

RECEPTIVE FIELDS

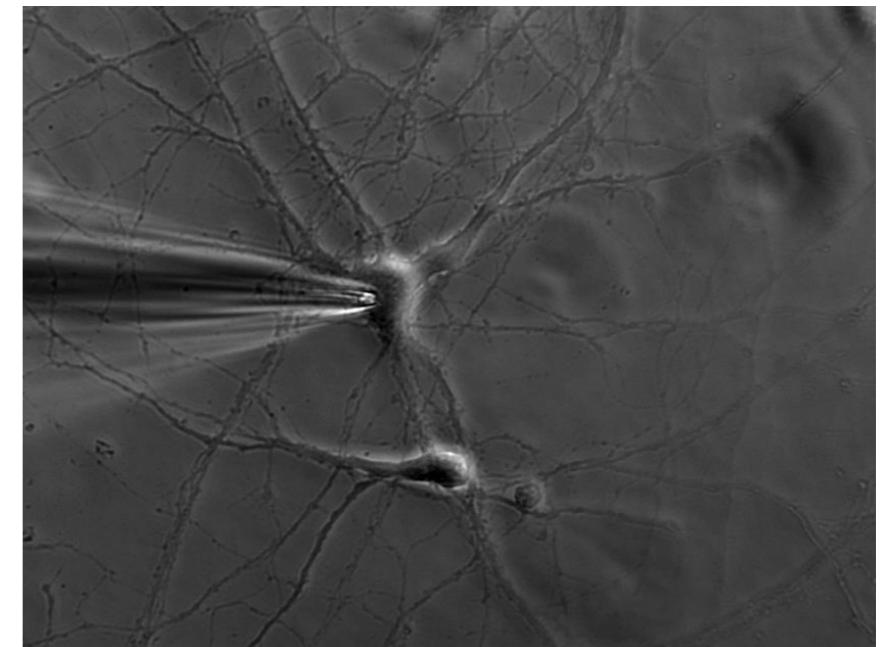
11.12.2018

PROBLEM SET 4

- * is posted!
- * due in two weeks (11/26)

VISUAL EXPERIMENT

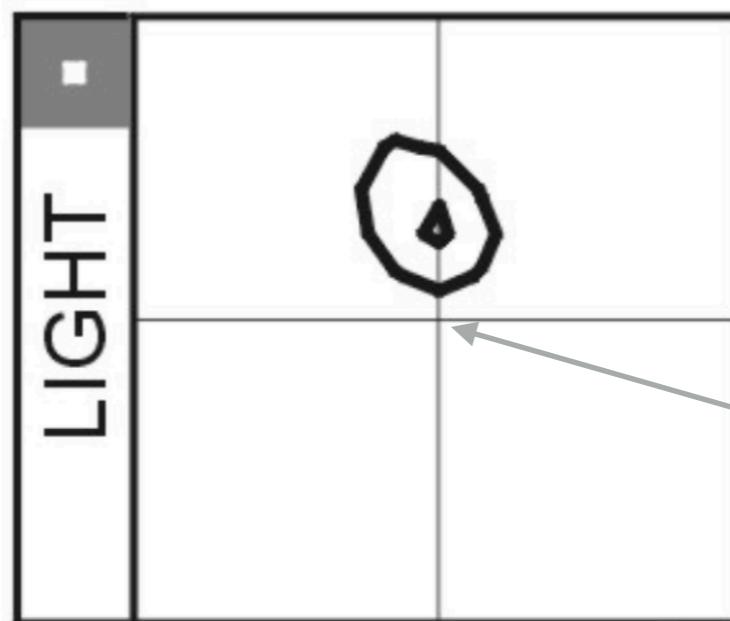
- * suppose we've placed an electrode into primary visual cortex (V1) to record spikes from a single neuron
- * we can choose what visual stimuli to display, and know where our subject is fixating
- * how do we figure out what this neuron is doing?



