

WHAT YOUR FAVORITE MAP PROJECTION SAYS ABOUT YOU

MERCATOR



YOU'RE NOT REALLY INTO MAPS.

ROBINSON



YOU HAVE A COMFORTABLE PAIR OF RUNNING SHOES THAT YOU WEAR EVERYWHERE. YOU LIKE COFFEE AND ENJOY THE BEATLES. YOU THINK THE ROBINSON IS THE BEST-LOOKING PROJECTION, HANDS DOWN.

VAN DER GRINTEN



YOU'RE NOT A COMPLICATED PERSON. YOU LOVE THE MERCATOR PROJECTION; YOU JUST WISH IT WEREN'T SQUARE. THE EARTH'S NOT A SQUARE, IT'S A CIRCLE. YOU LIKE CIRCLES. TODAY IS GONNA BE A GOOD DAY!

Dymaxion



YOU LIKE ISAAC ASIMOV, XML, AND SHOES WITH TOES. YOU THINK THE SEGWAY GOT A BAD RAP. YOU OWN 3D GOGGLES, WHICH YOU USE TO VIEW ROTATING MODELS OF BETTER 3D GOGGLES. YOU TYPE IN DVORAK.

WINKEL-TRIPEL



NATIONAL GEOGRAPHIC ADOPTED THE WINKEL-TRIPEL IN 1998, BUT YOU'VE BEEN A WT FAN SINCE LONG BEFORE "NAT GEO" SHOWED UP. YOU'RE WORRIED IT'S GETTING PLAYED OUT, AND ARE THINKING OF SWITCHING TO THE KAVRAYSKY. YOU ONCE LEFT A PARTY IN DISGUST WHEN A GUEST SHOWED UP WEARING SHOES WITH TOES. YOUR FAVORITE MUSICAL GENRE IS "POST-".

HOBÓ-DYER



YOU WANT TO AVOID CULTURAL IMPERIALISM, BUT YOU'VE HEARD BAD THINGS ABOUT GALL-PETERS. YOU'RE CONFLICT-AVERSE AND BUY ORGANIC. YOU USE A RECENTLY-INVENTED SET OF GENDER-NEUTRAL PRONOUNS AND THINK THAT WHAT THE WORLD NEEDS IS A REVOLUTION IN CONSCIOUSNESS.

GOODE HOMOLOSINE



THEY SAY MAPPING THE EARTH ON A 2D SURFACE IS LIKE FLATTENING AN ORANGE PEEL, WHICH SEEMS EASY ENOUGH TO YOU. YOU LIKE EASY SOLUTIONS. YOU THINK WE WOULDN'T HAVE SO MANY PROBLEMS IF WE'D JUST ELECT NORMAL PEOPLE TO CONGRESS INSTEAD OF POLITICIANS. YOU THINK AIRLINES SHOULD JUST BUY FOOD FROM THE RESTAURANTS NEAR THE GATES AND SERVE THAT ON BOARD. YOU CHANGE YOUR CAR'S OIL, BUT SECRETLY WONDER IF YOU REALLY NEED TO.

PLATE CARRÉE
(EQUIRECTANGULAR)



YOU THINK THIS ONE IS FINE. YOU LIKE HOW X AND Y MAP TO LATITUDE AND LONGITUDE. THE OTHER PROJECTIONS OVERCOMPLICATE THINGS. YOU WANT ME TO STOP ASKING ABOUT MAPS SO YOU CAN ENJOY DINNER.

A GLOBE!



YES, YOU'RE VERY CLEVER.

PEIRCE QUINCUNCIAL



YOU THINK THAT WHEN WE LOOK AT A MAP, WHAT WE REALLY SEE IS OURSELVES. AFTER YOU FIRST SAW INCEPTION, YOU SAT SILENT IN THE THEATER FOR SIX HOURS. IT FREAKS YOU OUT TO REALIZE THAT EVERYONE AROUND YOU HAS A SKELETON INSIDE THEM. YOU HAVE REALLY LOOKED AT YOUR HANDS.

WATERMAN BUTTERFLY



REALLY? YOU KNOW THE WATERMAN? HAVE YOU SEEN THE 1909 CAHILL MAP IT'S BASED— ... YOU HAVE A FRAMED REPRODUCTION AT HOME?! WHOA. ... LISTEN, FORGET THESE QUESTIONS. ARE YOU DOING ANYTHING TONIGHT?

GALL-PETERS



I HATE YOU.

Questions from session 1

Q1.1 (Polina)

The replication crisis in psychology raised questions about whether many "established" findings are real. What mechanistic role, if any, could computational modelling play in addressing this crisis?

Q1.2 (Juan)

We have presented models as tools for understanding. However, models are increasingly also tools for prediction and control, specially in AI. When does a model's usefulness for control become ethically problematic, even if it aids understanding? Should neuroscientists consider how their models might be used?

Q1.3 (Shiru)

Foucault argued that scientific categories do not merely describe reality but actively shape it through institutional power. If this is true, what responsibility do scientists bear for the categories they create and propagate? Can a model be simultaneously useful, socially constructed, and in some sense true?

Q1.4 (Miguel)

For Judith Butler, gender is not discovered but performed into existence: a model that consolidates itself through use. Do you agree with Butler? Consider Box's aphorism. In what sense is gender a "wrong" model? In what sense is it a "useful" model? And useful for whom? Apply the same reasoning to sexuality.

Q1.6 (Izei)

The same data often support multiple, incompatible interpretations. This is the underdetermination problem in philosophy of science. If data can't uniquely determine theory, what does? Aesthetics? Social consensus? Pragmatic success? What implications does this have for neuroscience?

Q1.9 (Jacy)

Consider Chomsky's generative grammar. Where does it fit within Marr's three levels? Where does it not? Has the separation of levels been productive for linguistics? What are the costs and benefits of developing theory largely independently of neural data?

Q1.11 (Francesca)

V1 neurons respond to oriented edges. How do we get from this to recognising your grandmother's face? Sketch out what you think the computational problem is at each subsequent stage. Where do you think the biggest gaps in our understanding are?

Representation of information

Computational Neuroscience - Lecture 2

Alejandro Tabas

Computational models in Neuroscience research

1. What is a representation?
2. Single-unit codes and receptive fields
3. Measuring receptive fields
4. Population codes
5. Transformation of information
6. Conclusions
7. Mind reading

Computational models in Neuroscience research

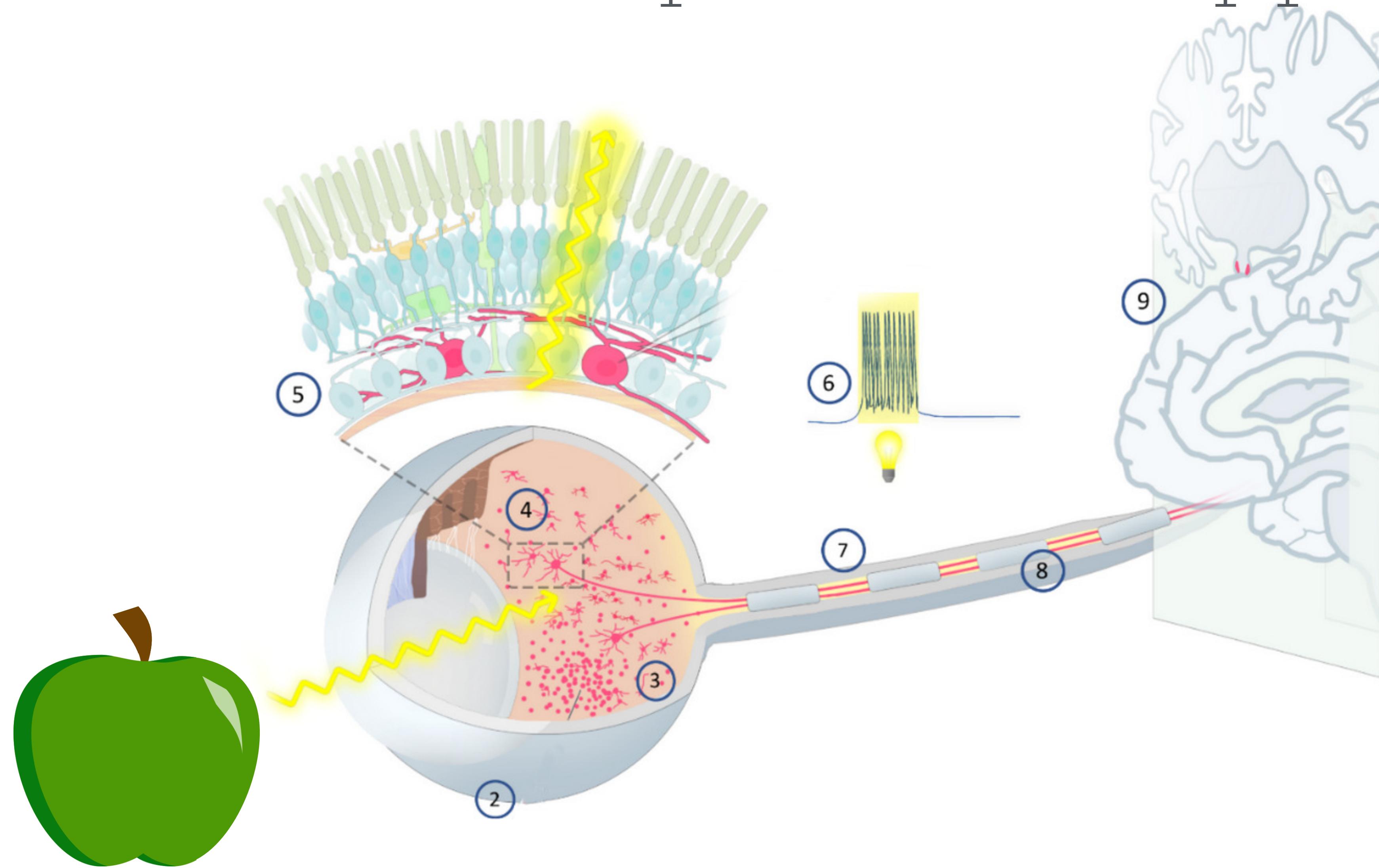
- 1. What is a representation?**
2. Single-unit codes and receptive fields
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What is a representation?

Lets forget about the brain for a minute...

- What does it mean to represent something? What is a representation?
- Can you think of any example of a representation?
- What makes a good/bad representation?

How does the brain represent an apple?



How does the brain represent information?

Think of some examples of information that you are sure the brain represents

- Does everything we know and think have to be represented in the brain?

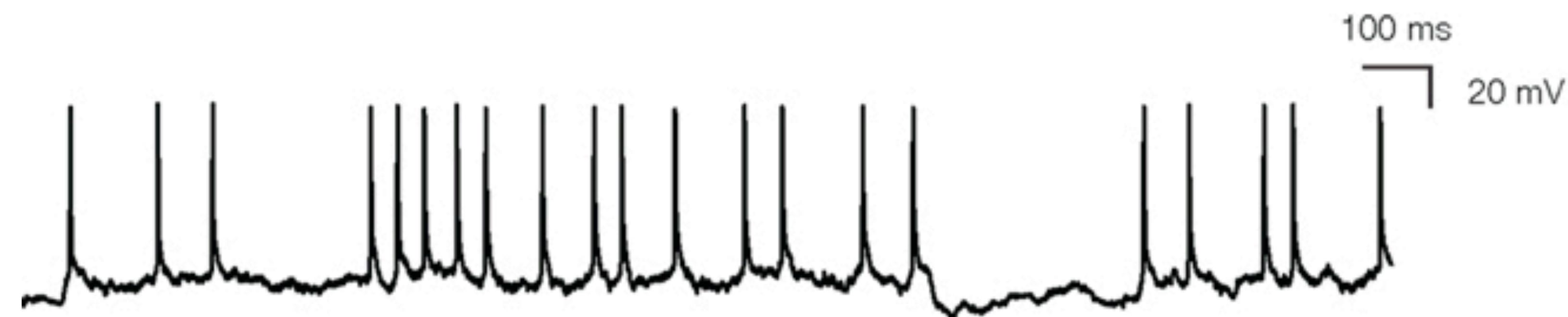
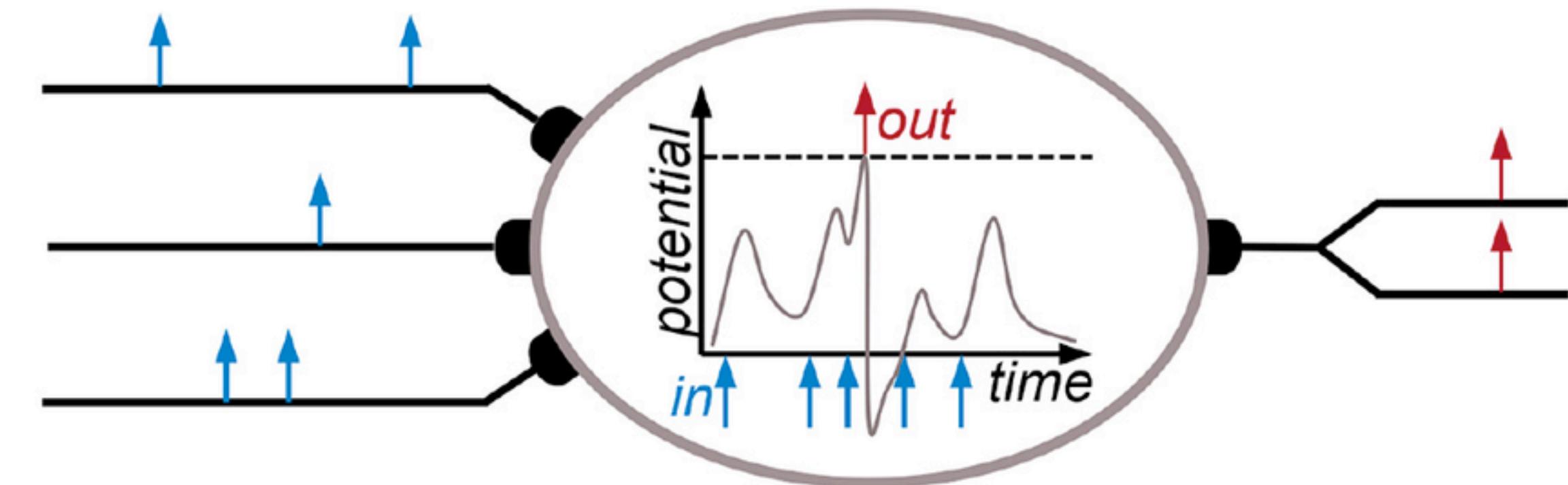
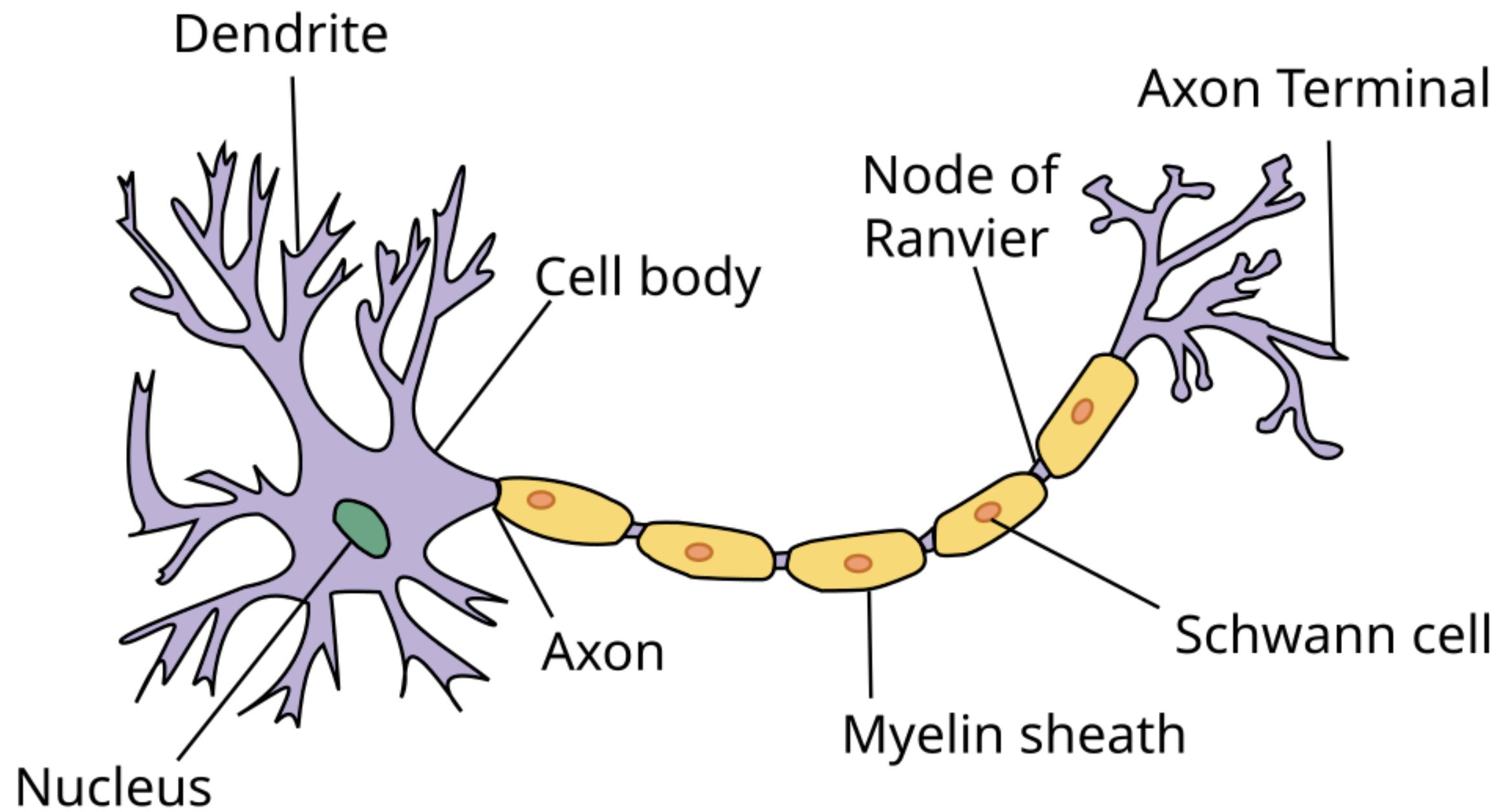
What specific resources does the brain have to represent information?

- Do all neuronal spikes encode a representation of information?
- Do all synaptic connections encode a representation of information?

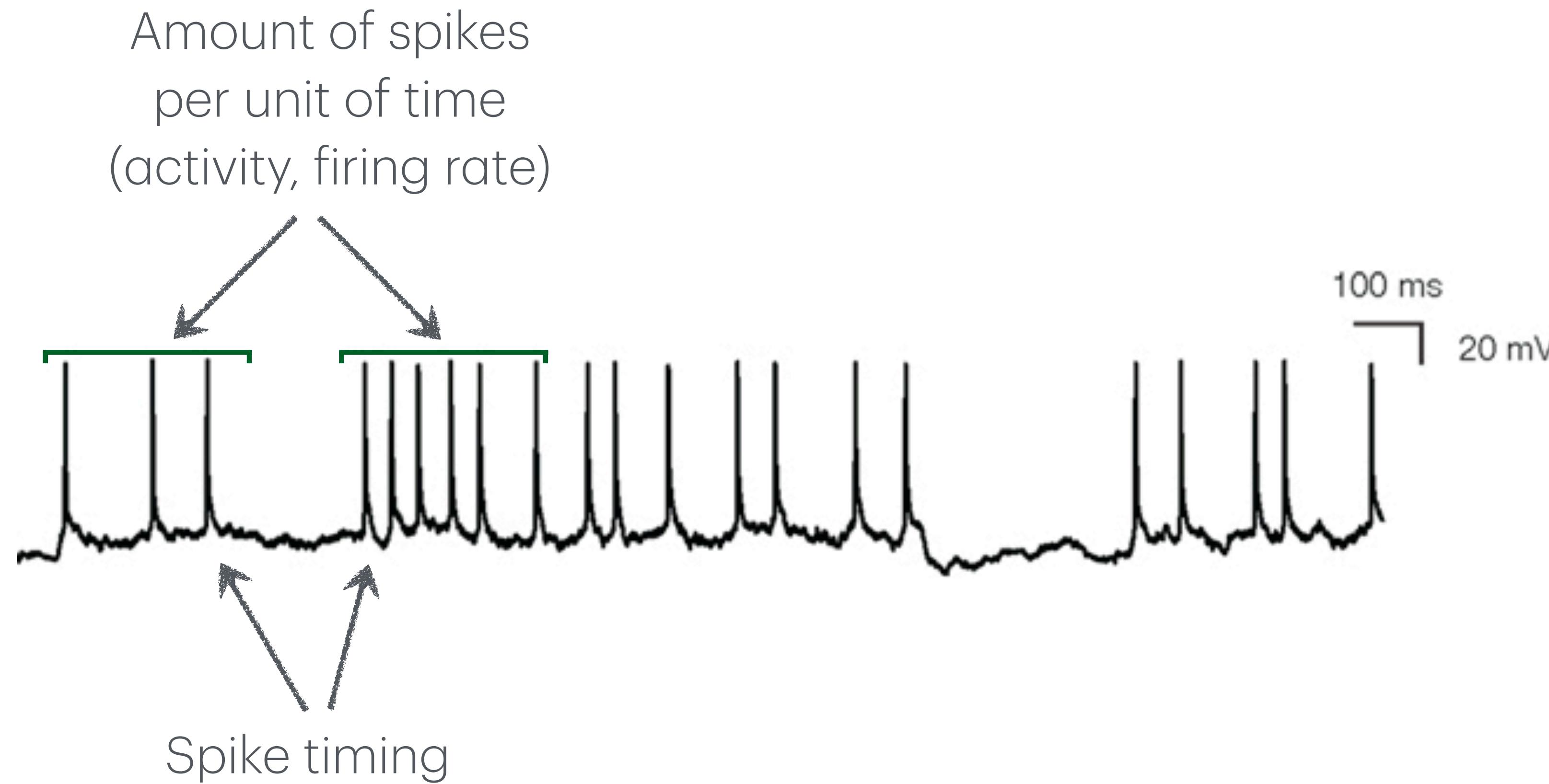
Computational models in Neuroscience research

