ as strong as stimulus effects

Spike Train Prediction

- improved prediction, but not of mean rate!

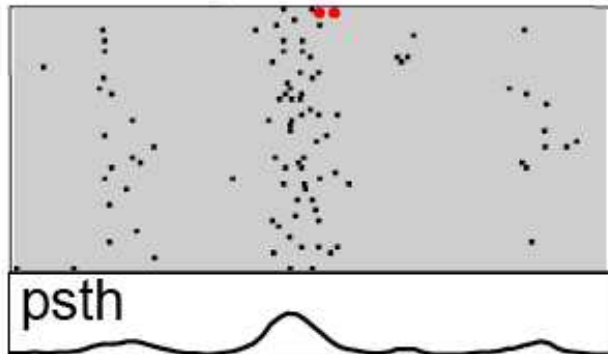


Network predictability analysis



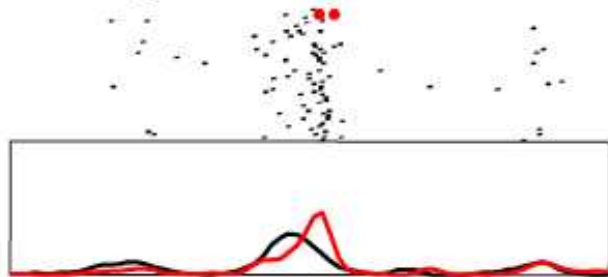
rgc raster

- fix all other neurons for a single trial

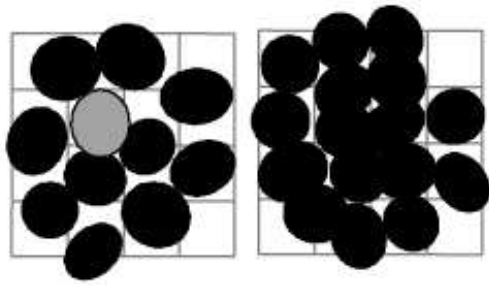


psth

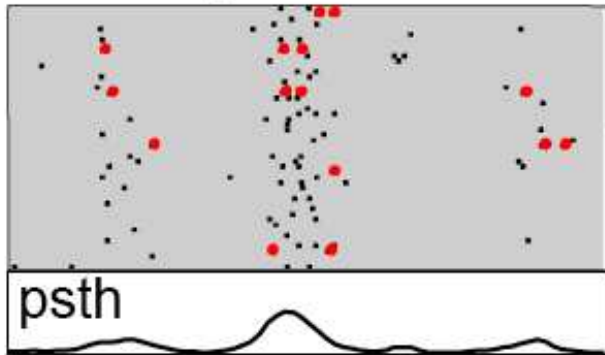
single-trial prediction



- draw single-trial predictions of this cell's spike train