VISUAL EXPERIMENT

- * each neuron in V1 has a **receptive field**
 - * this is the region of visual space where stimuli cause the neuron to fire

an example
receptive field



fixation point

RECEPTIVE FIELDS

- * the same idea applies to other sensory neurons
 - * in *somatosensory cortex*, each neuron has a receptive field localized somewhere on the body (e.g. the top of the left pinky finger)
 - * in *auditory cortex*, each neuron has (typically) a spectral receptive field, meaning it only responds to certain frequencies of sound

RECEPTIVE FIELDS

- * this leads to a more general definition of a receptive field:
 - * a receptive field is the portion of a “sensory space” that can elicit neural responses when stimulated
- * so what is a “sensory space”? how is it different from, e.g., visual space?

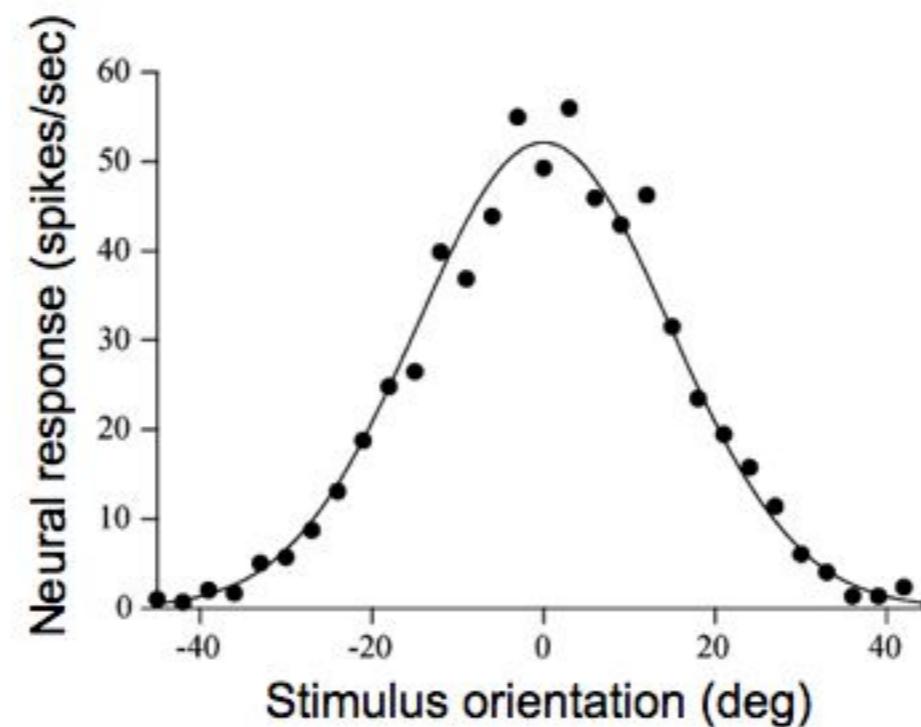
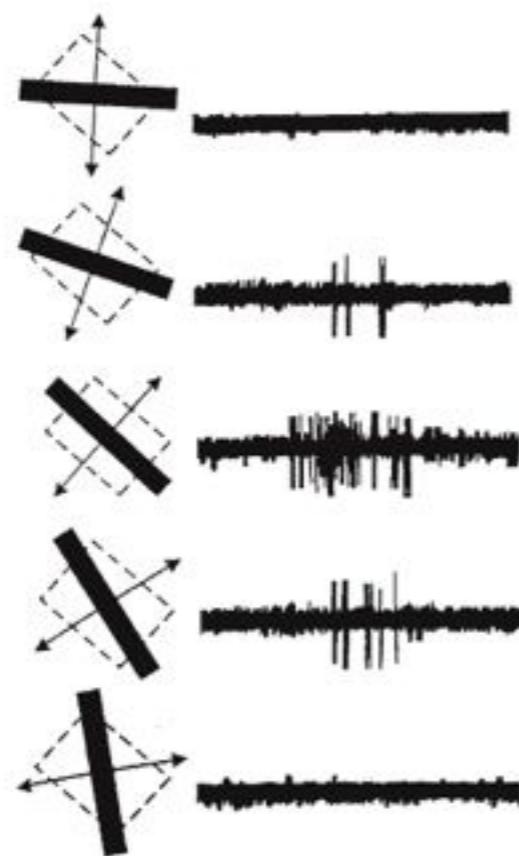
VISUAL SENSORY SPACE