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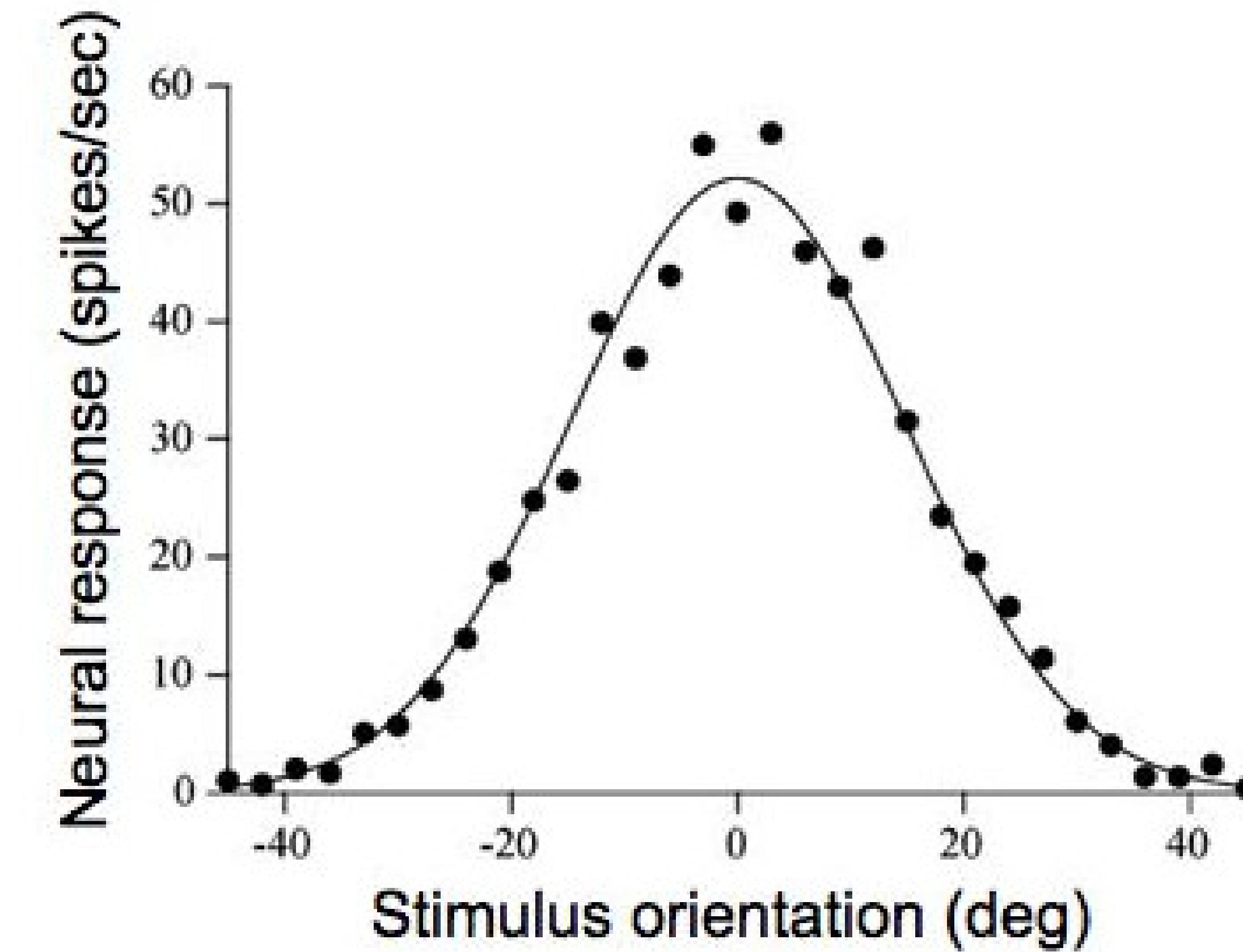
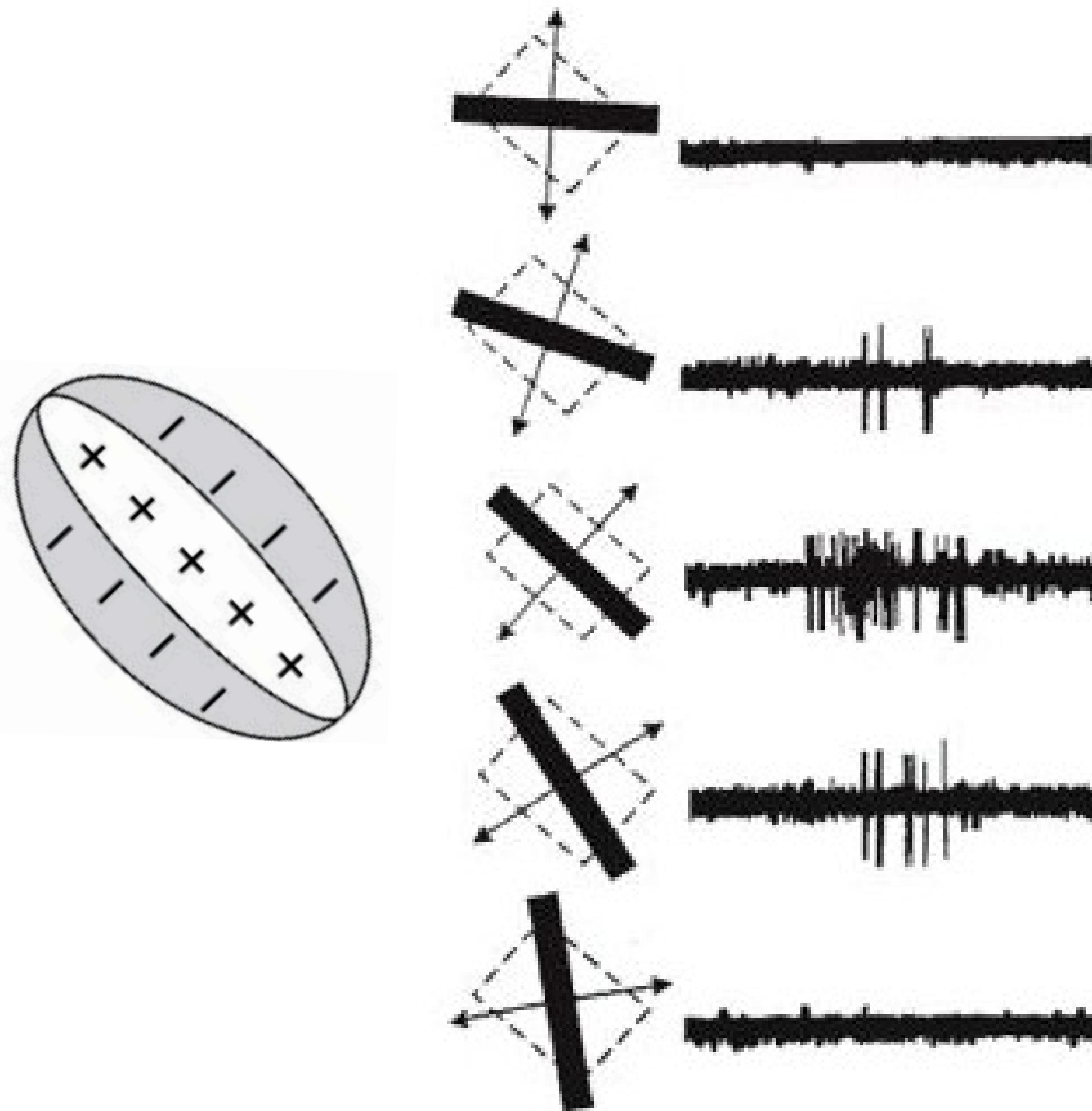
Neuronal spikes



How can single neurons encode information?

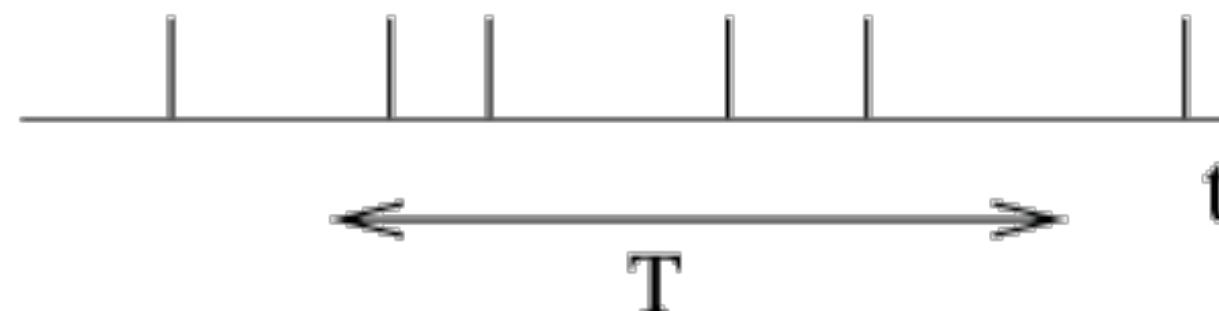


Example: V1 receptive fields



Three ways to define firing rate

Rate = average over time
(single neuron, single run)



Spike count

$$v = \frac{n_{sp}}{T}$$

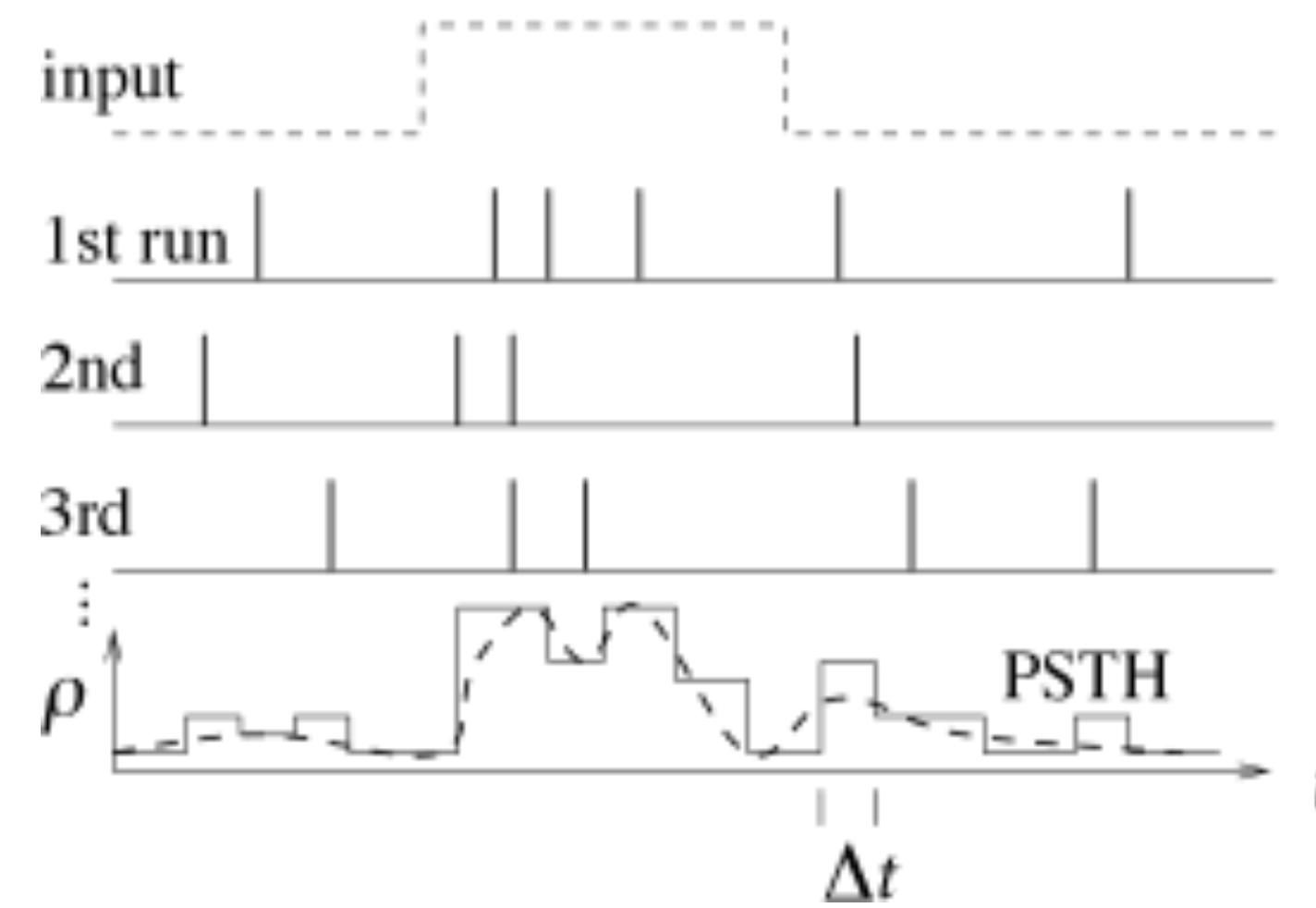
rate = average over pool of equivalent neurons
(several neurons, single run)



activity

$$A = \frac{1}{\Delta t} \frac{n_{act}(t; t+\Delta t)}{N}$$

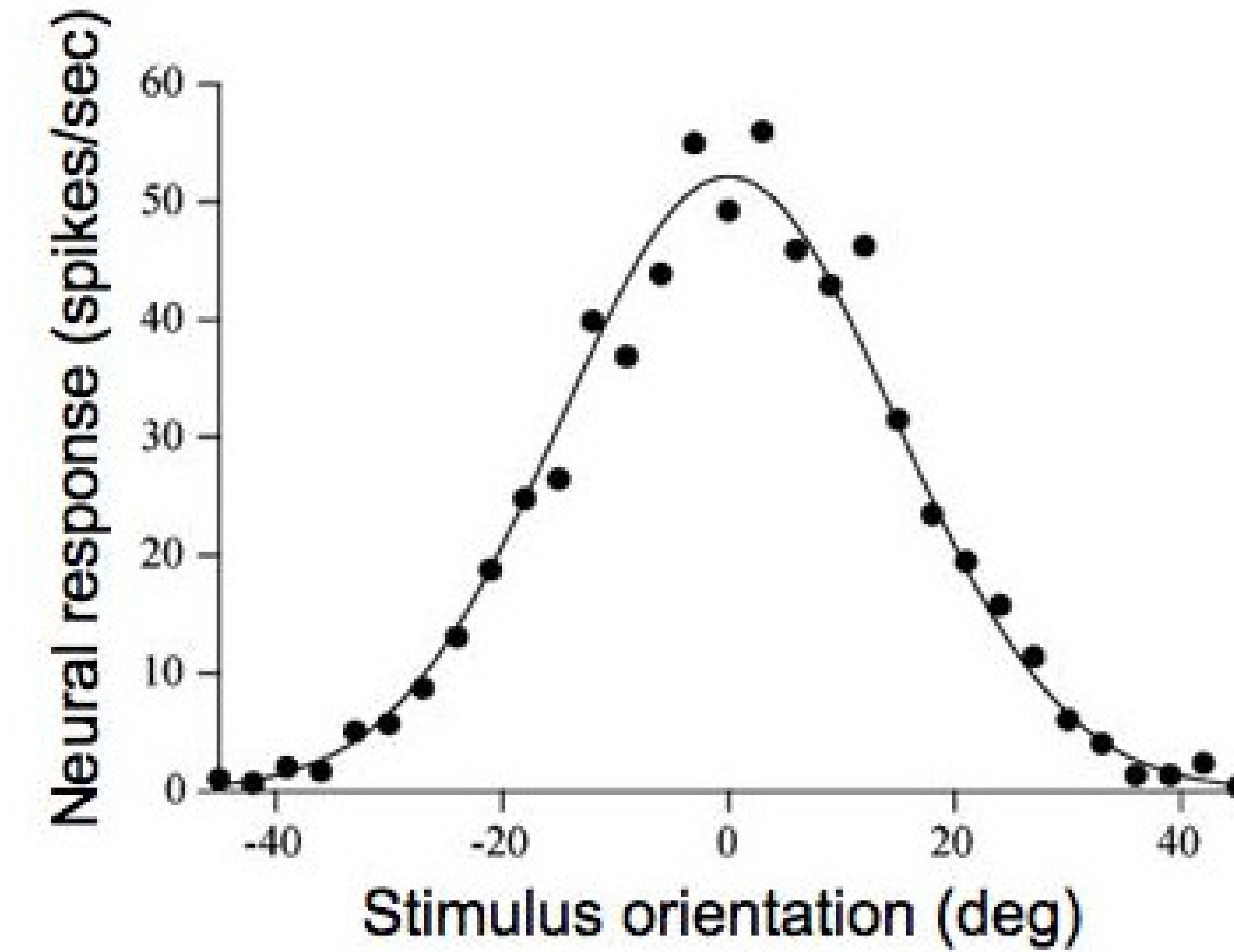
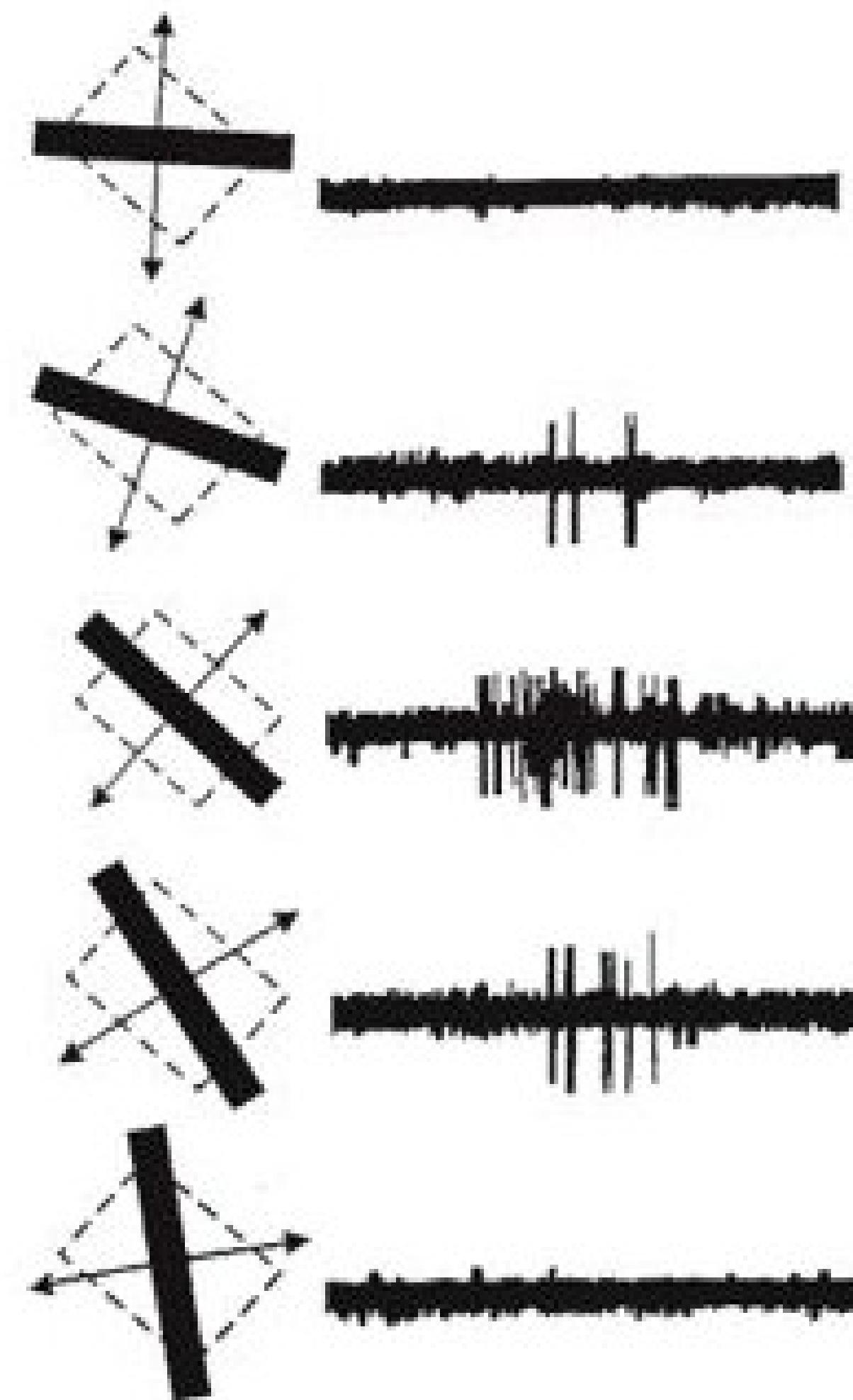
rate = average over several runs
(single neuron, repeated runs)