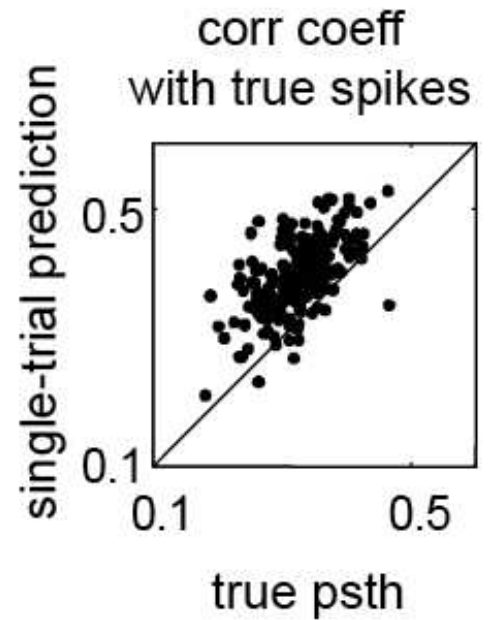
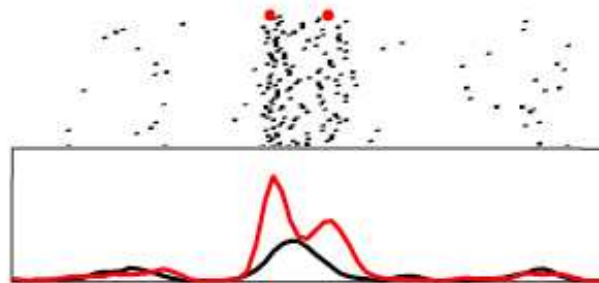
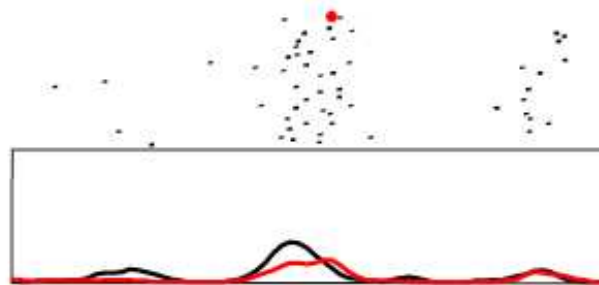
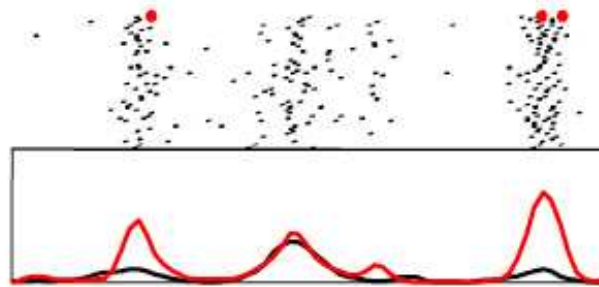
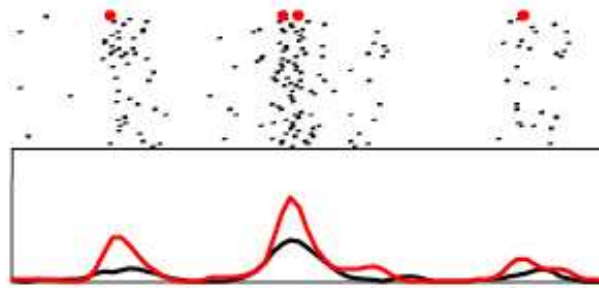
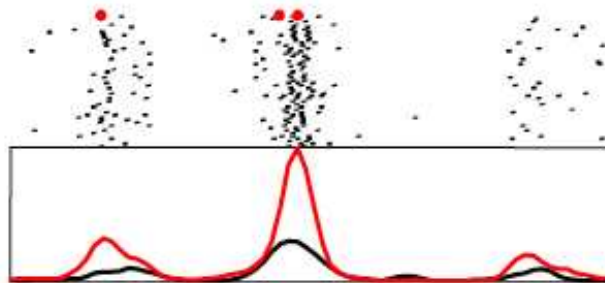
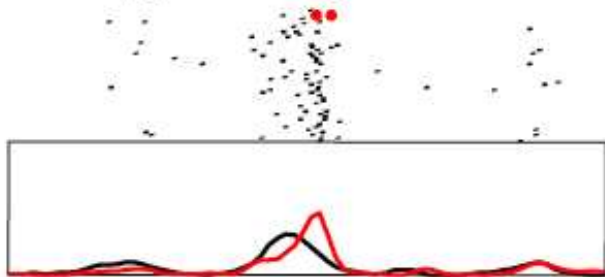


rgc raster



psth

single-trial prediction



- single-trial variability predicted by local network activity

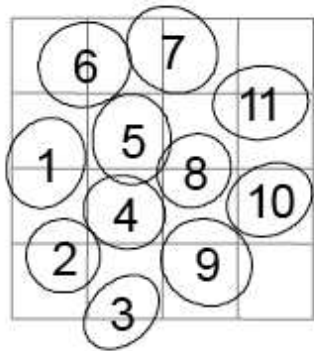
Model captures spatiotemporal cross-corrs

x-corrs:

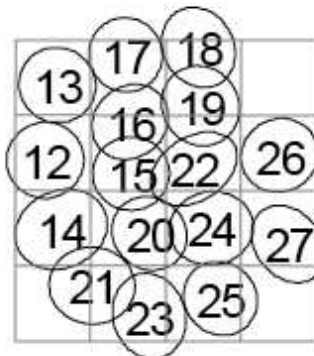
ON-ON

OFF-OFF

ON cells

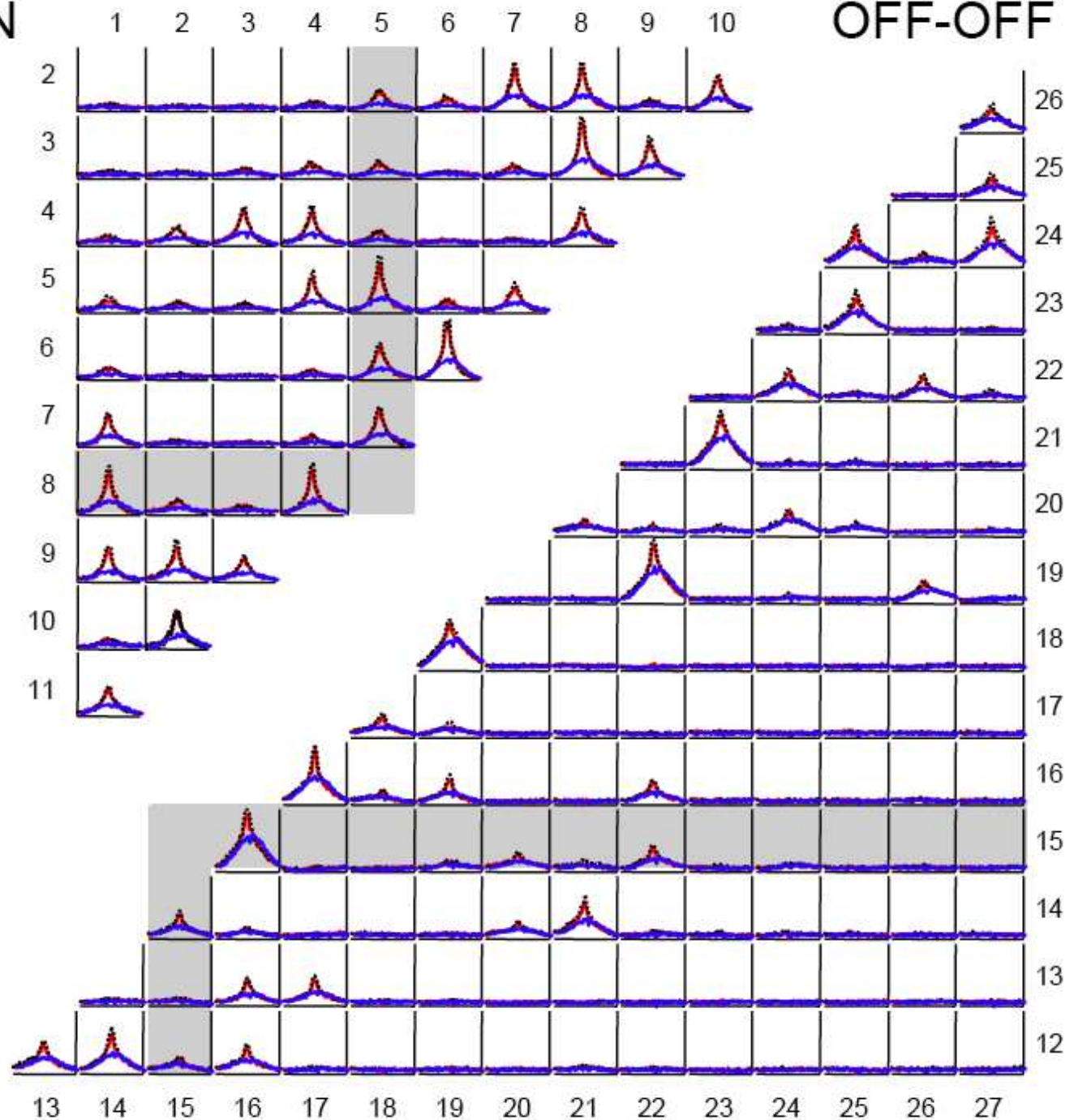


OFF cells



75 sp/s

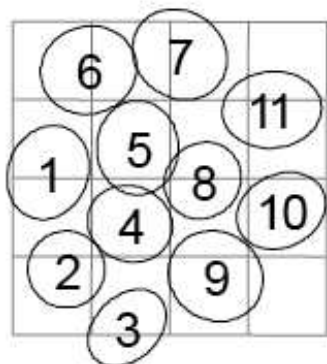
50 ms



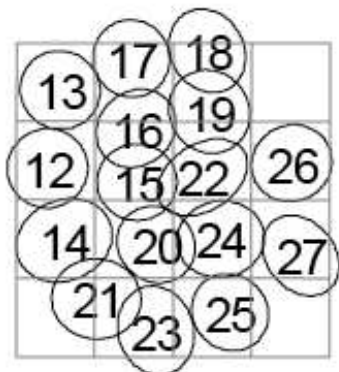
X-corrs:

ON-OFF

ON cells

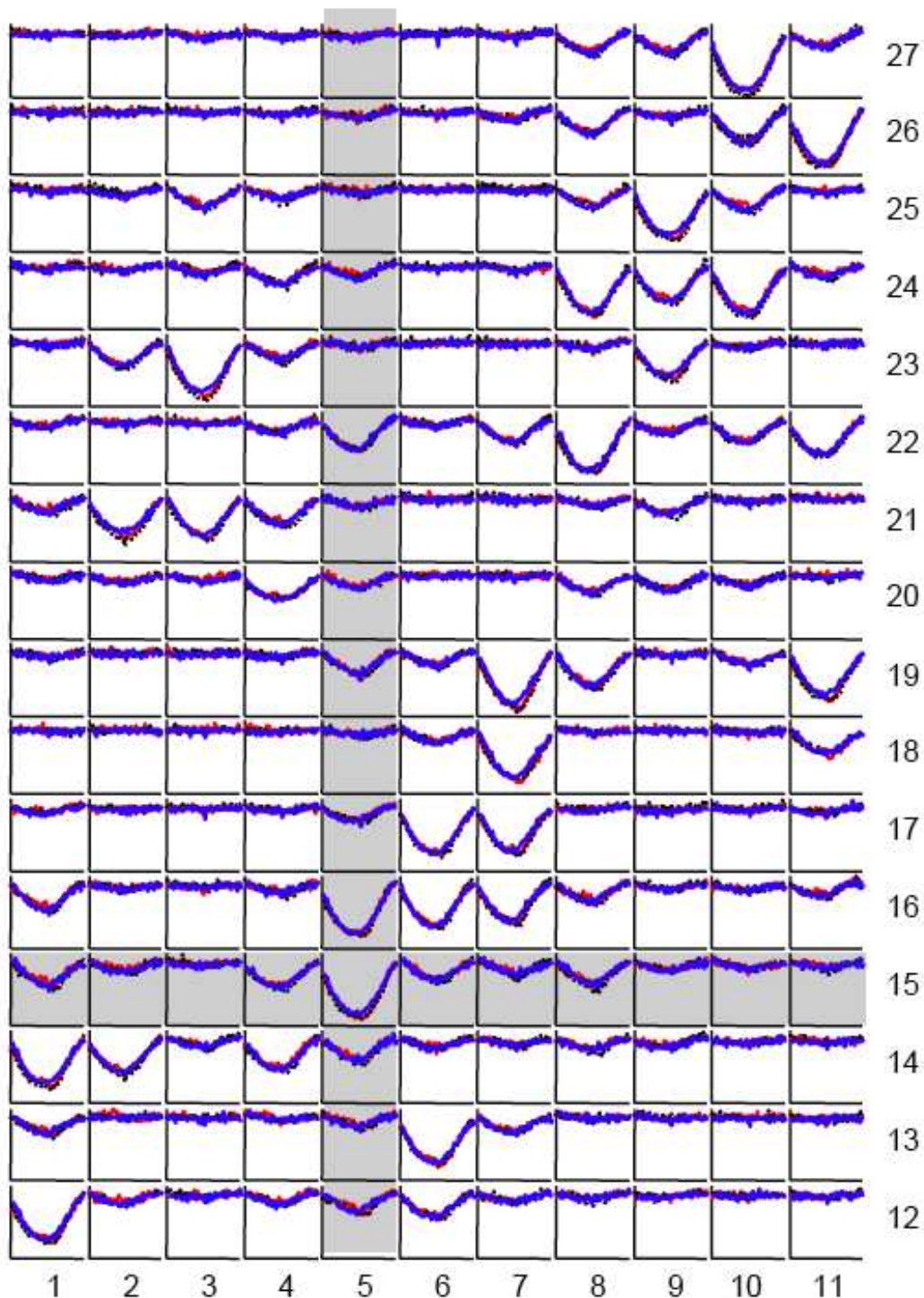


OFF cells



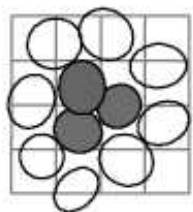
37 sp/s

50 ms

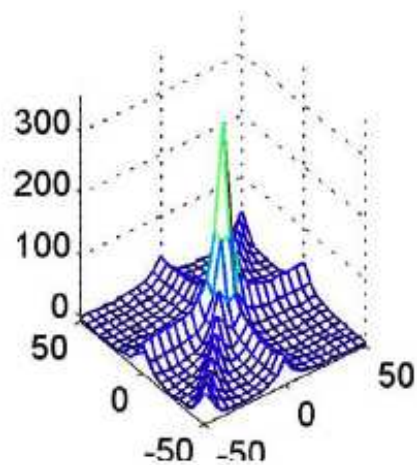


Triplet correlations

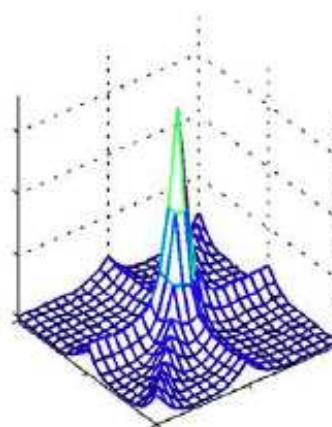