- * what are some other visual “features” that are important components of the visual sensory space?

VISUAL SENSORY SPACE

- * suppose we choose some other dimension of visual sensory space, like orientation
- * we can measure the response of a V1 neuron to each orientation of a stimulus
- * this gives us an “orientation tuning curve”

VISUAL EXPERIMENT



Hubel & Wiesel, 1968

VISUAL SENSORY SPACE

- * now it's getting a bit complicated.
neurons in V1 have:
 - * a visual receptive field
 - * an orientation tuning curve
 - * an ocular tuning curve
 - * “simple” or “complex” behavior (response to moving stimuli)

GENERALIZED RECEPTIVE FIELDS

- * we could summarize all of these properties as a function
- * suppose that Y is the firing rate of our neuron & X is the stimulus
- * we can relate the two through a function,
$$f(X) = Y,$$
where f is a function that predicts the response of the neuron to any stimulus

GENERALIZED RECEPTIVE FIELDS

- * we can think of this function as a more general version of the “receptive field” of our neuron
- * what is the form of this function?
- * how do we parameterize the stimulus?

GENERALIZED RECEPTIVE FIELDS

- * we can assume that the function f is determined by a set of parameters p
- * these parameters could describe, e.g., the location of the visual receptive field, the orientation selectivity, and so on
- * estimating the parameters from a set of (stimulus, response) pairs is called *system identification*

GENERALIZED RECEPTIVE FIELDS

- * the simplest form that this function could take (probably) is *linear*
- * this means that the response Y is a linear function of the stimulus features
 - * $Y = \beta_1X_1 + \beta_2X_2 + \beta_3X_3 + \dots$
- * we will write this form of model using matrix multiplication notation: $Y=X\beta$