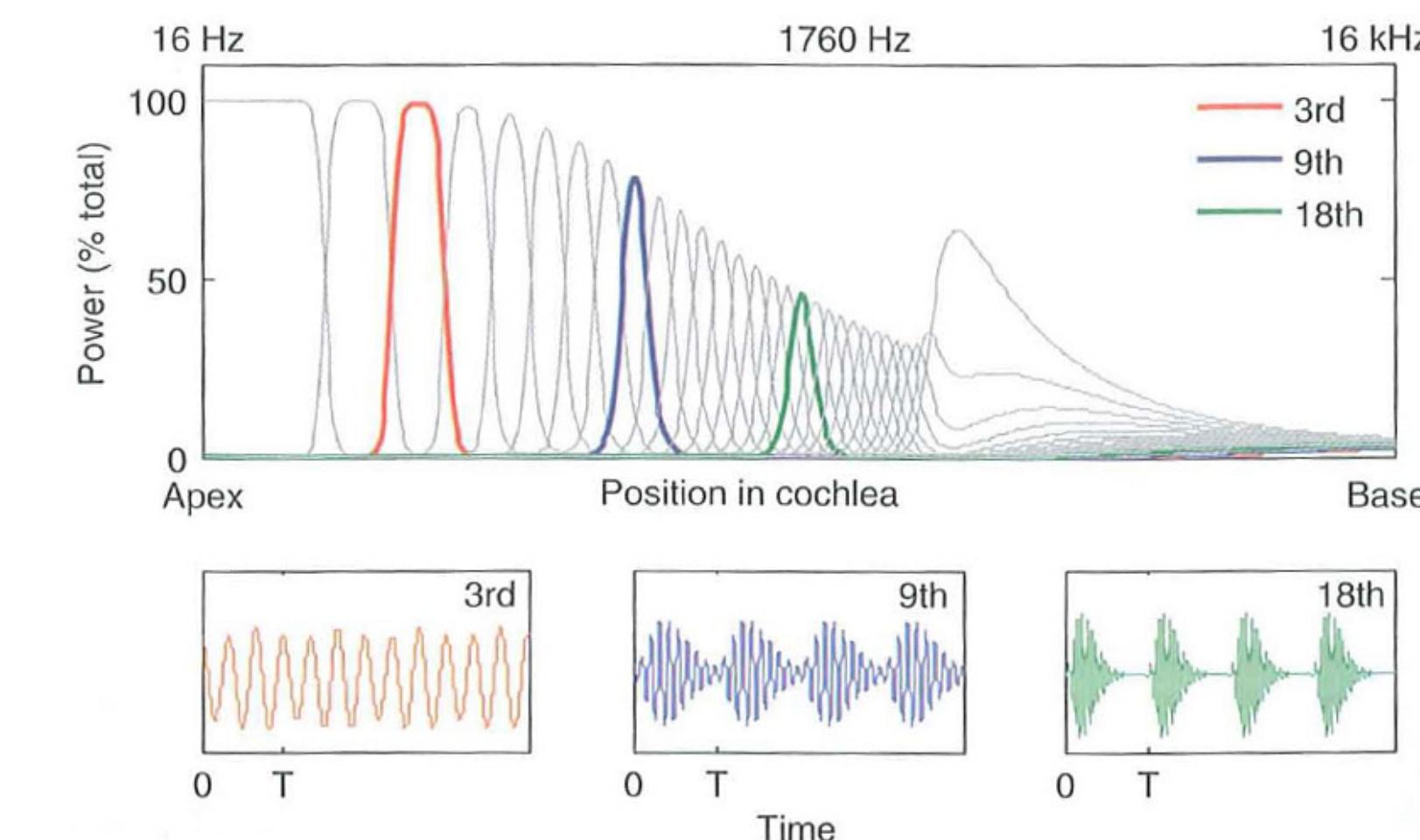
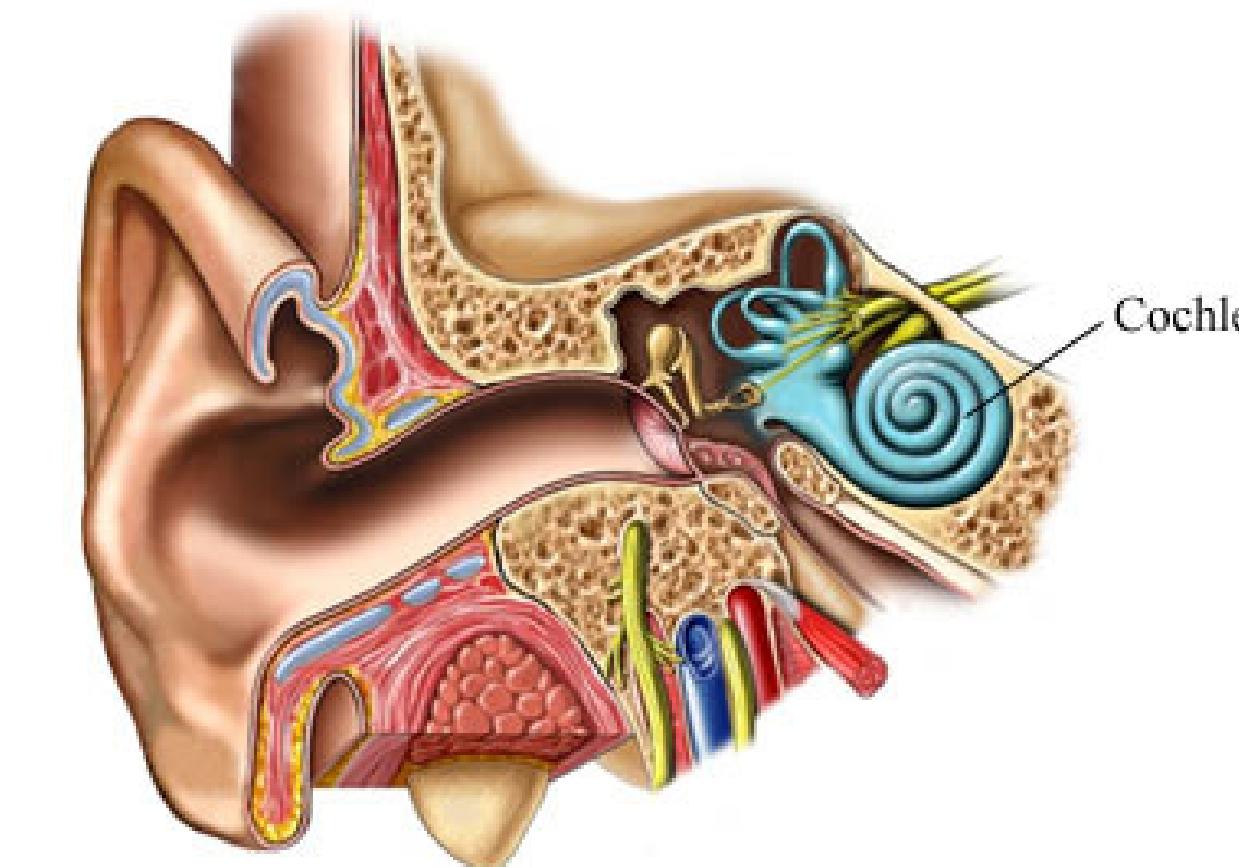
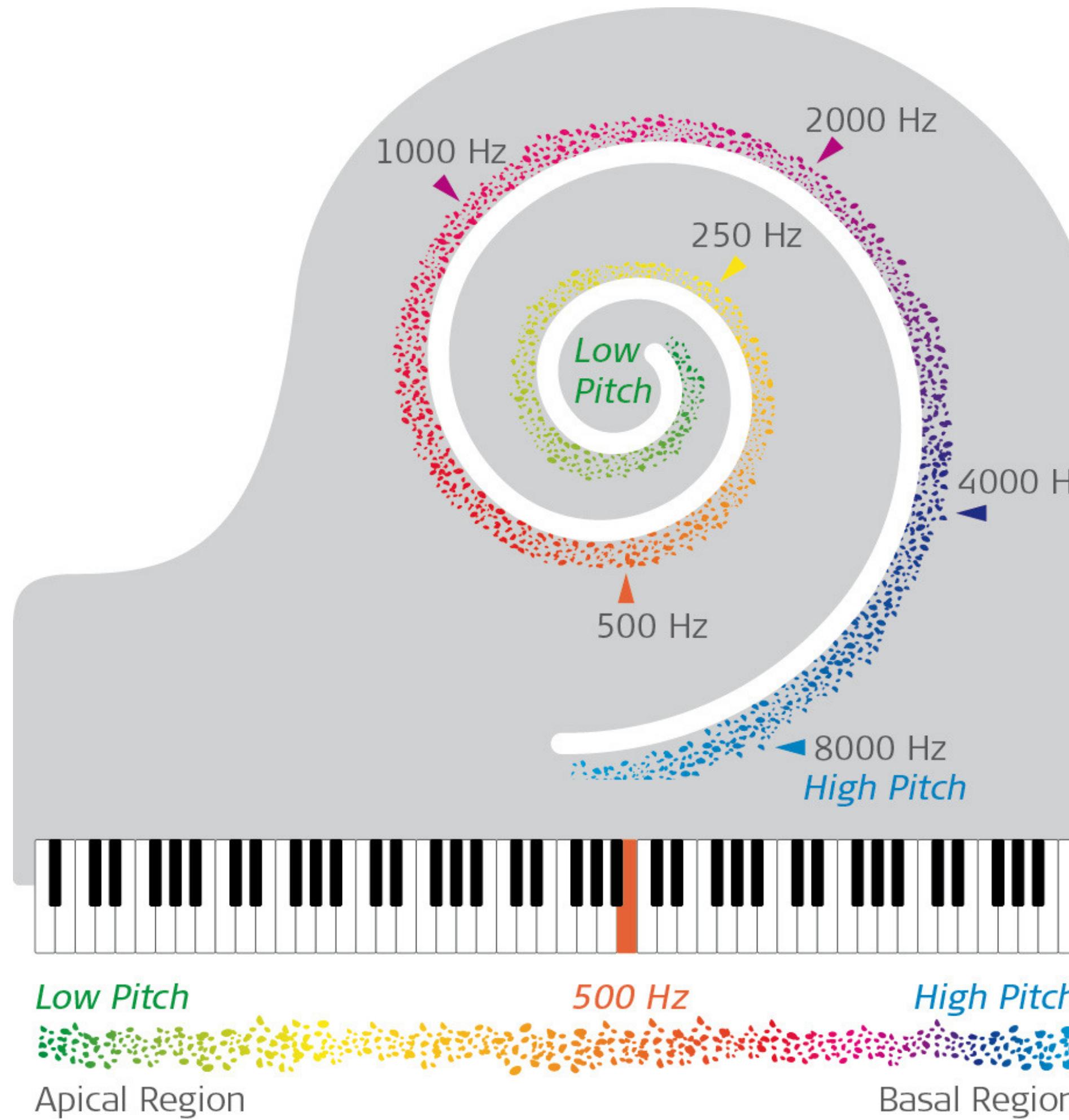
spike density
in PSTH

$$\rho = \frac{1}{\Delta t} \frac{1}{K} n_K(t; t+\Delta t)$$

Example: V1 receptive fields

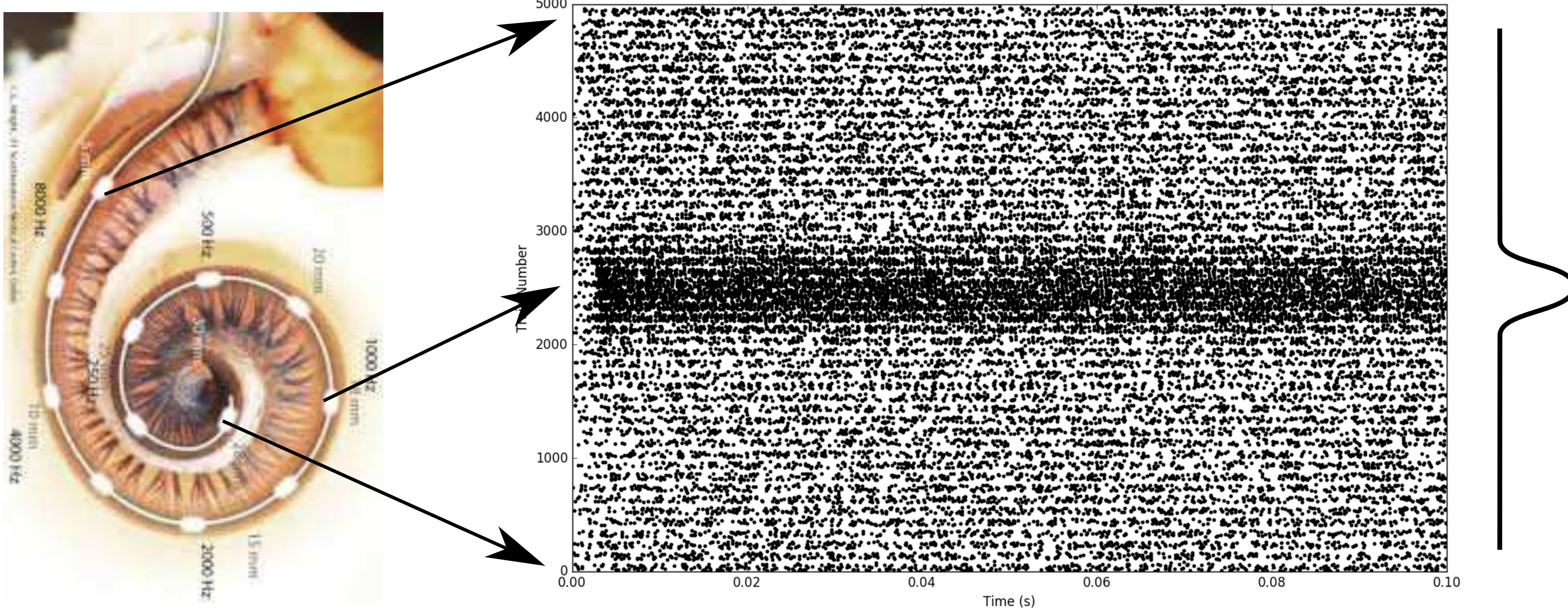


Example: auditory receptive fields



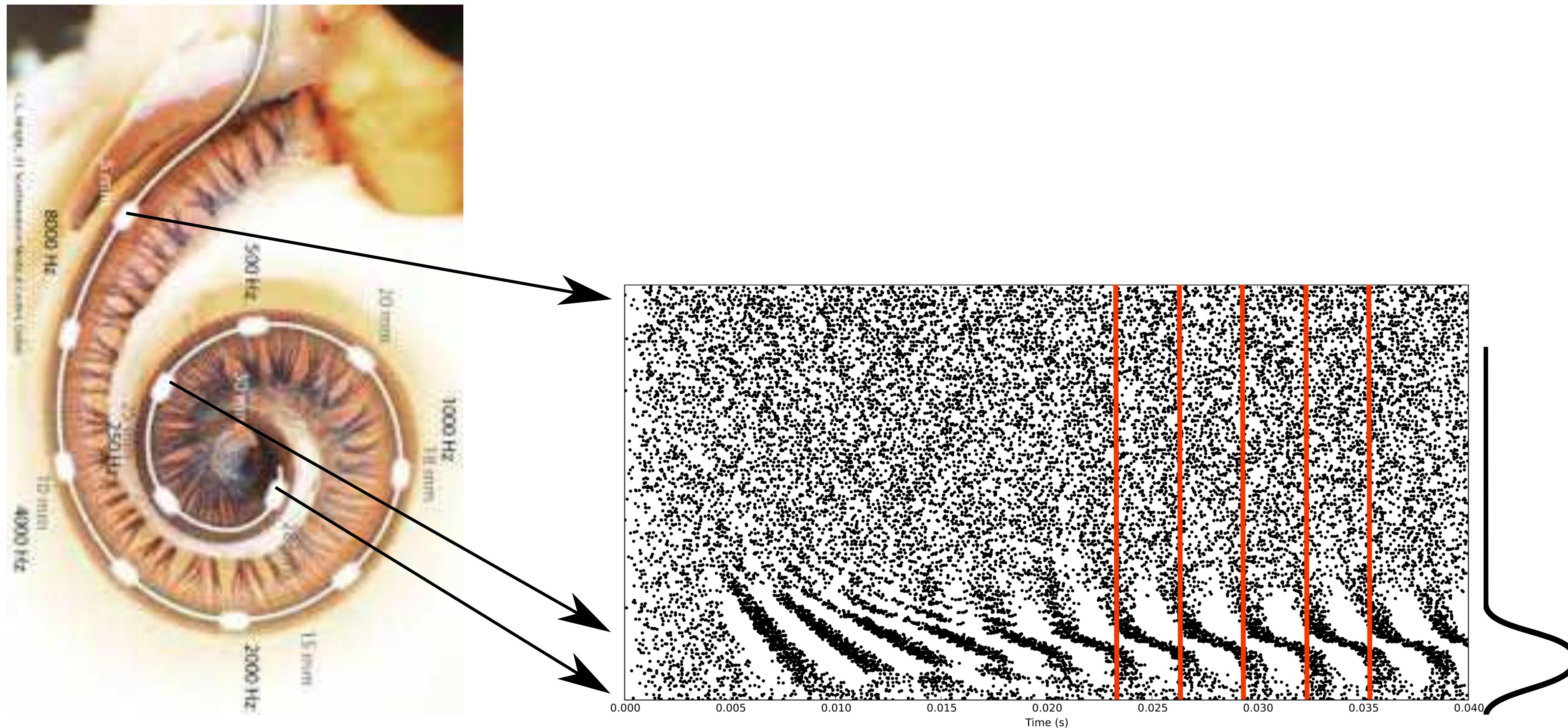
Example: auditory receptive fields

raster plot for an iterated rippled noise with $f_0 = 1000$ Hz

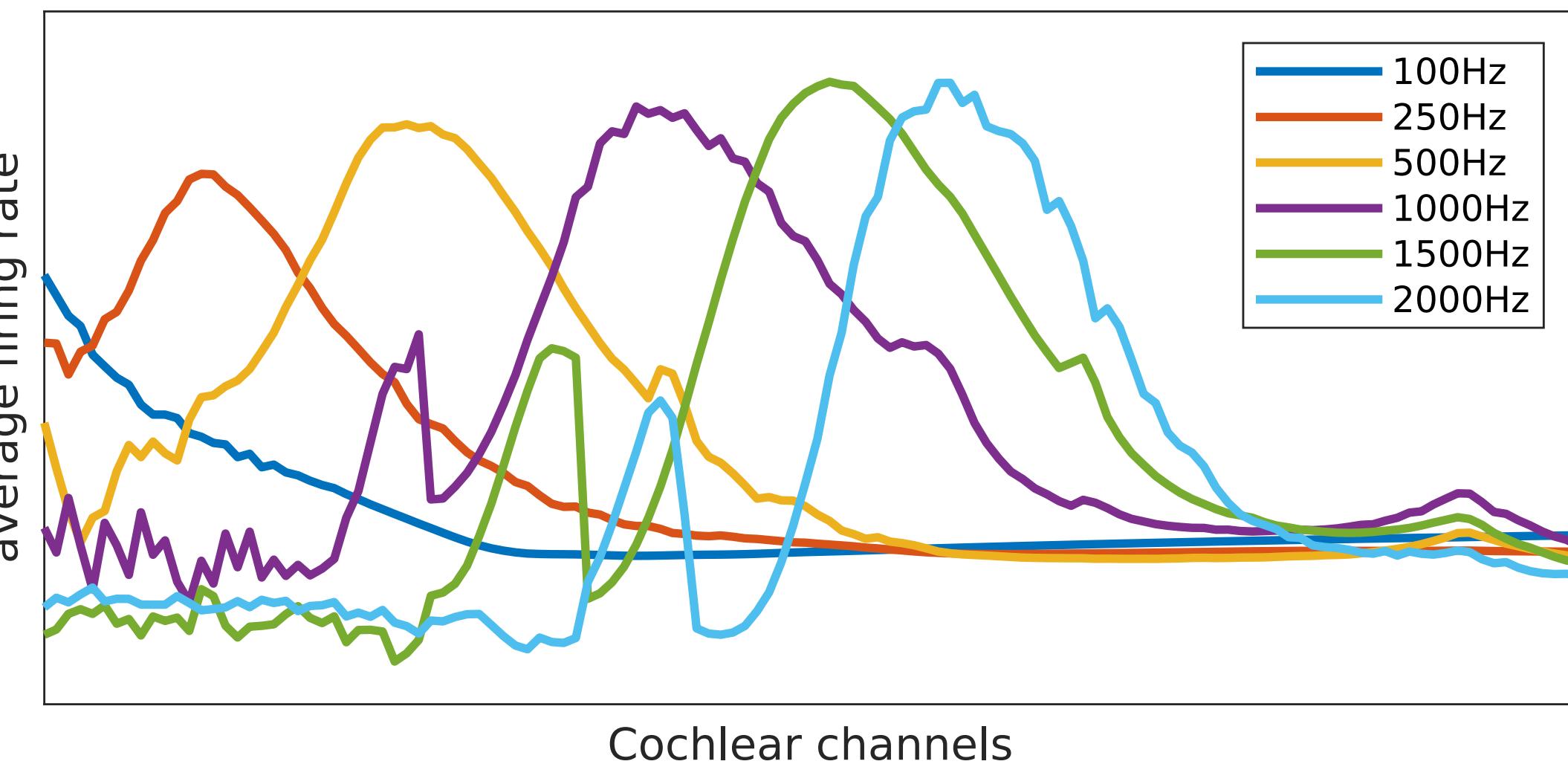
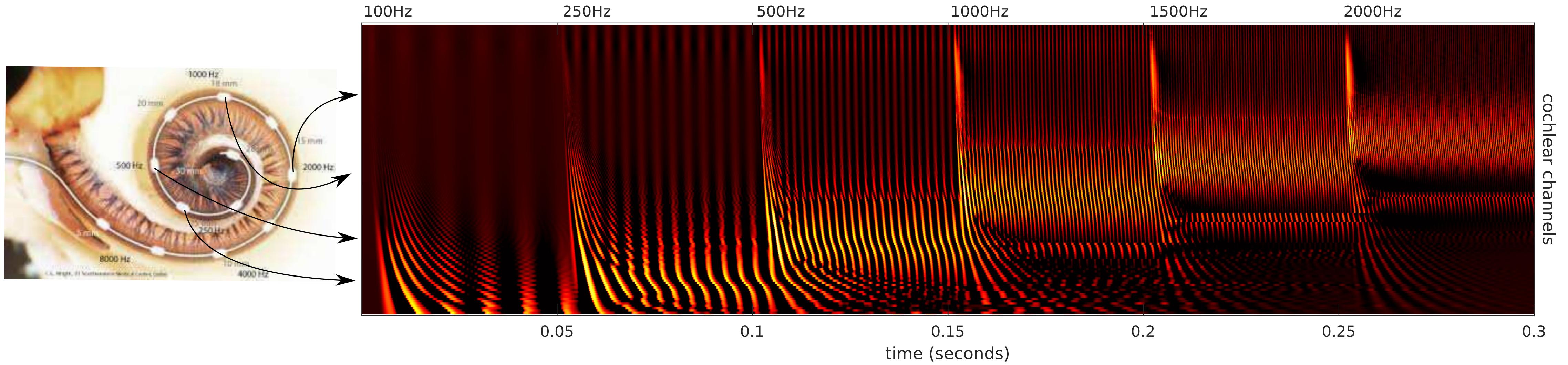