3 ON cells



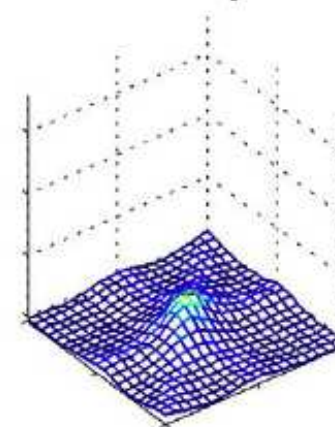
RGC



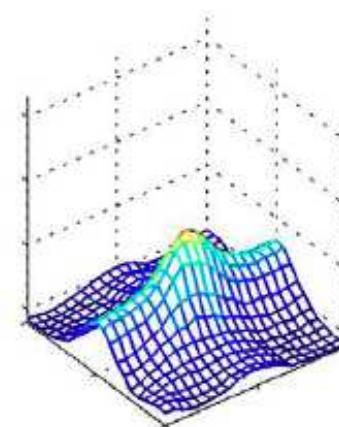
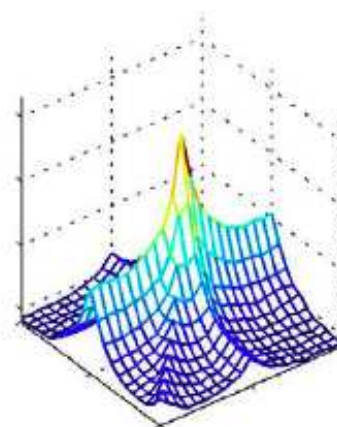
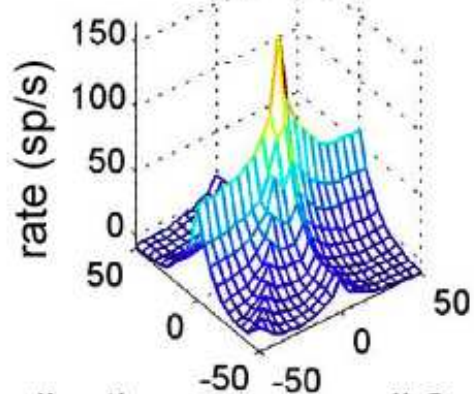
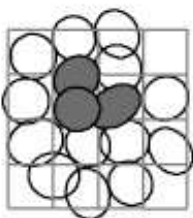
full model