LINEAR MODELS

$$Y = X\beta$$

|
image pixels

X1, Y=0.7



X2, Y=0.3



X3, Y=0.0



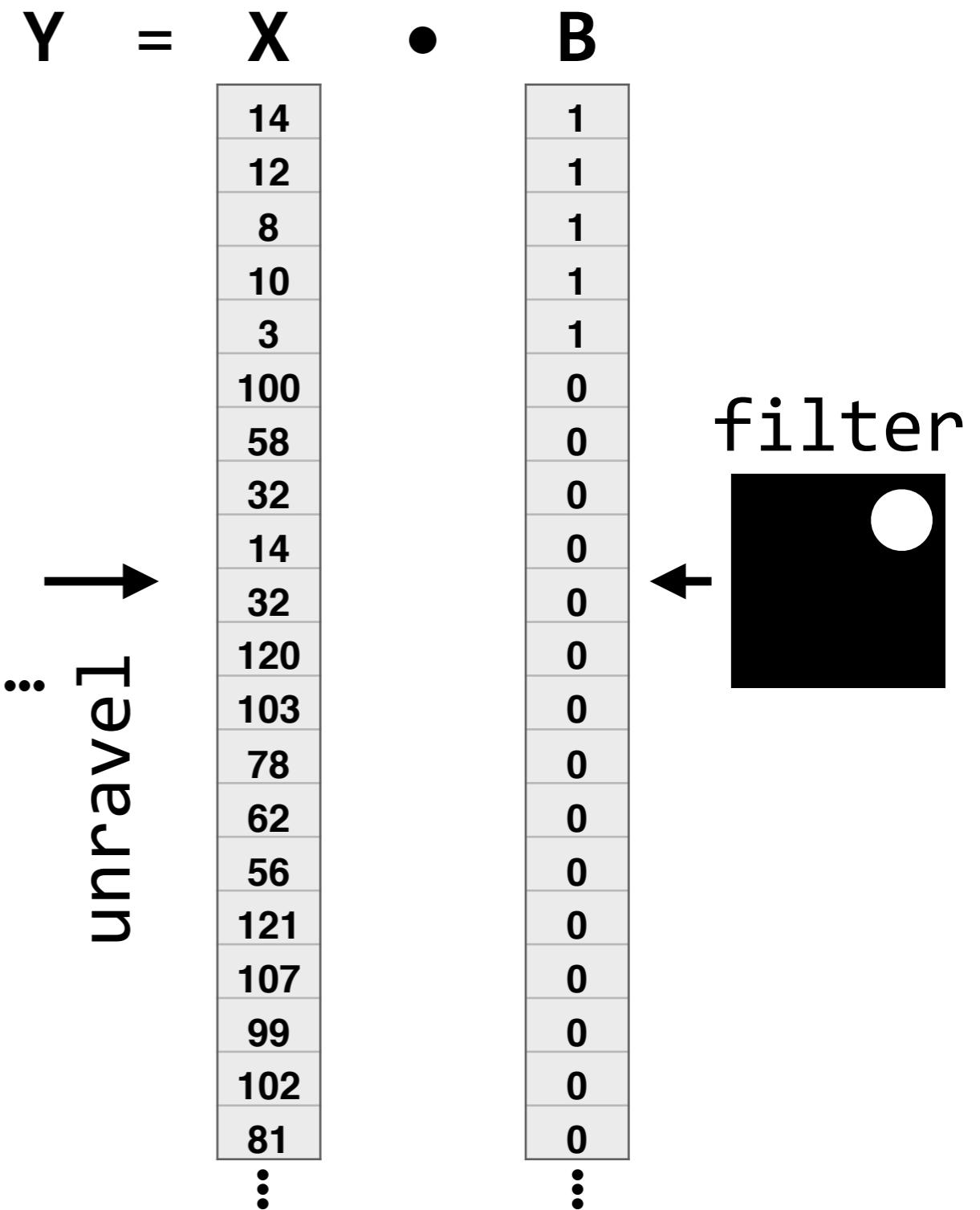
LINEAR MODELS

$$Y = X\beta$$

|
image pixels



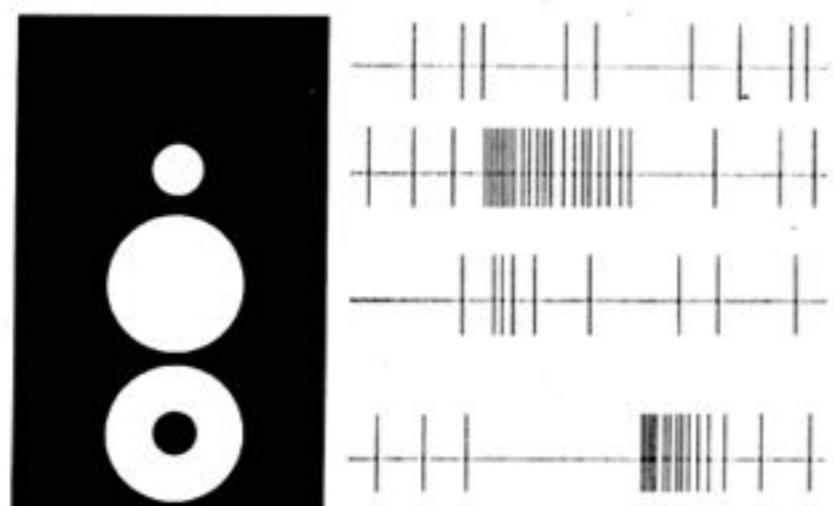
14	100	120	121
12	58	103	107
8	32	78	99
10	14	62	102
3	32	56	81
⋮			