Example of a time code: phase locking

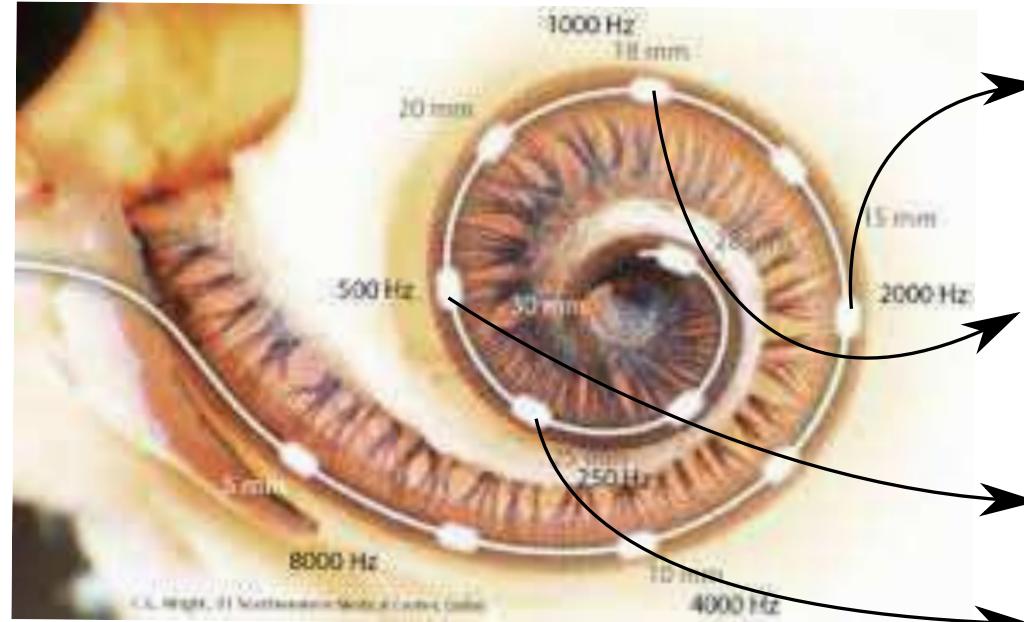
raster plot for a pure tone with $f = 200$ Hz



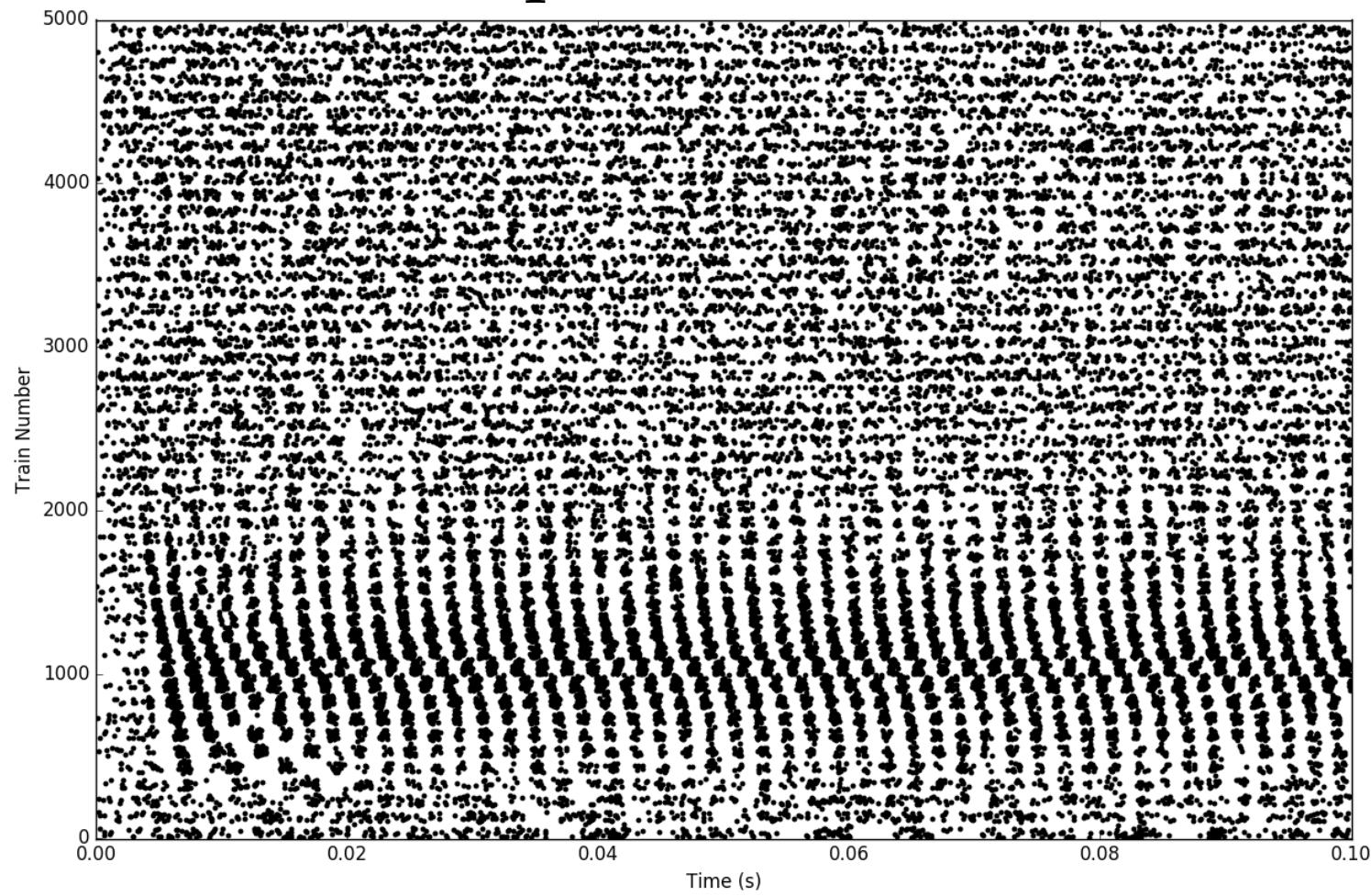
Auditory receptive fields: rate and time codes



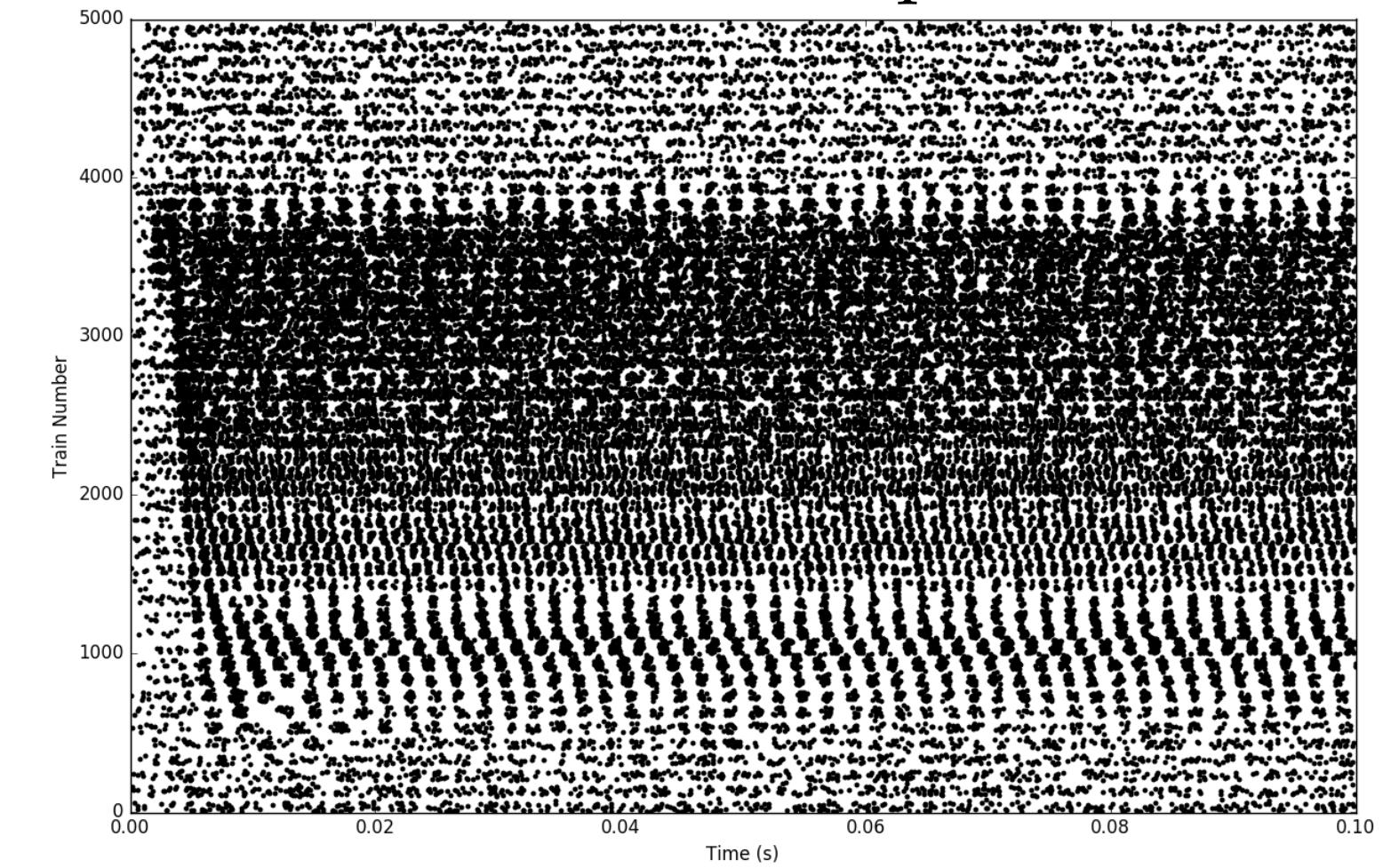
Auditory receptive fields: rate and time codes