uncoupled



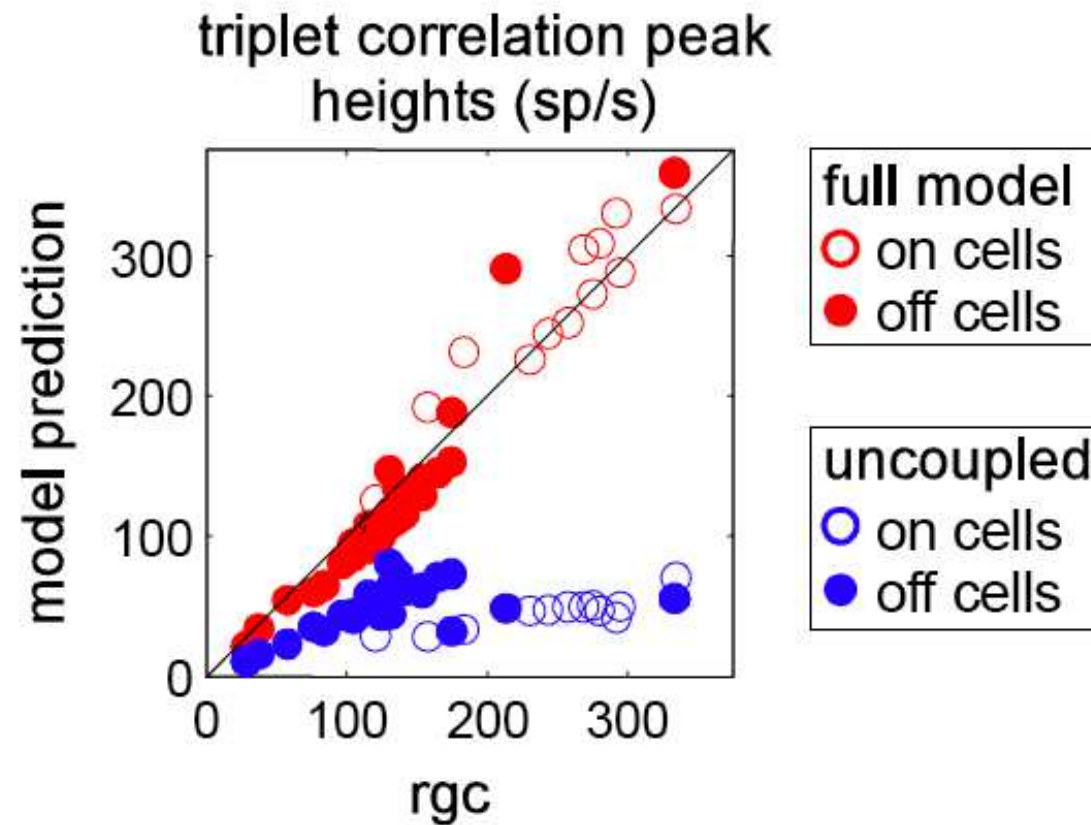
3 OFF cells



cell 1 spike time

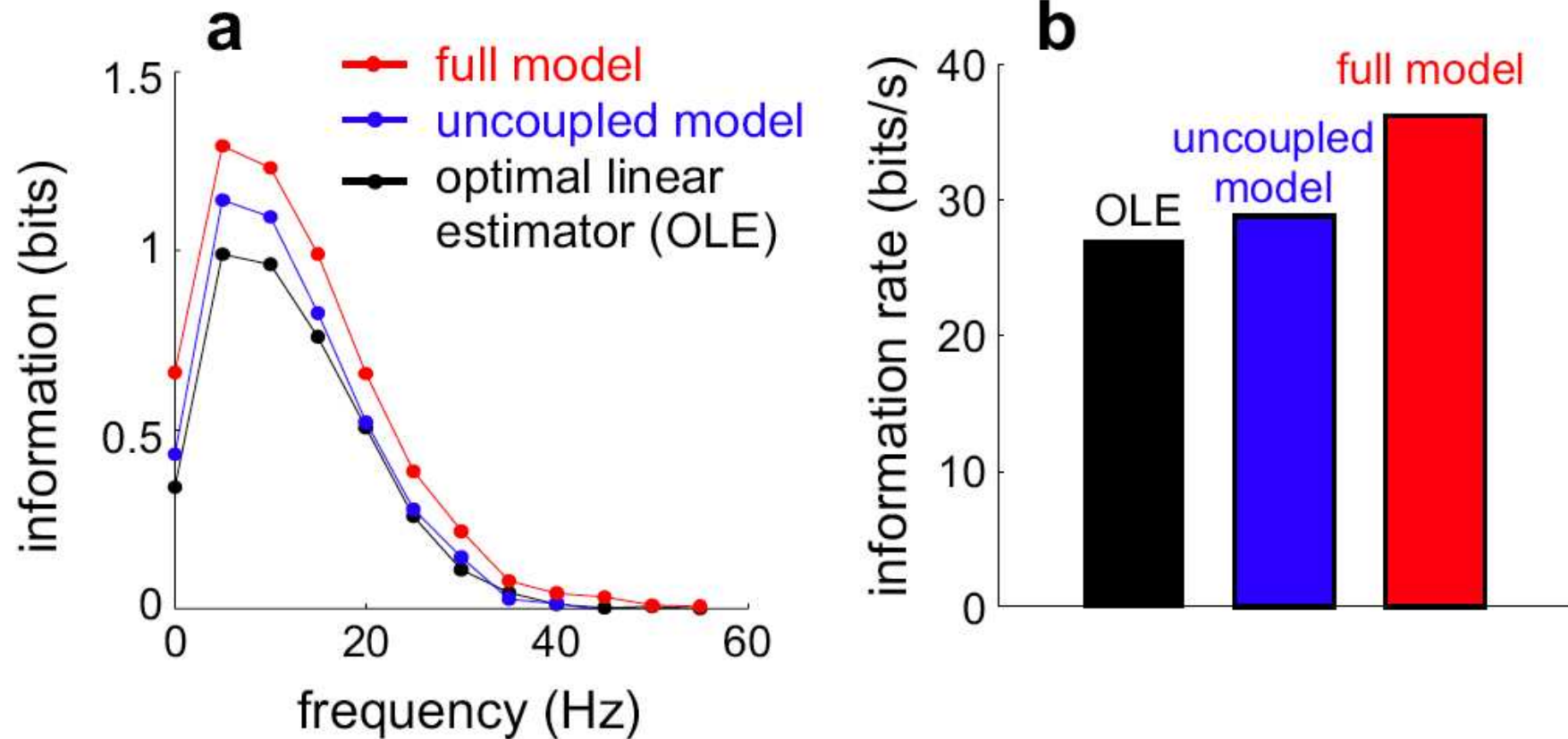
cell 2 spike time

Triplet correlations