LINEAR MODELS

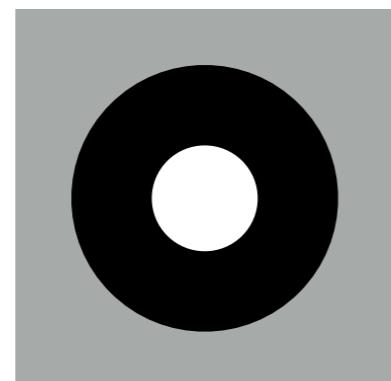
Retinal ganglion cell responses

on-center RGC

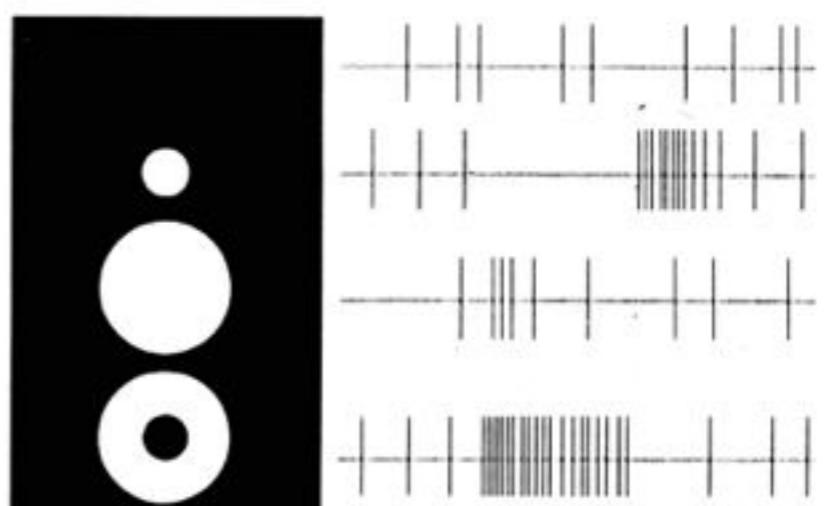


stimulus: on — off

Beta
(on-center)

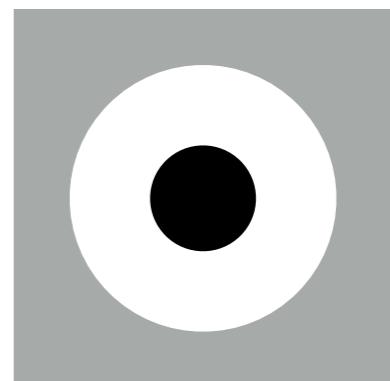


off-center RGC



stimulus: on — off

Beta
(off-center)