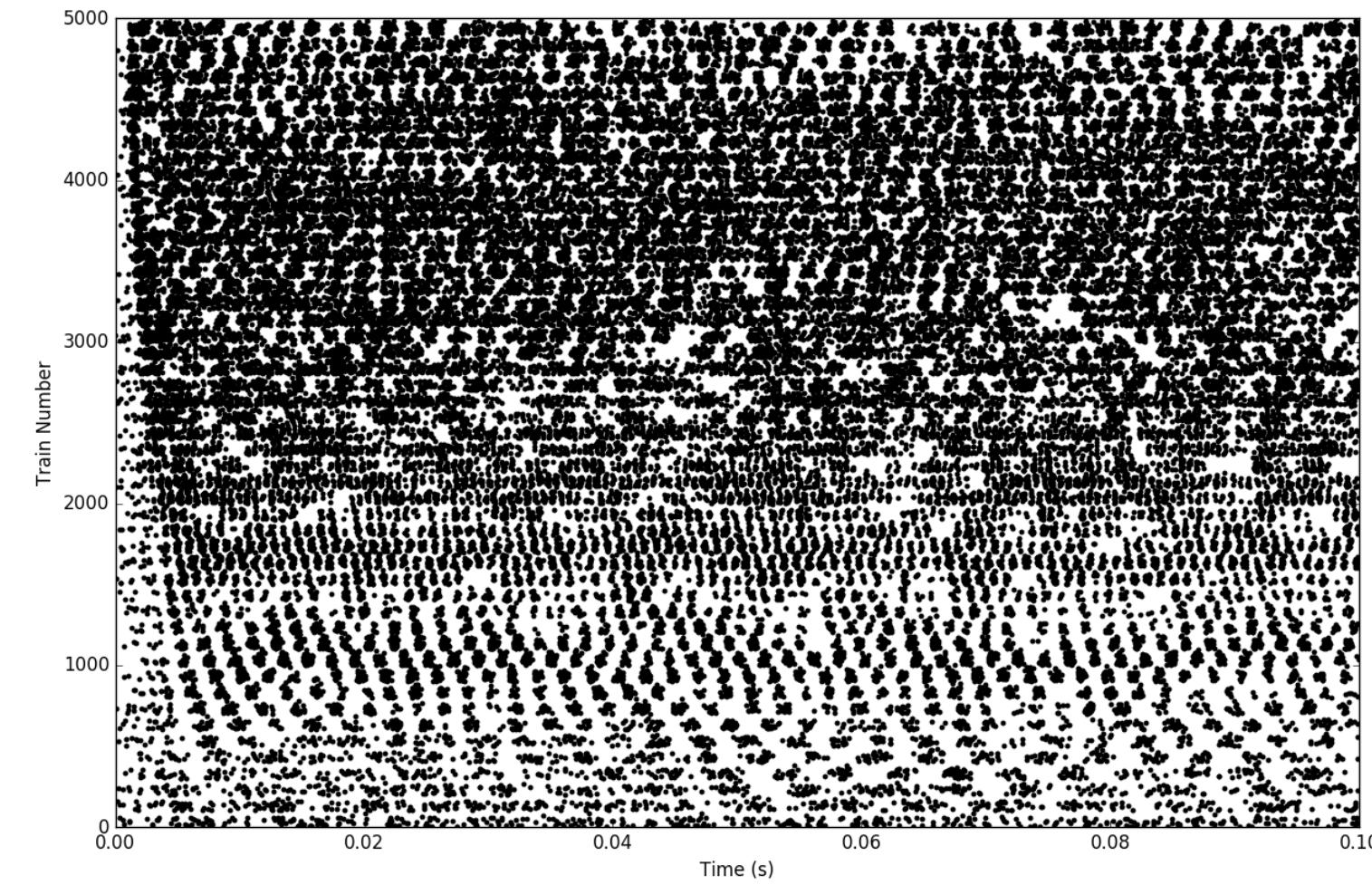
pure tone



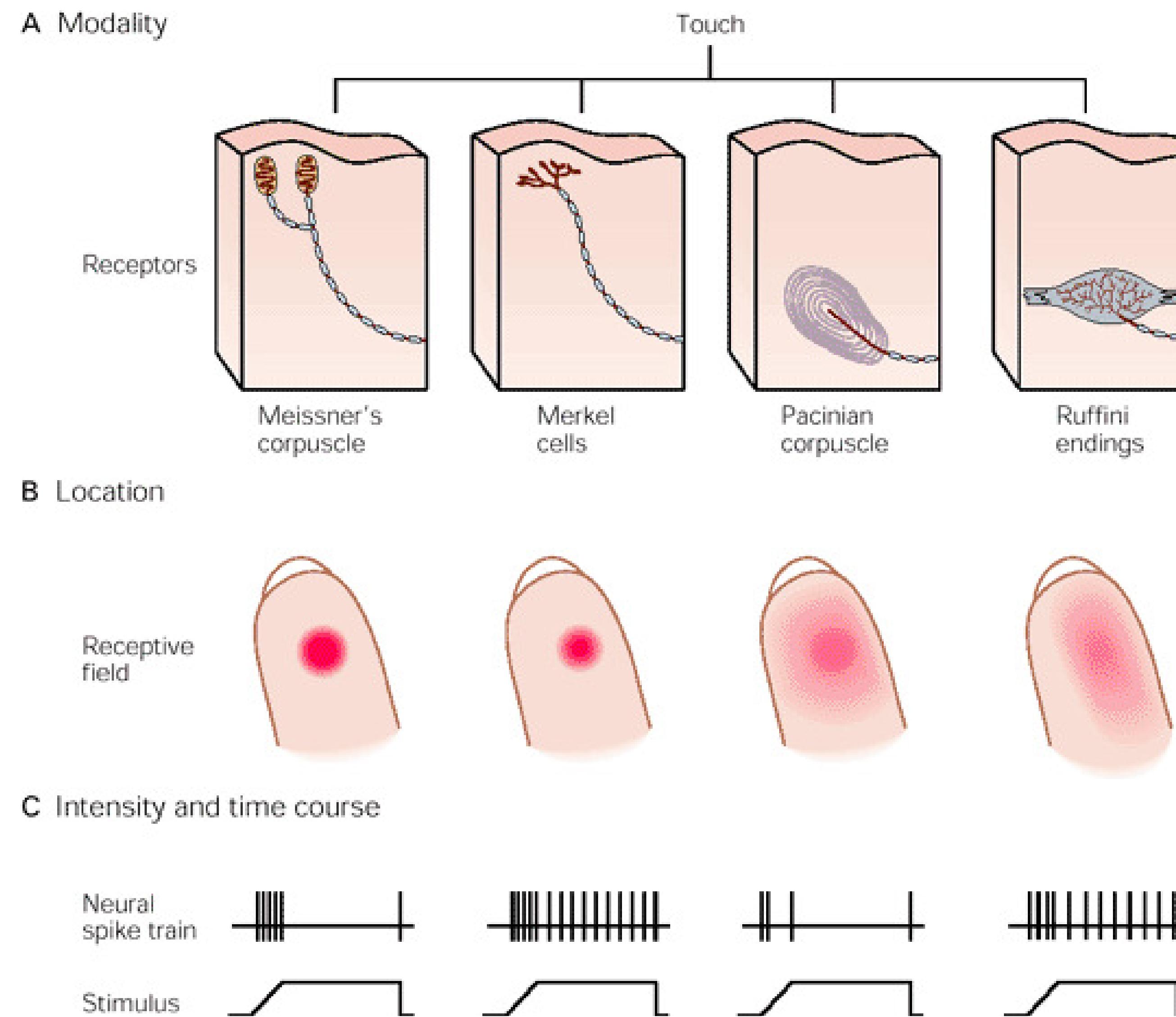
harmonic complex



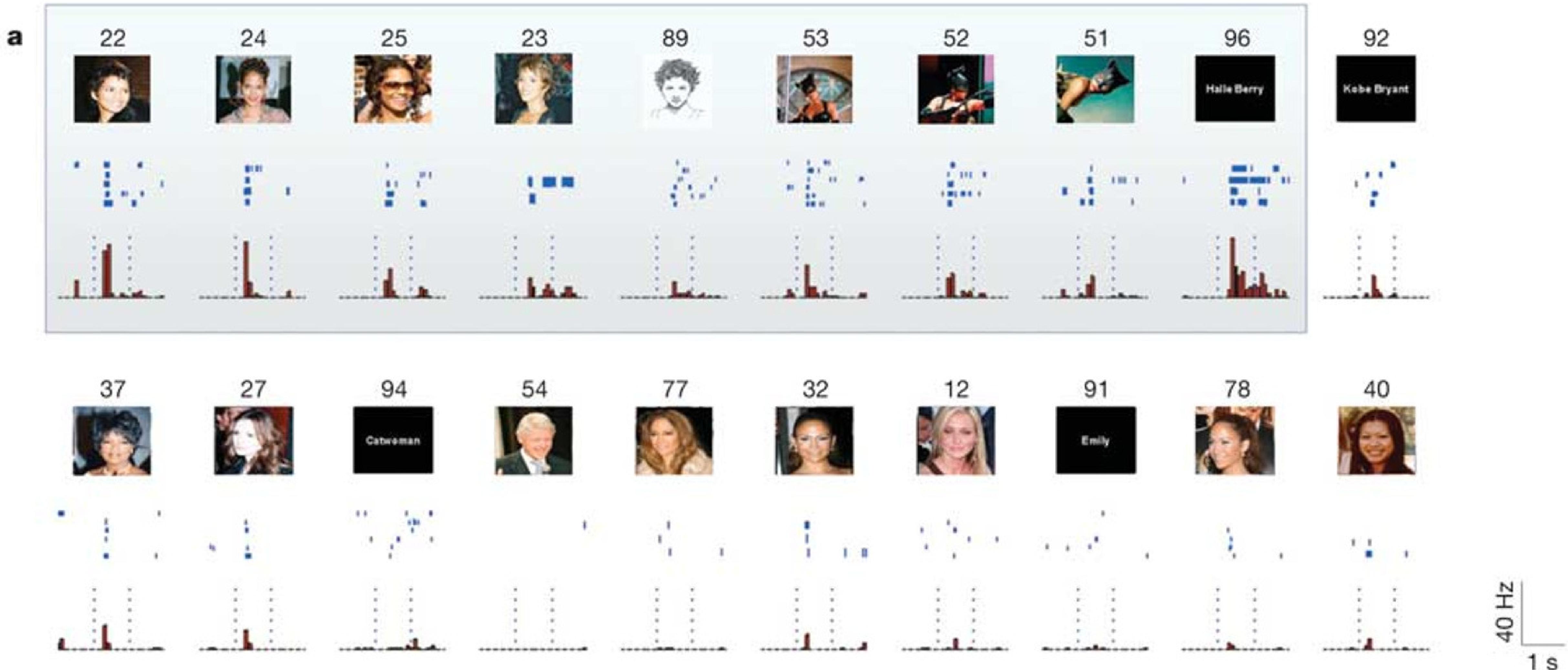
iterated rippled noise



Example: Somatosensory receptive fields

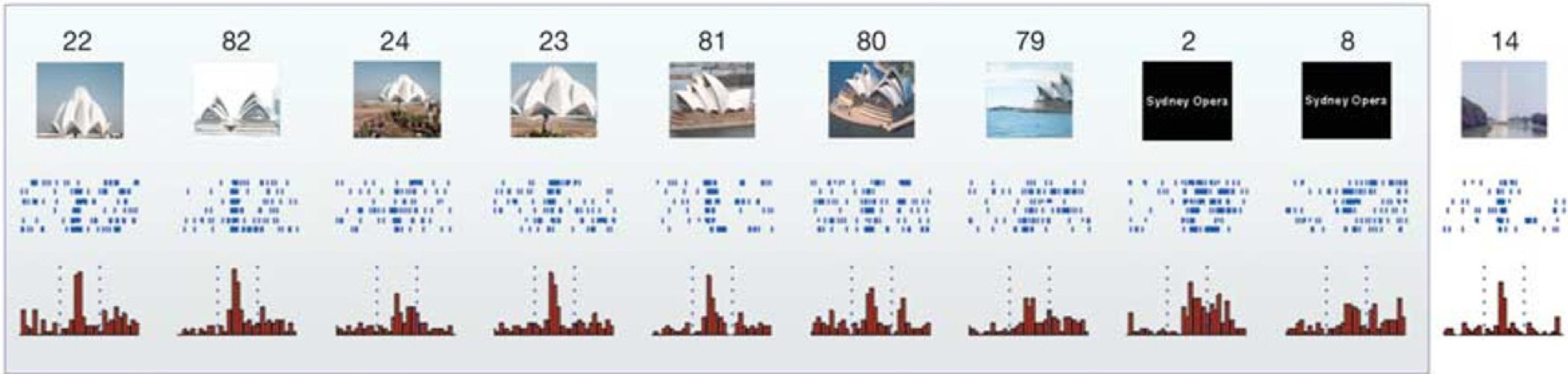


Example: Abstract receptive fields



Example: Abstract receptive fields

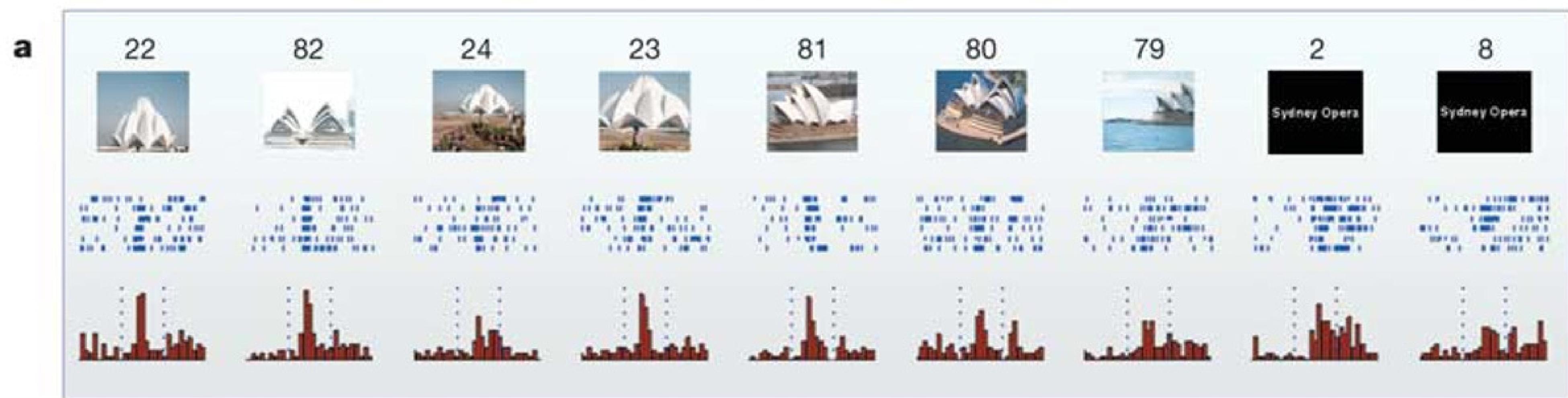
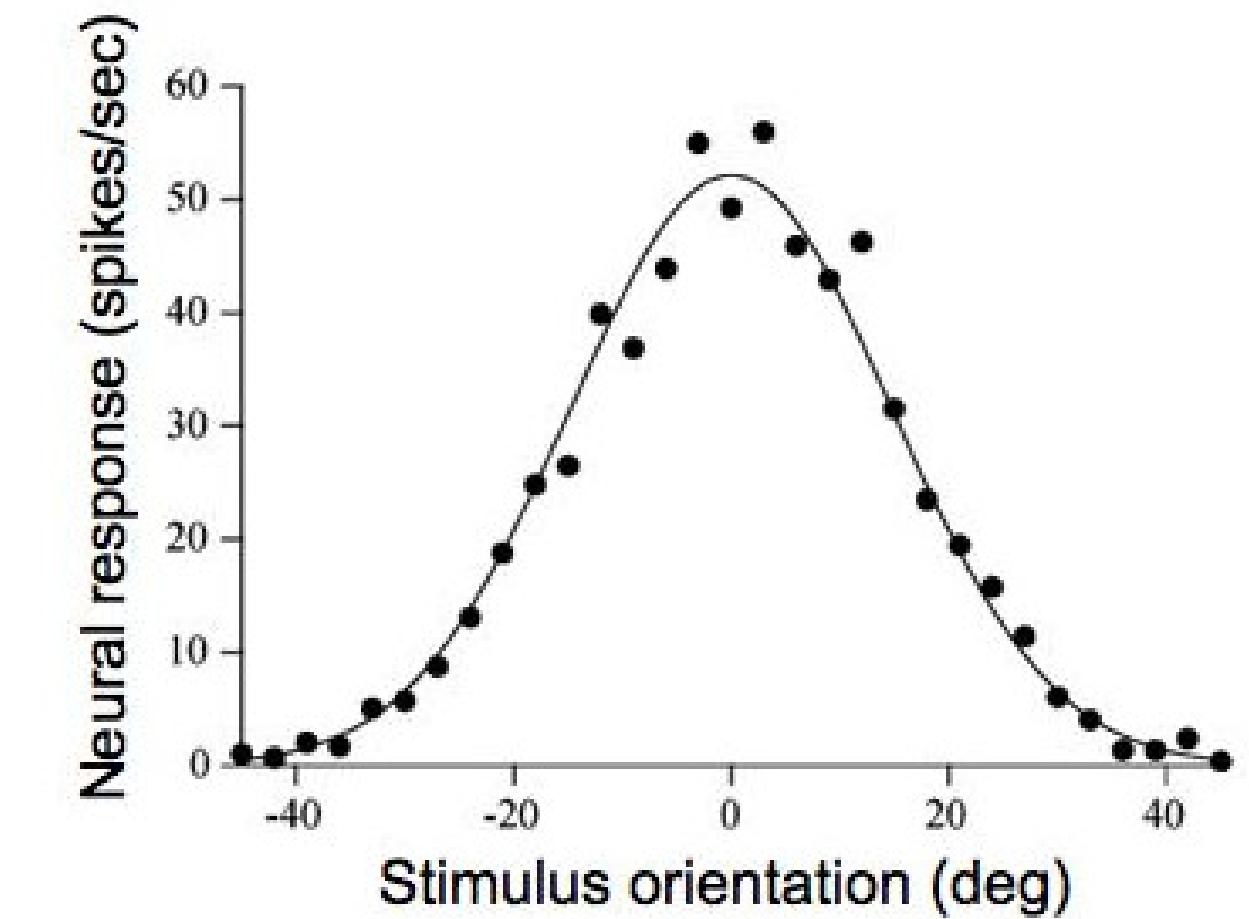
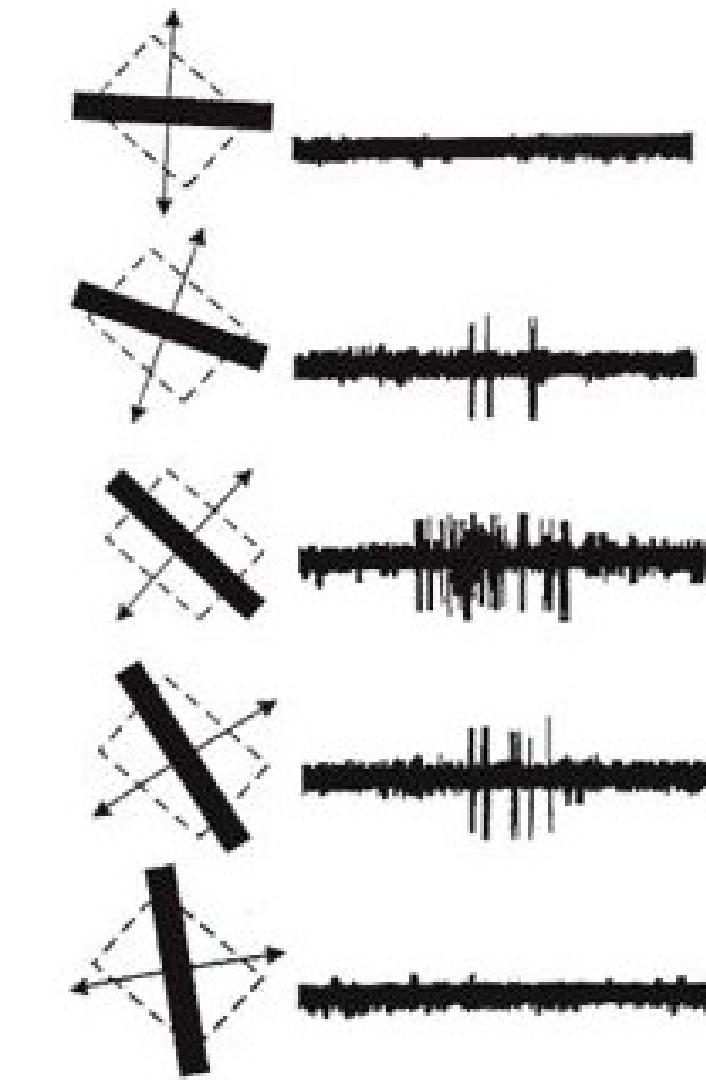
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Invariance

Consider these two neural codes

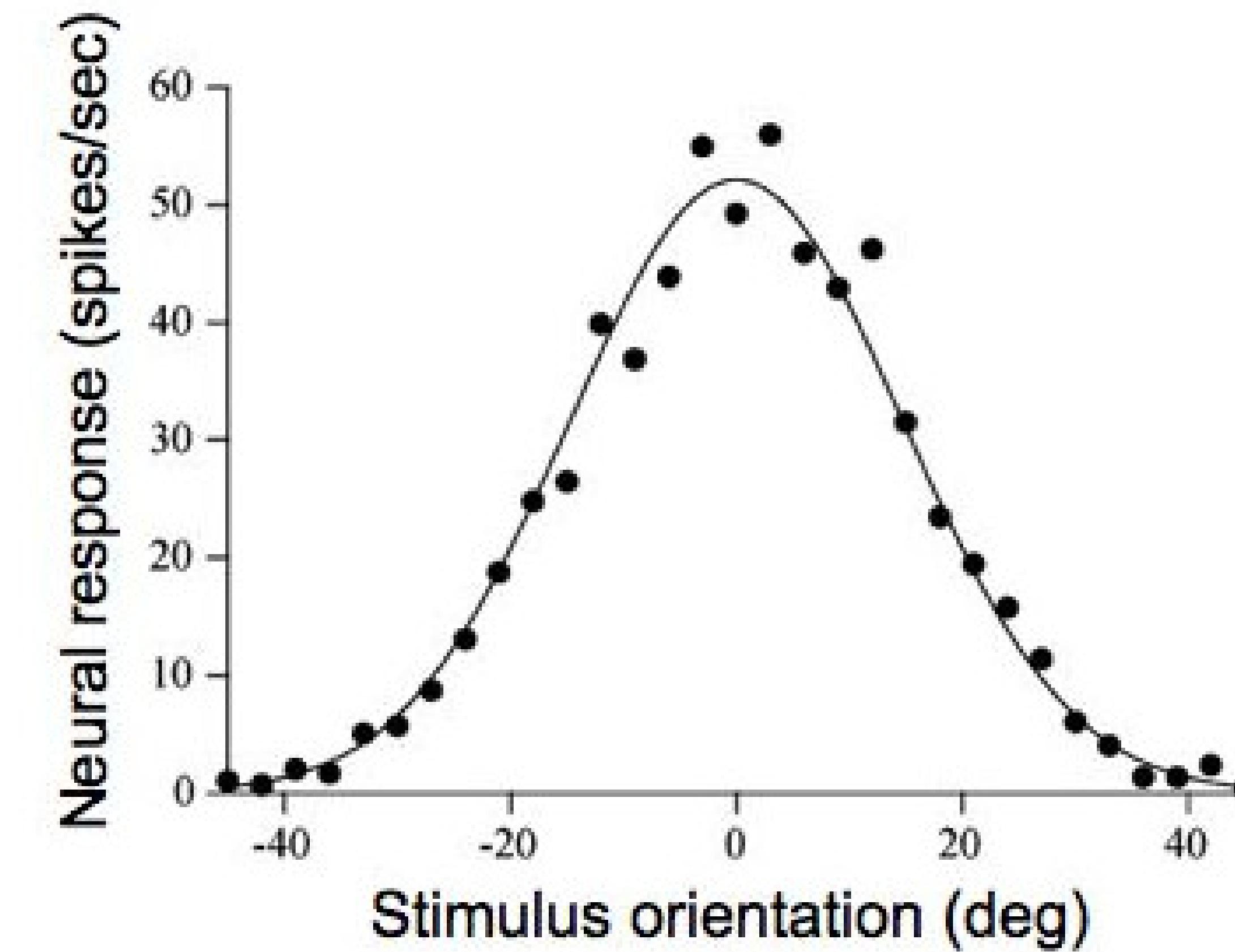
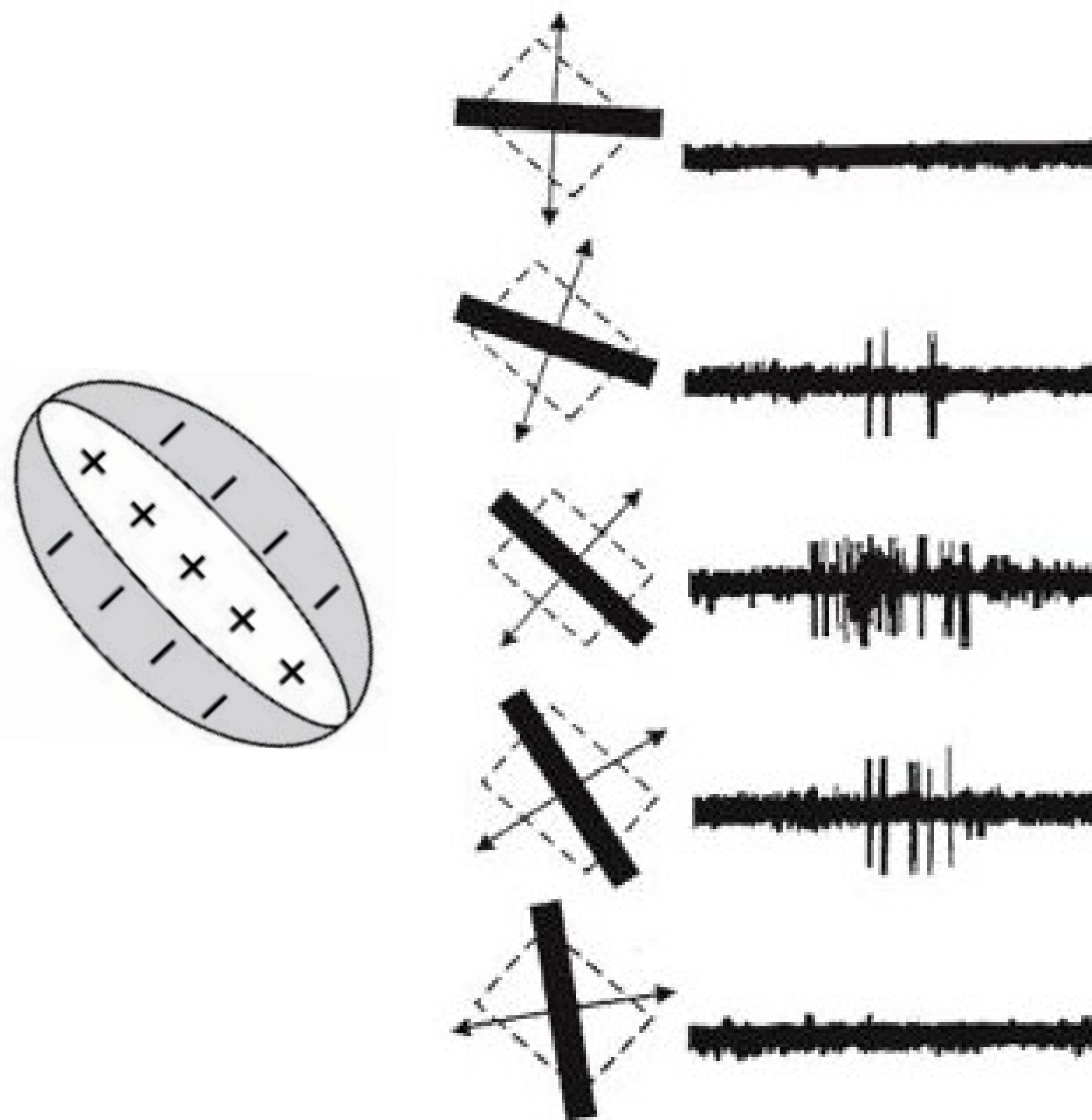
- What advantages does each of them have?
- Which one is more efficient?
- Where in the brain would you expect to find each of them?



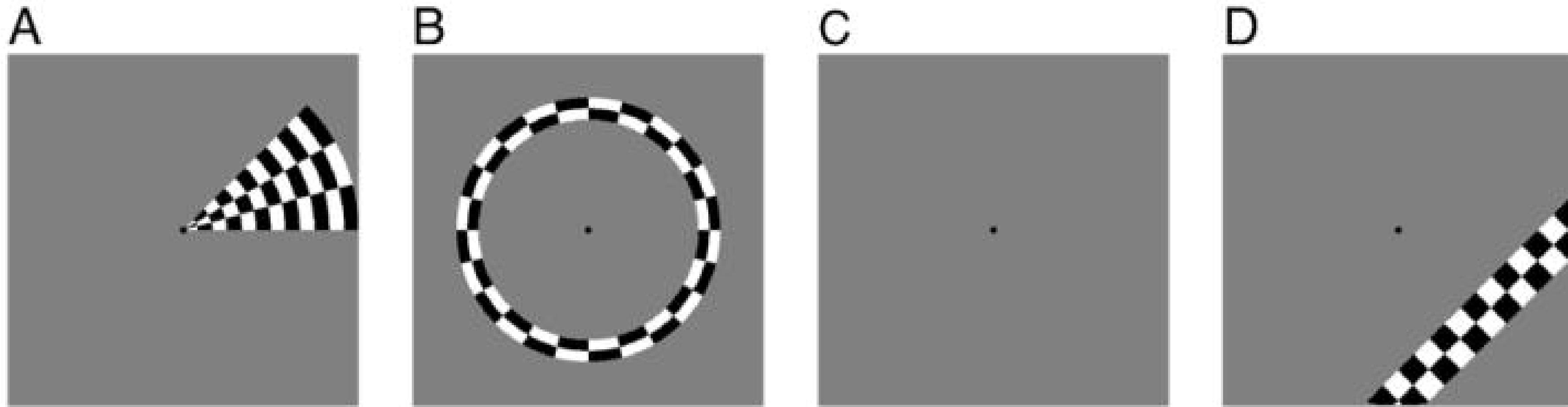
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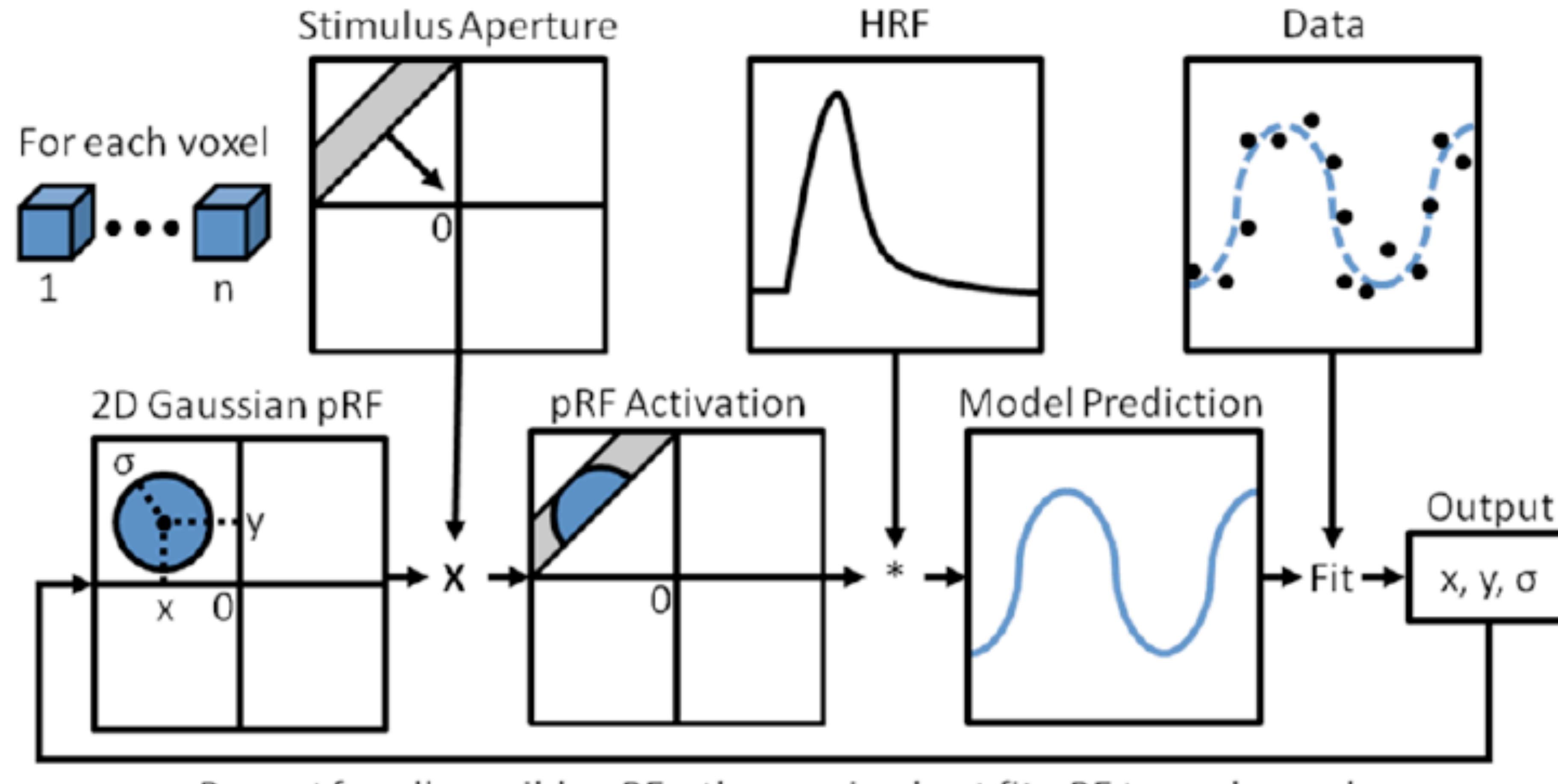
V1 receptive fields