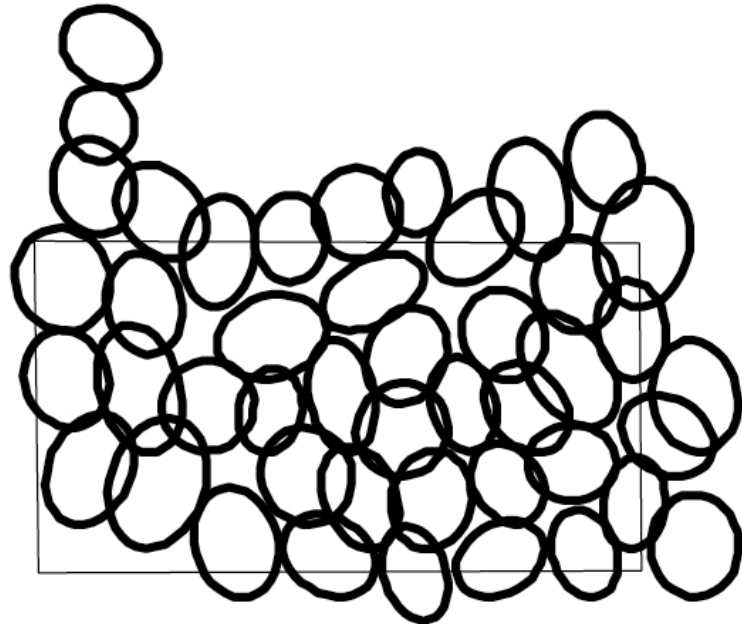
Optimal Bayesian decoding

— Compute $E(stim|resp)$ via MCMC under each model



— Including network terms improves decoding accuracy.