FITTING LINEAR MODELS

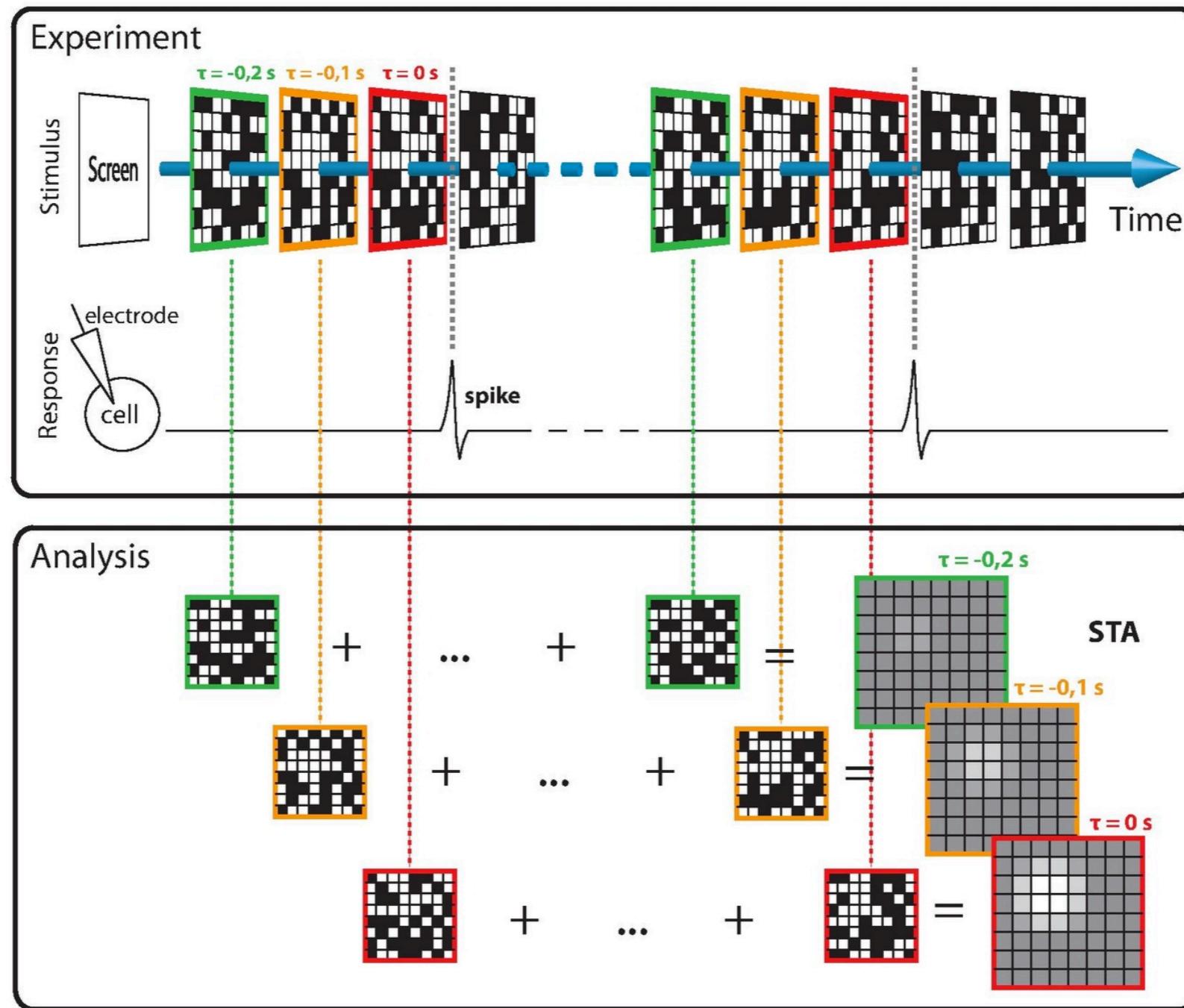
- * suppose we have a bunch of (X, Y) pairs
(i.e. the response of the neuron to many images)
- * how do we figure out what β is?

FITTING LINEAR MODELS

- * one way is **spike-triggered average (STA)**
- * in STA, we assume that each stimulus either elicited a spike or did not (responses are 1 or 0)
- * then we simply take the average of all the stimuli that elicited a spike

FITTING LINEAR MODELS

Spike-triggered average (STA)



FITTING LINEAR MODELS

- * this method is really bad in most cases
- * because it doesn't account for *correlations* among the stimulus channels

FITTING LINEAR MODELS

- * e.g. imagine this is the data:
- * $Y = [1, 1, 0, 1, 0]$
 $X_1 = [1, 1, 0, 1, 0]$
 $X_2 = [1, 0, 0, 1, 0]$
- * STA would say that $\beta_1=1$, and $\beta_2=0.66$
- * but a simpler explanation would be that $\beta_2=0$, since X_1 already explains everything about Y

FITTING LINEAR MODELS

* the solution? regression!

END