Measuring V1 RFs with fMRI



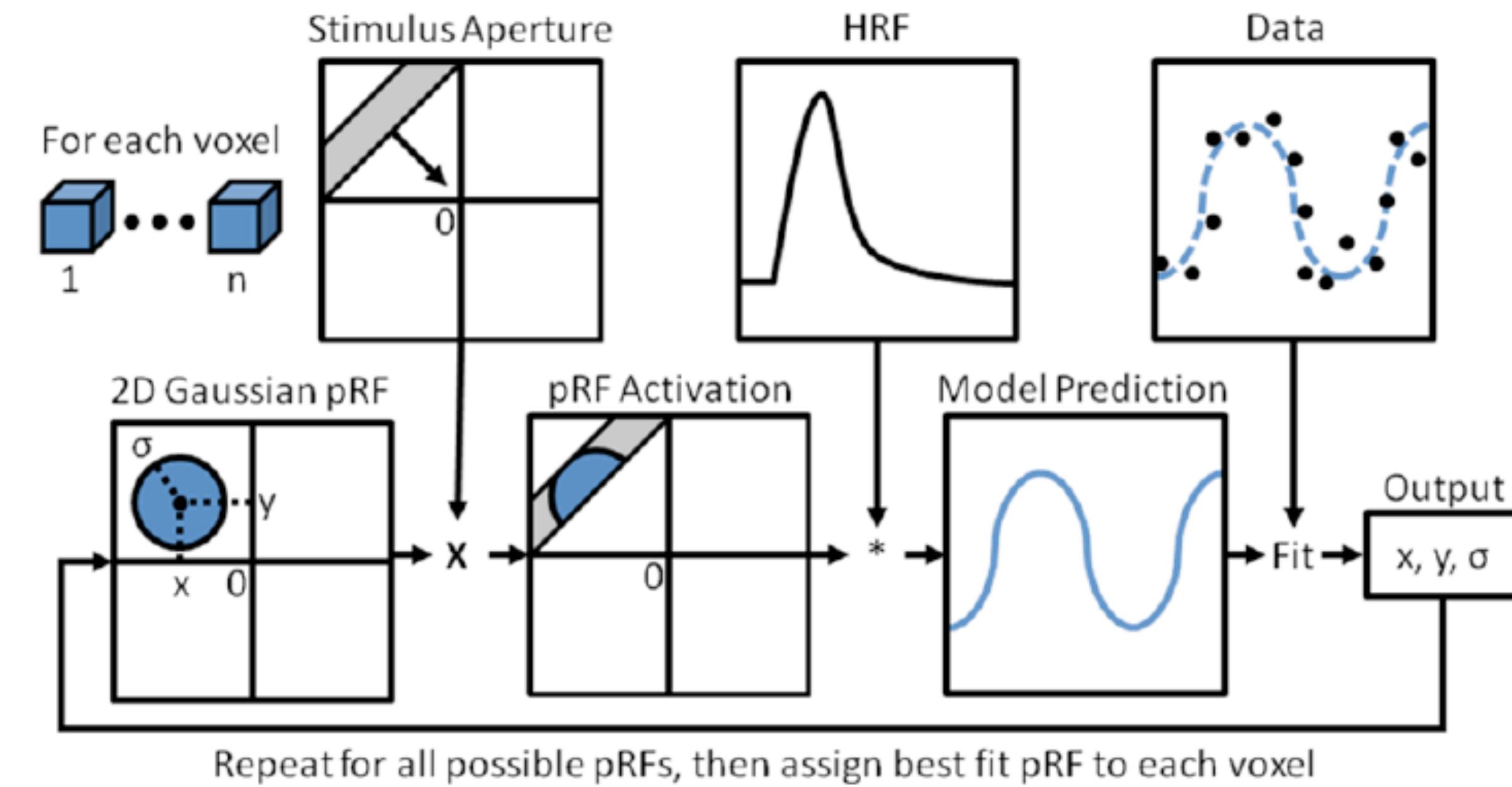
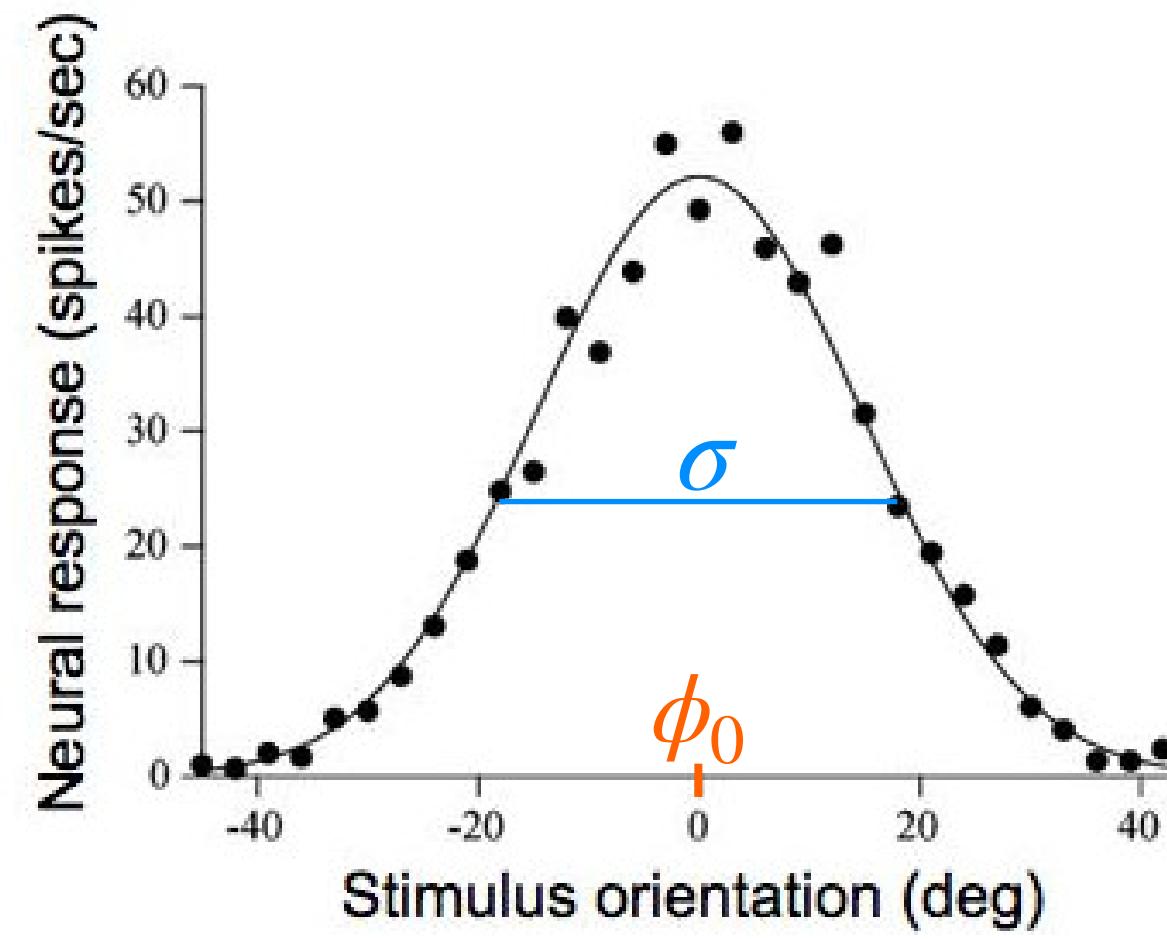
Measuring V1 RFs with fMRI



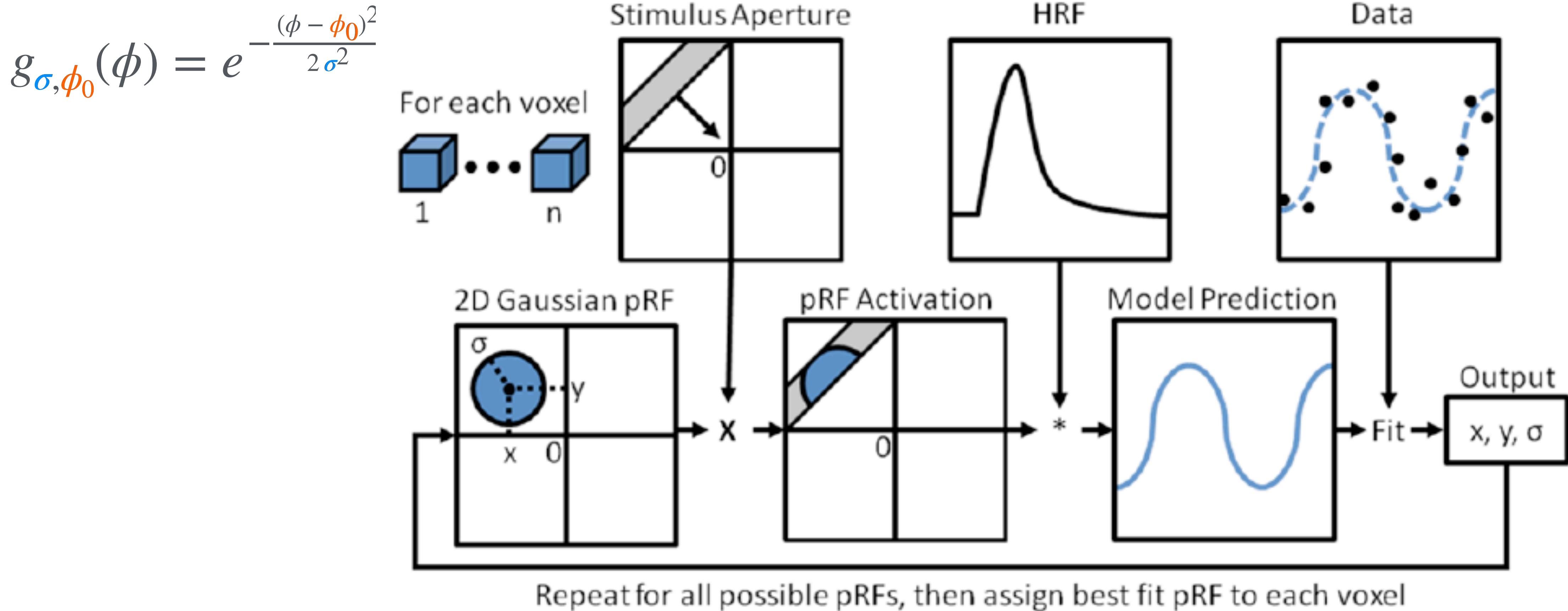
Repeat for all possible pRFs, then assign best fit pRF to each voxel

Parametrising orientation in V1 RFs

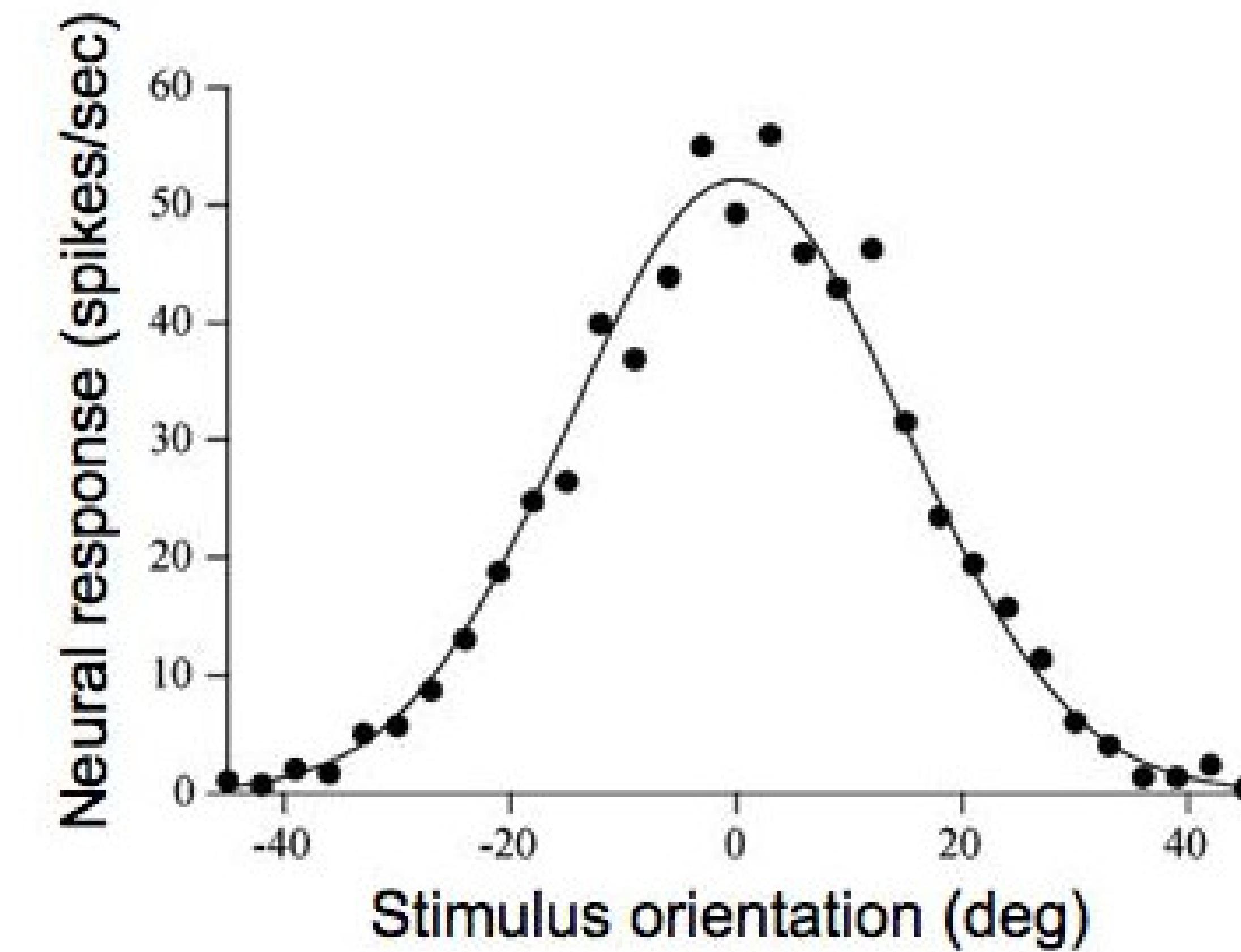
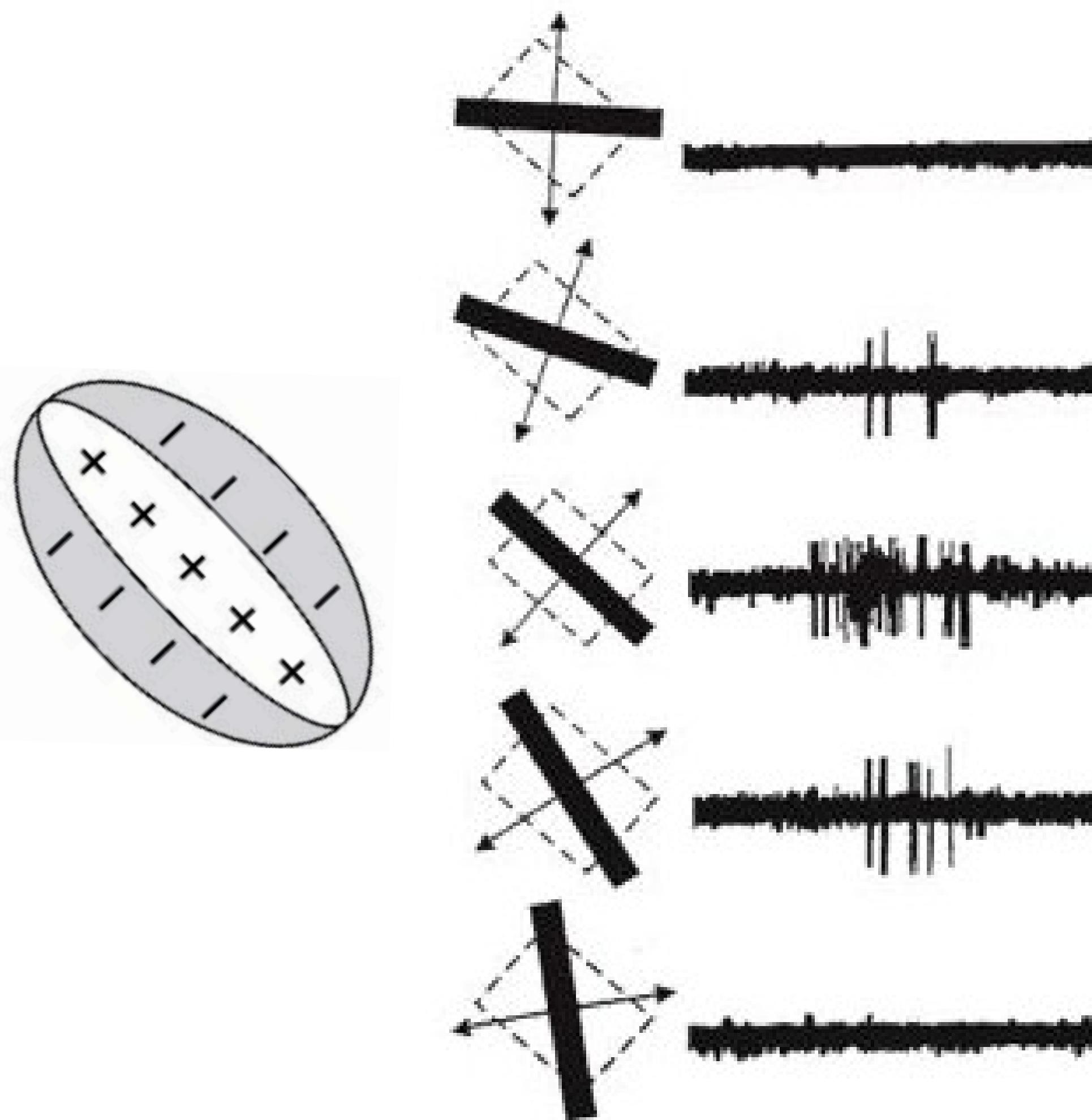
$$g_{\sigma, \phi_0}(\phi) = e^{-\frac{(\phi - \phi_0)^2}{2\sigma^2}}$$