Next: Large-scale network modeling



ON-Parasol



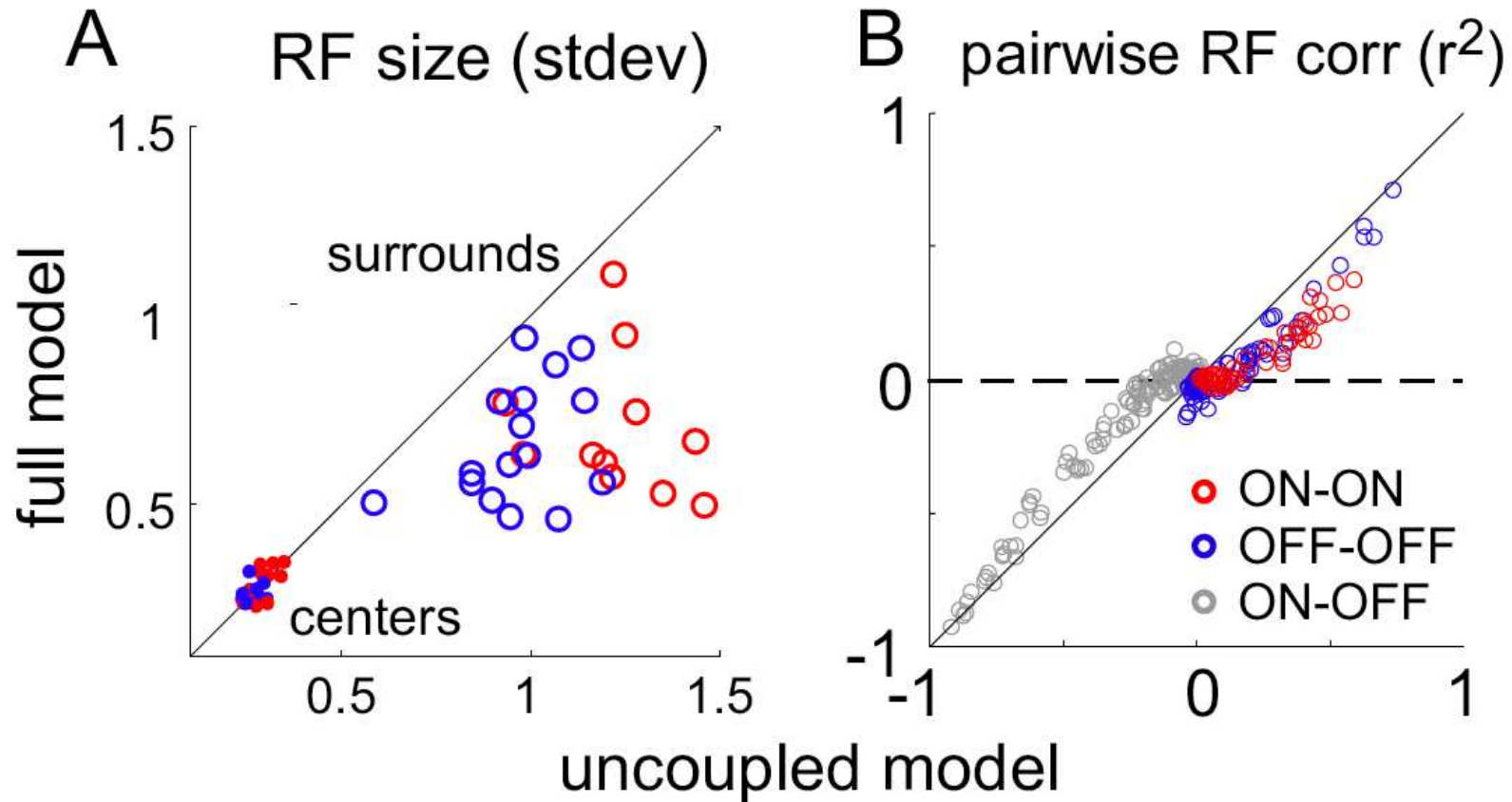
OFF-Parasol

— Do observed local connectivity rules lead to interesting network dynamics? What are the implications for retinal information processing?

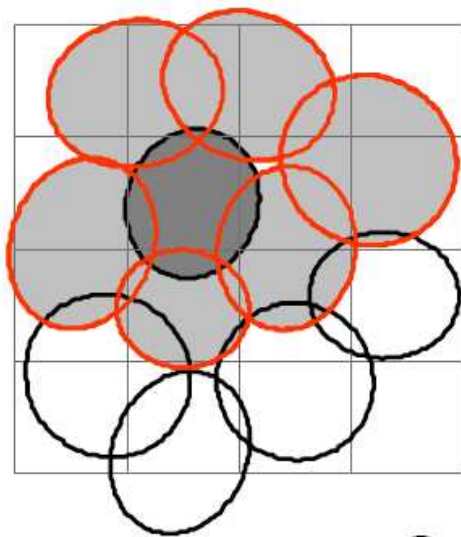
References

- Pillow, J. and Paninski, L. (2007). Model-based decoding, information estimation, and change-point detection in multi-neuron spike trains. *Submitted*.
- Pillow, J., Paninski, L., Shlens, J., Simoncelli, E., and Chichilnisky, E. (2005). Modeling multi-neuronal responses in primate retinal ganglion cells. *Comp. Sys. Neur.* '05.

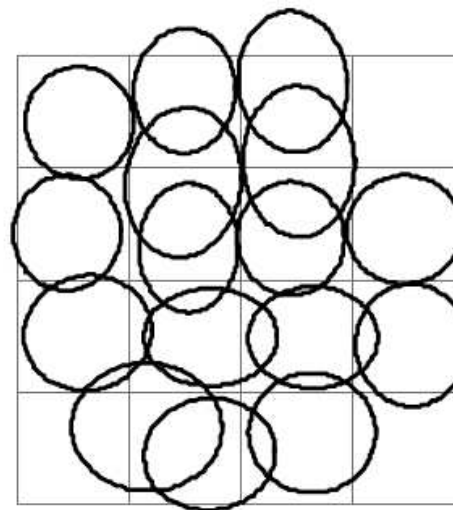
Fitting coupling terms exposes smaller receptive fields



ON
cells



OFF
cells



Cross-Correlations