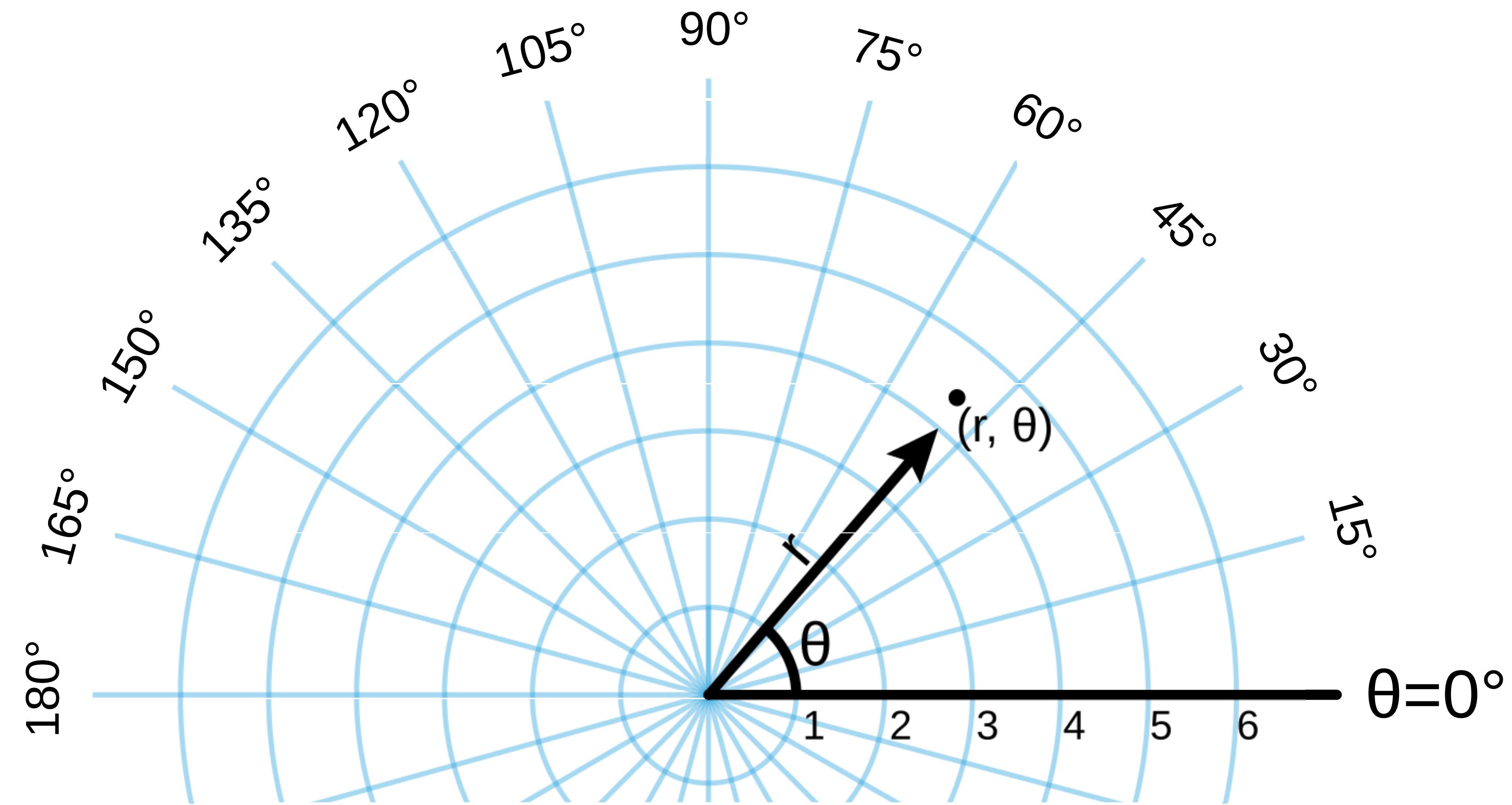
Parametrising orientation in V1 RFs



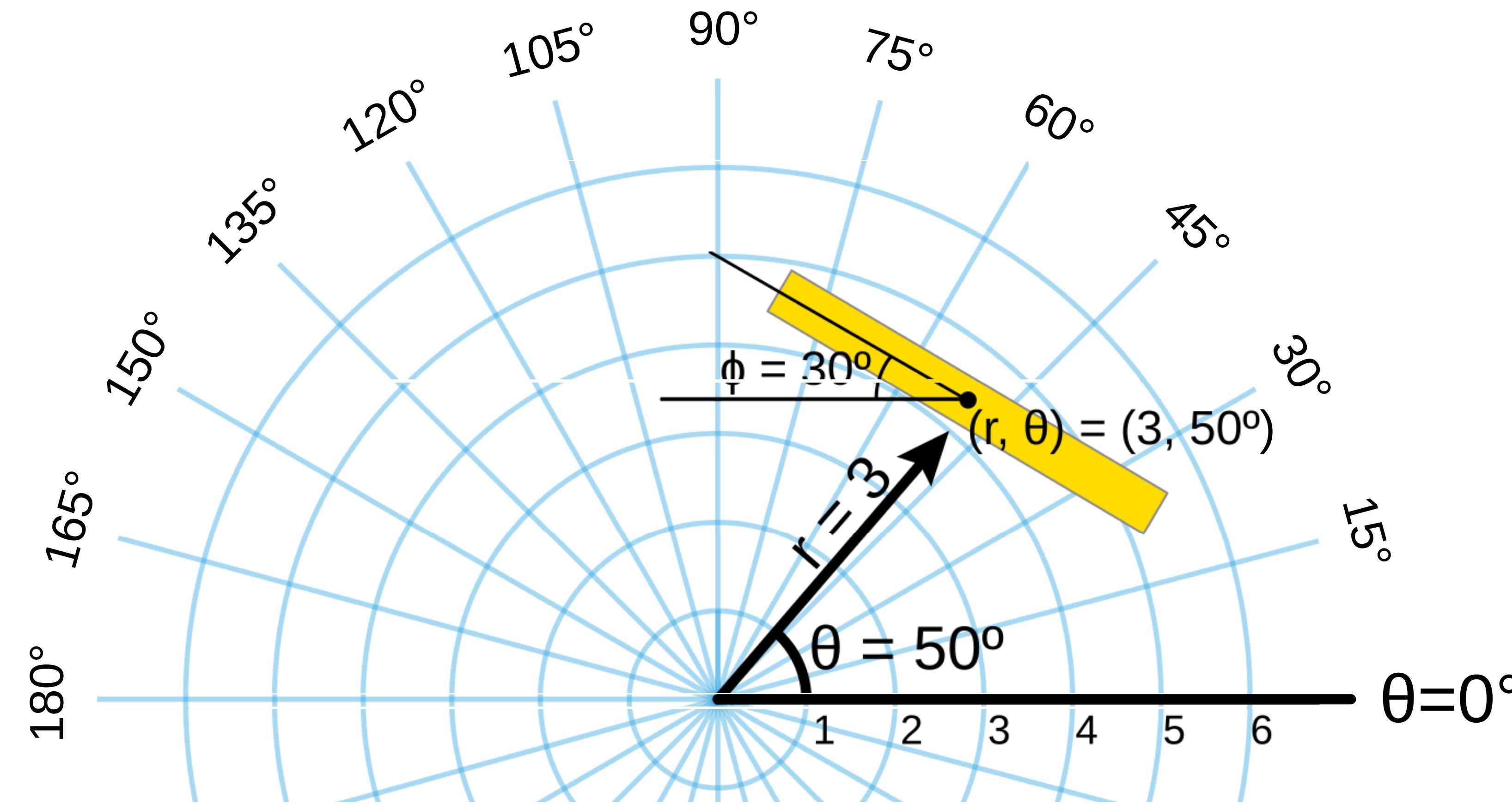
Parametrising position in V1 RFs



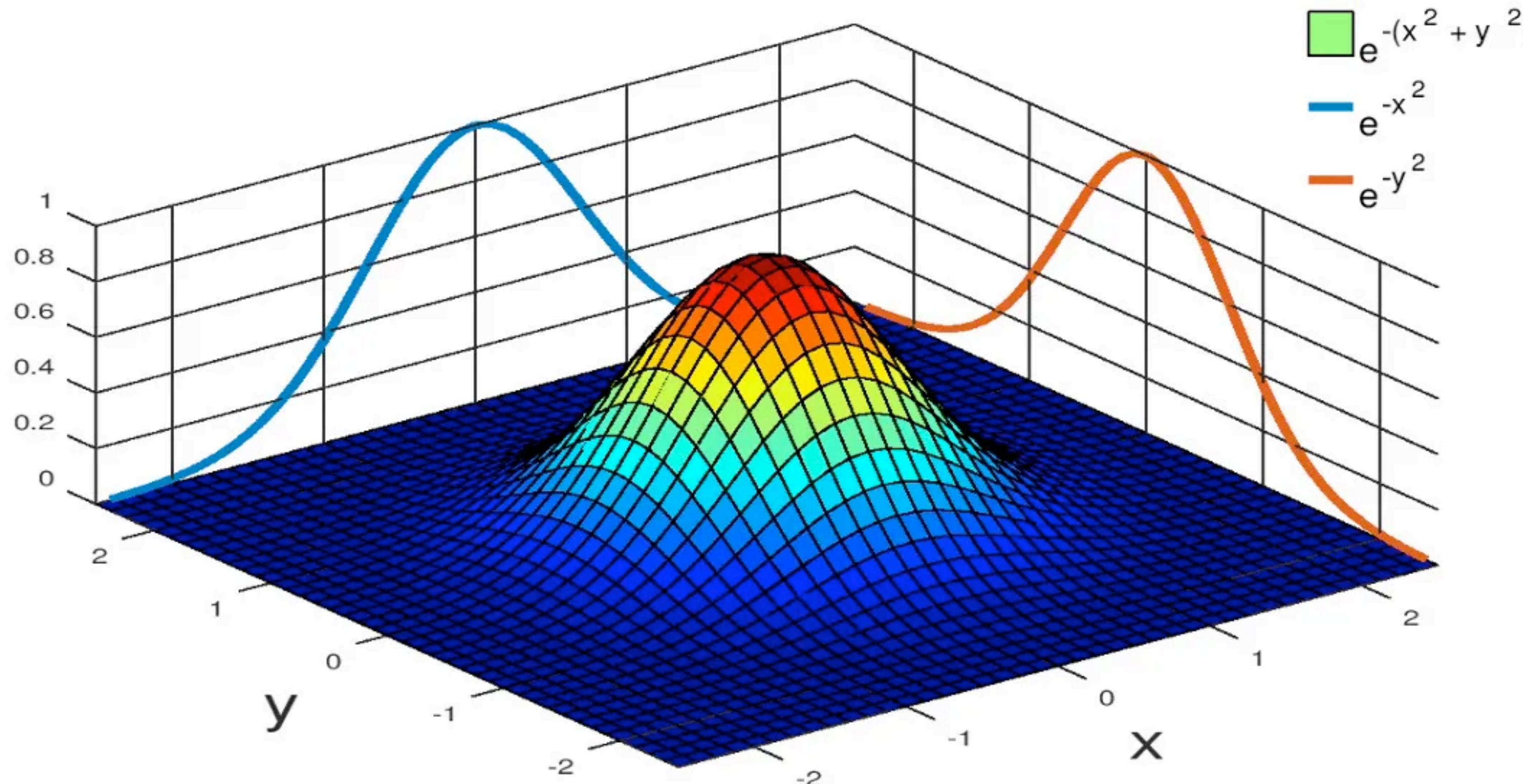
Parametrising position in V1 RFs



Parametrising position in V1 RFs

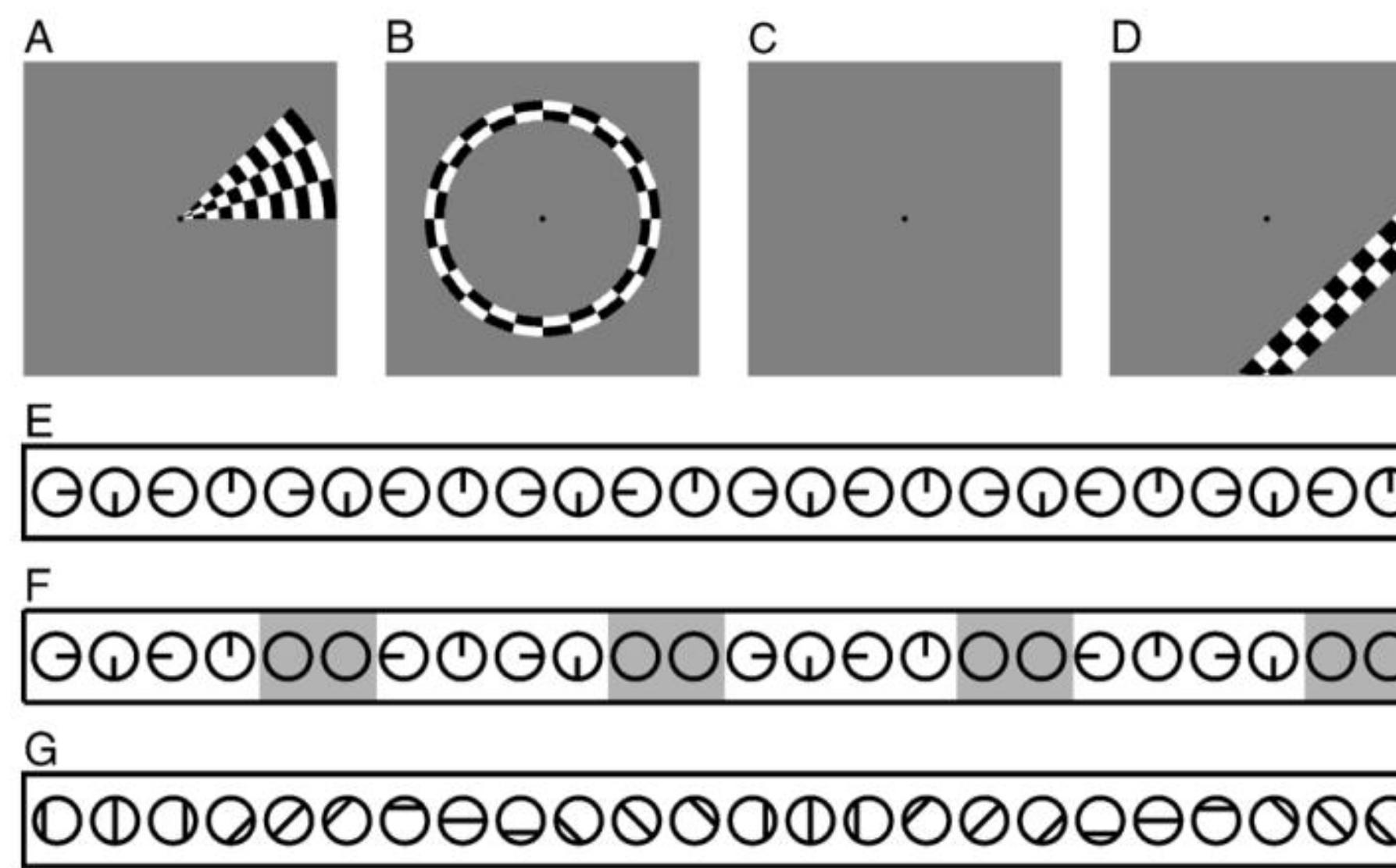


Parametrising position in V1 RFs

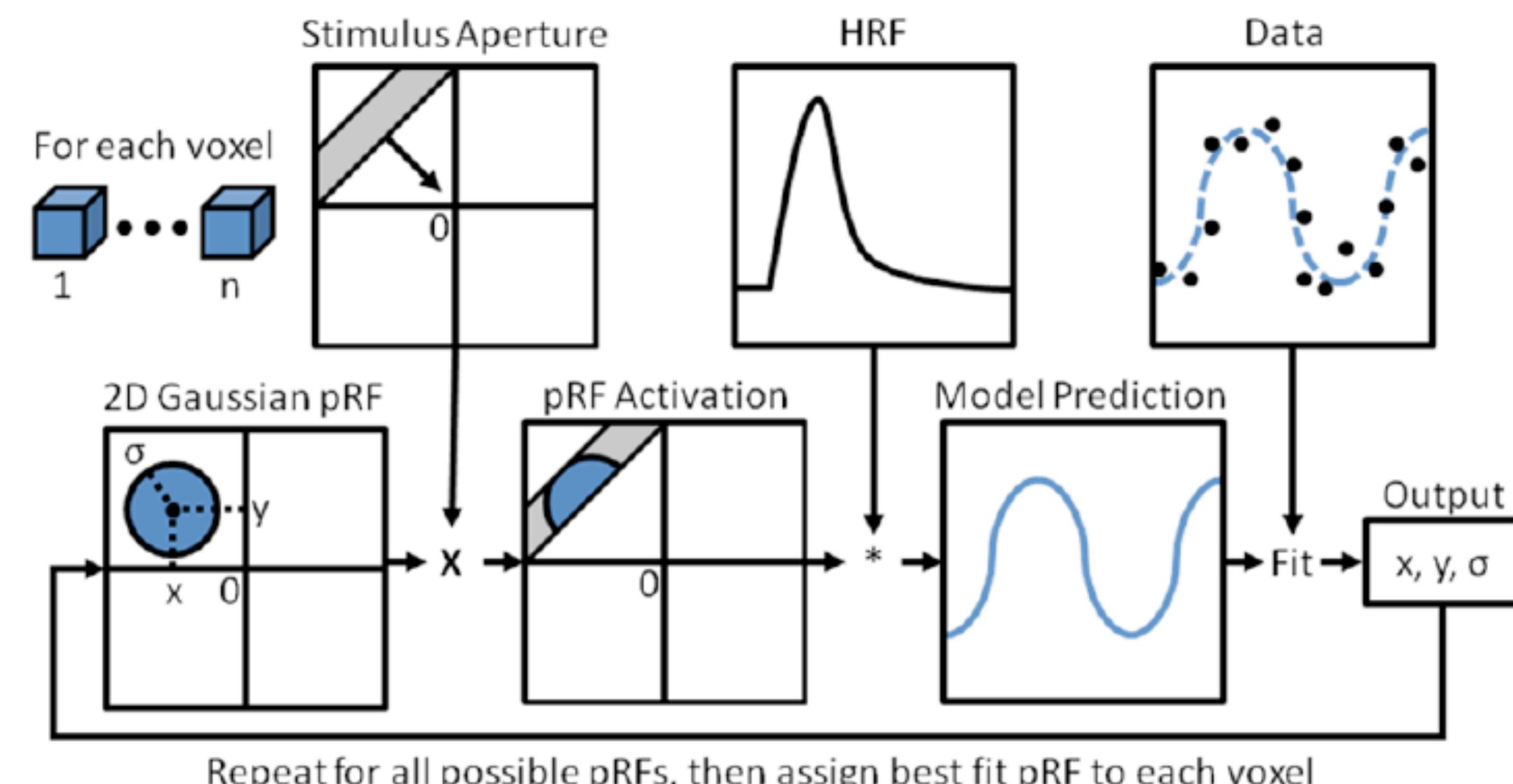
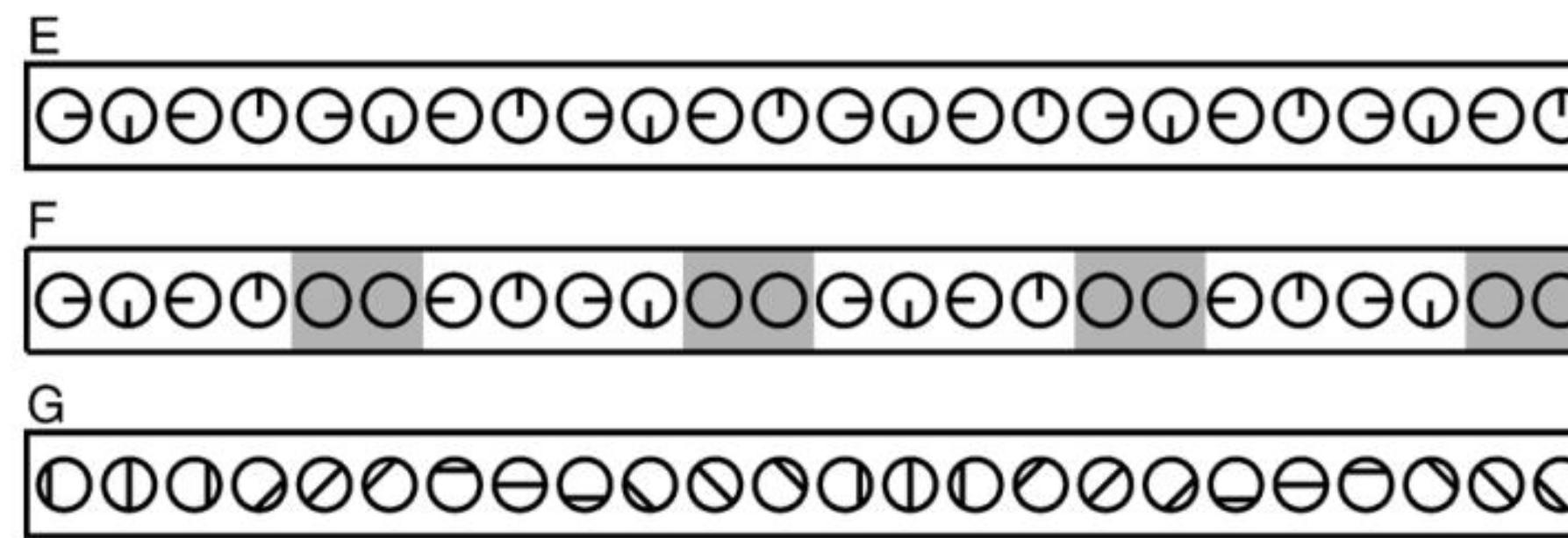


$$g_{r_0, \sigma_r, \theta_0, \sigma_\theta}(r, \theta) = e^{-\frac{(r - r_0)^2}{2\sigma_r^2} - \frac{(\theta - \theta_0)^2}{2\sigma_\theta^2}}$$

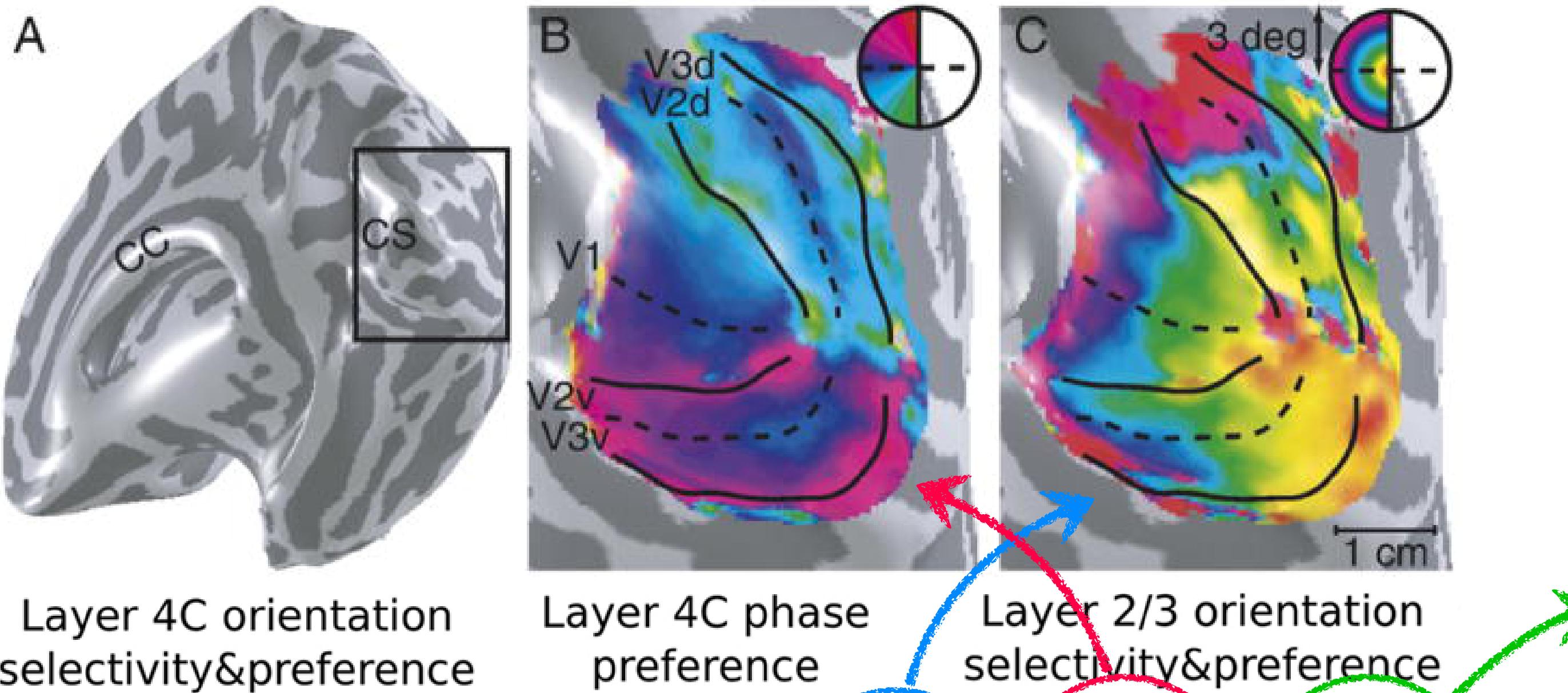
Parametrising orientation in V1 RFs



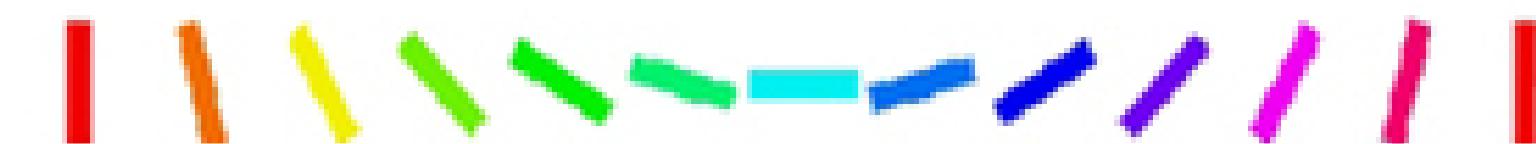
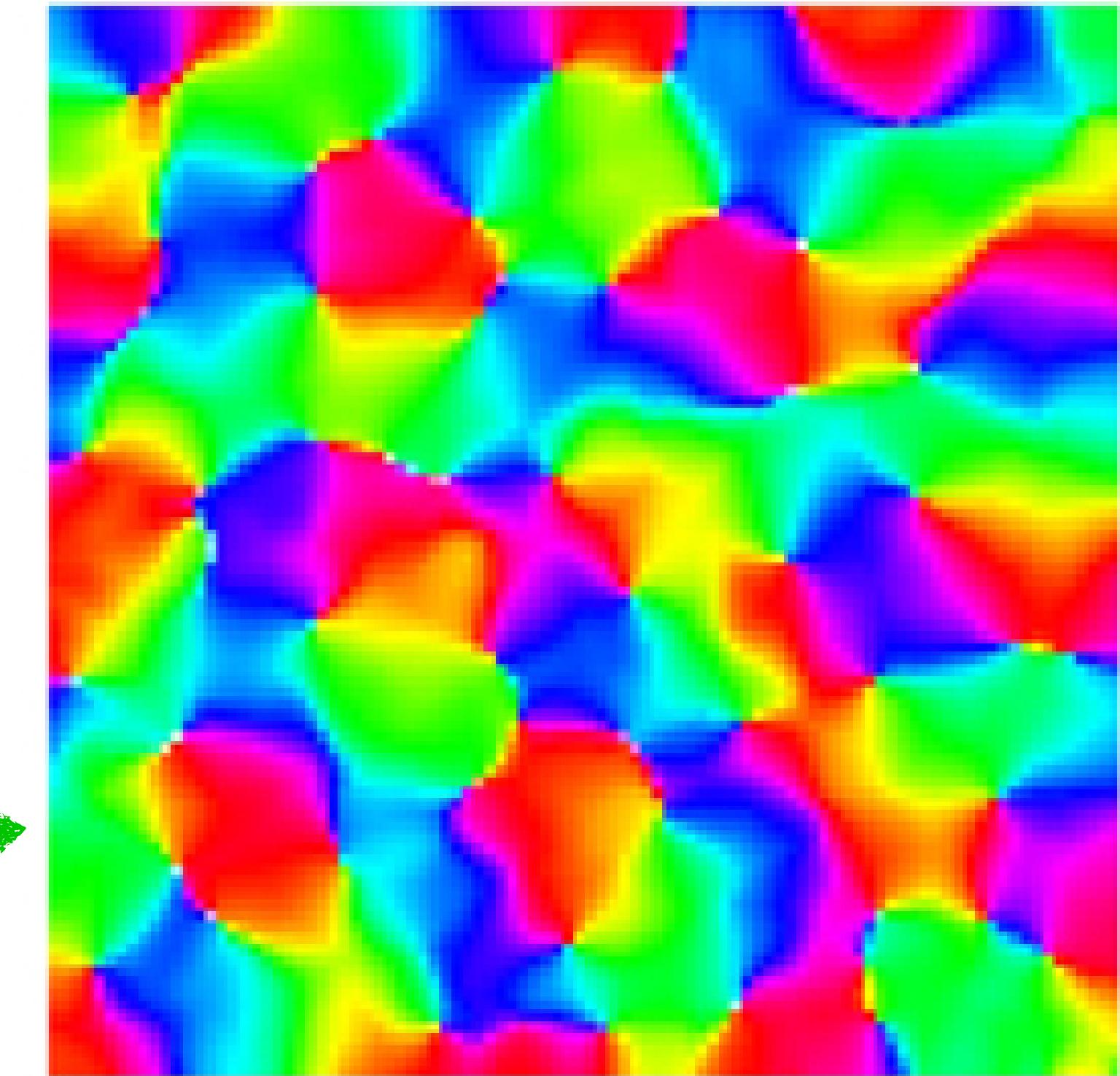
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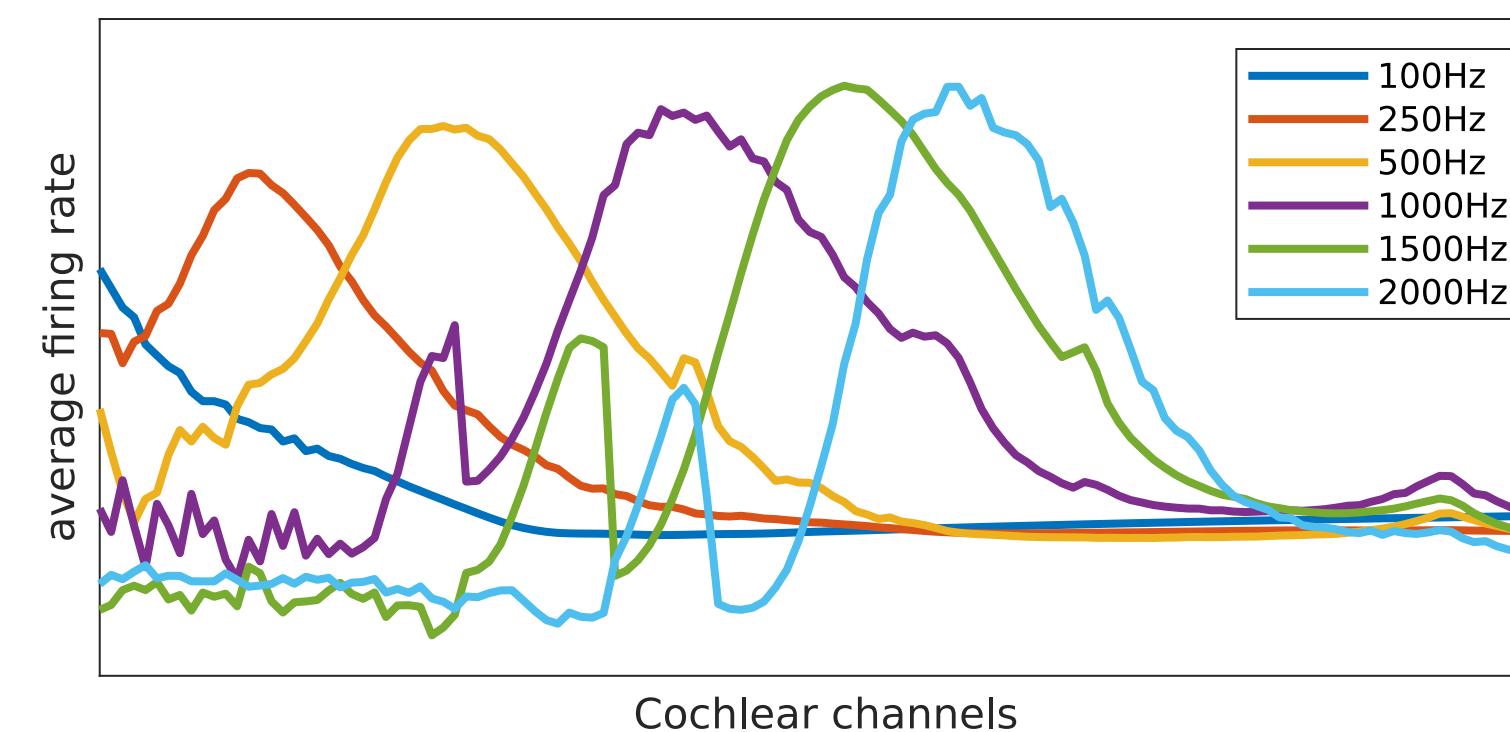
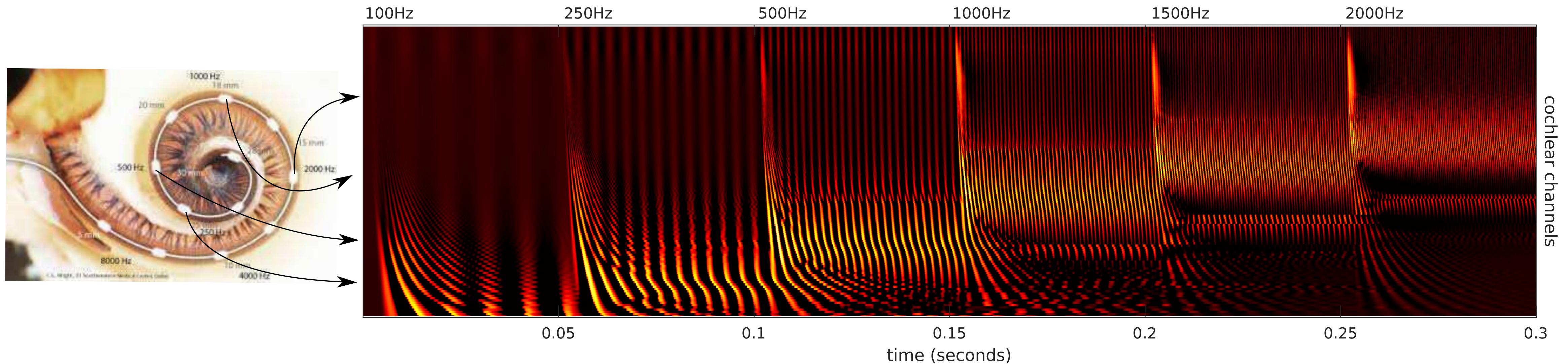
Measuring V1 RFs with fMRI



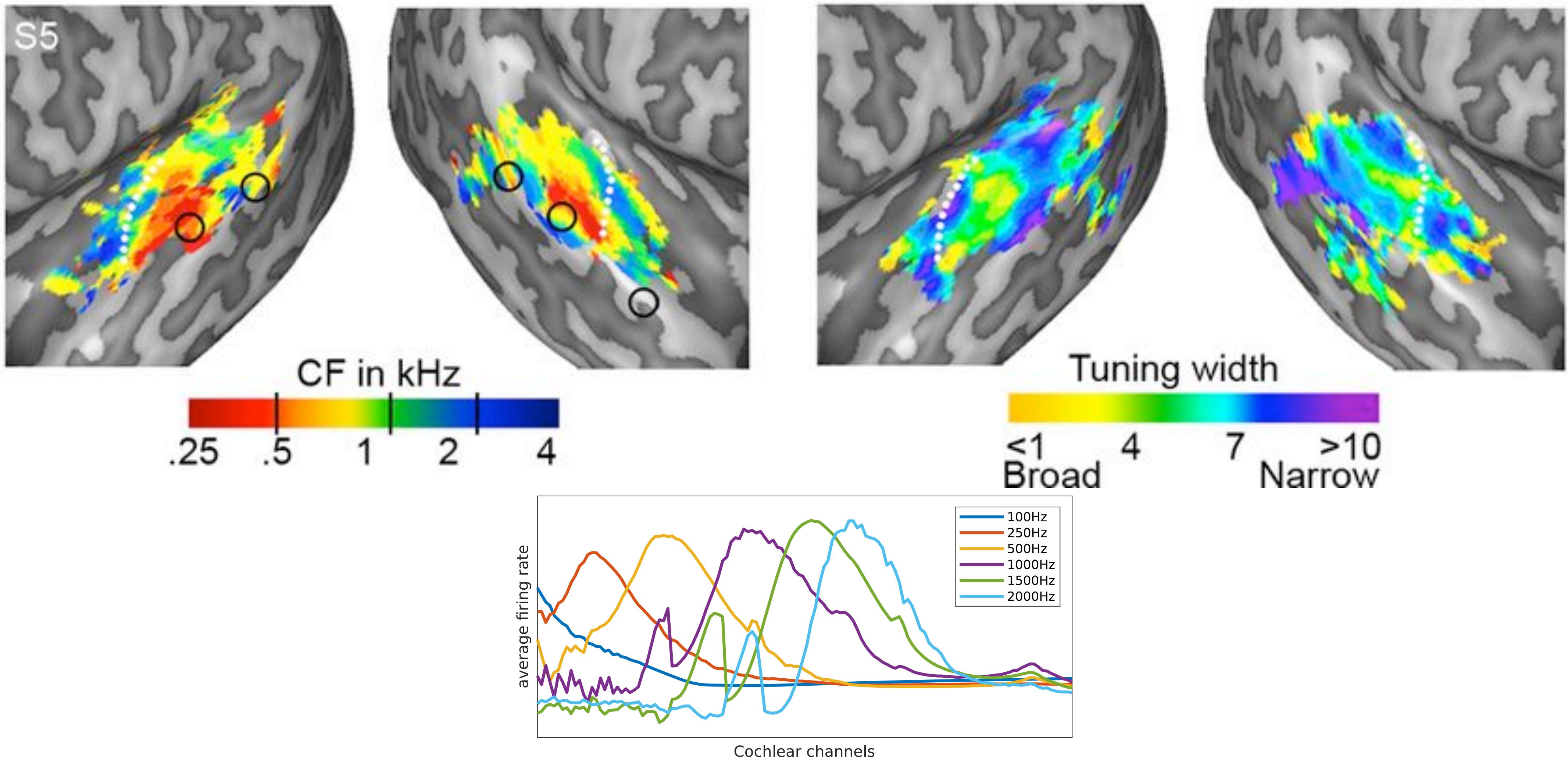
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Measuring auditory RFs with fMRI

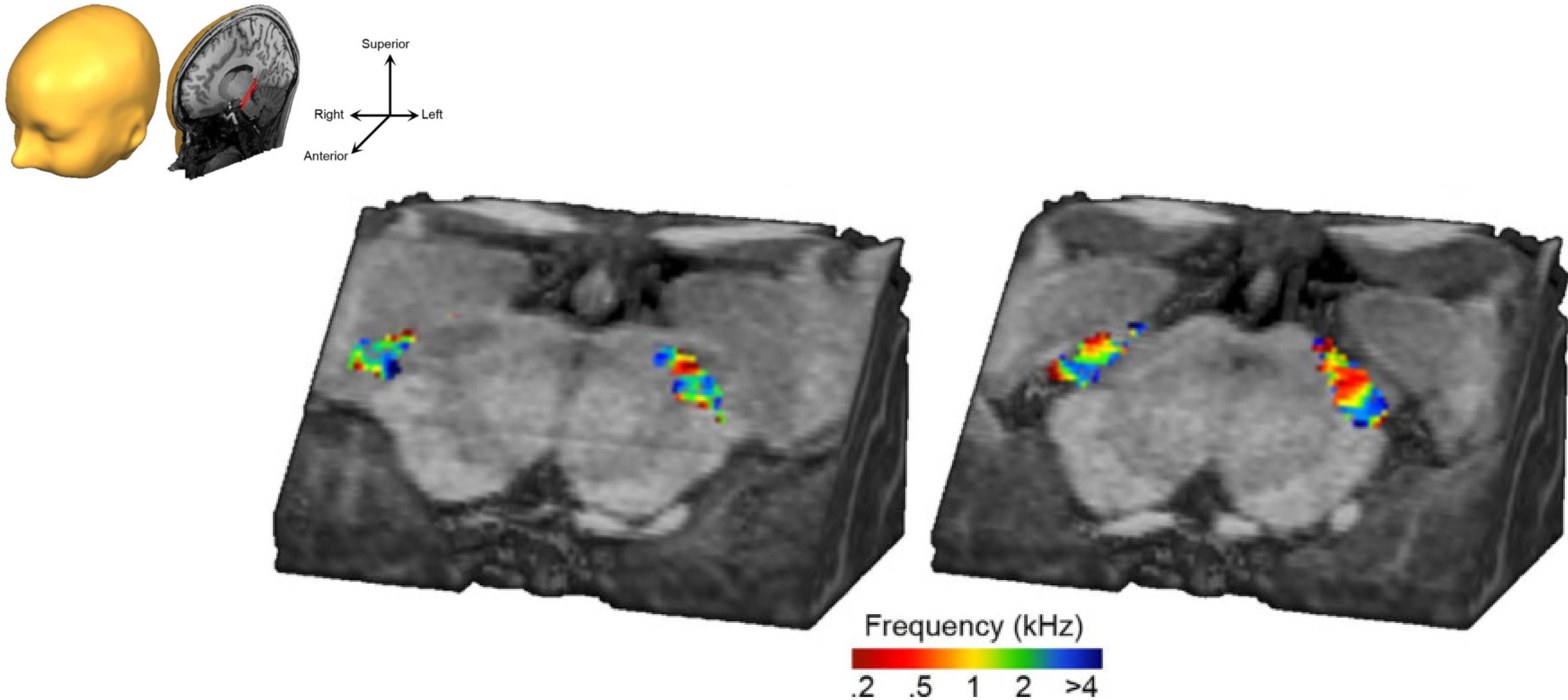


Measuring auditory RFs with fMRI

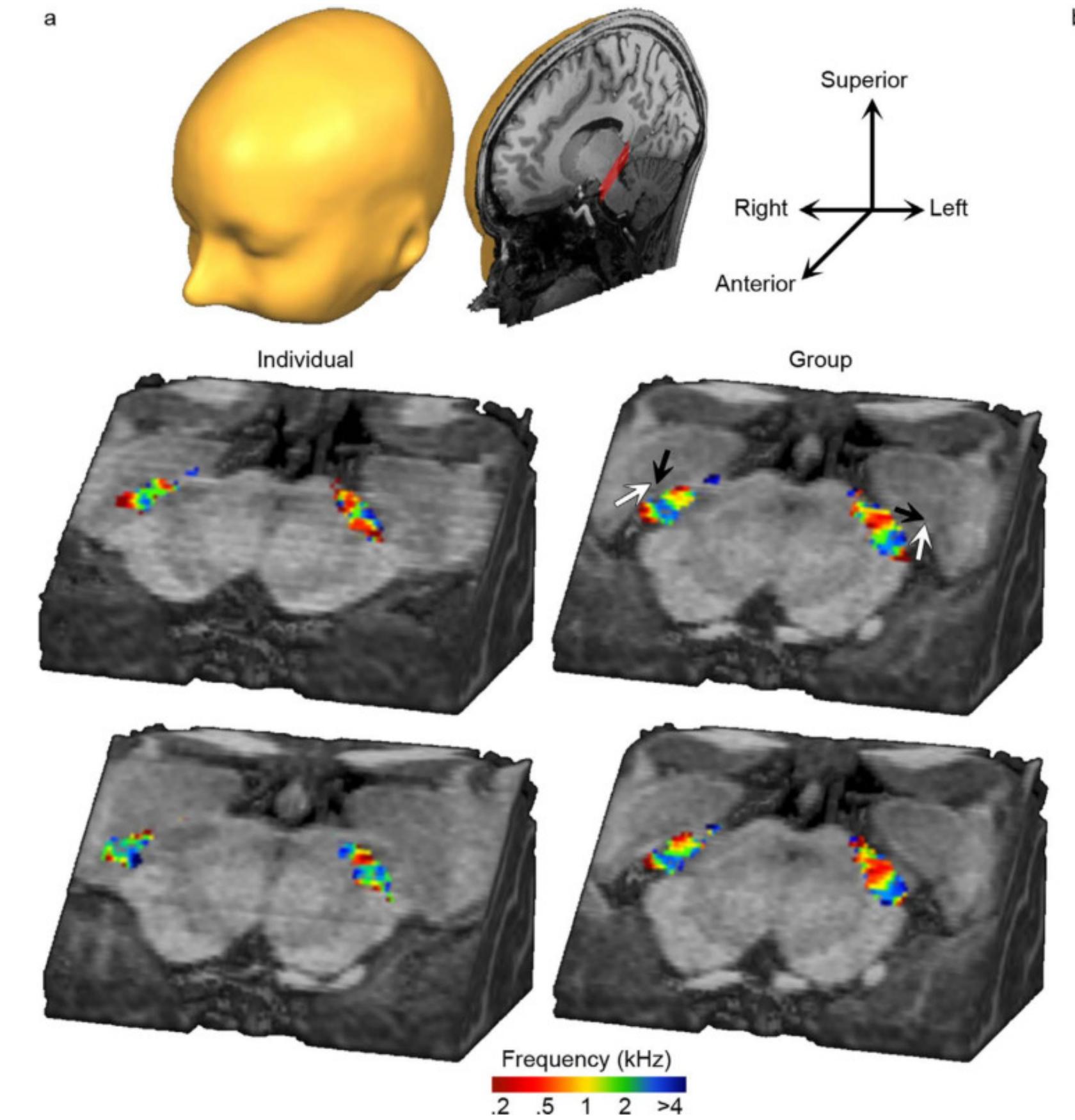
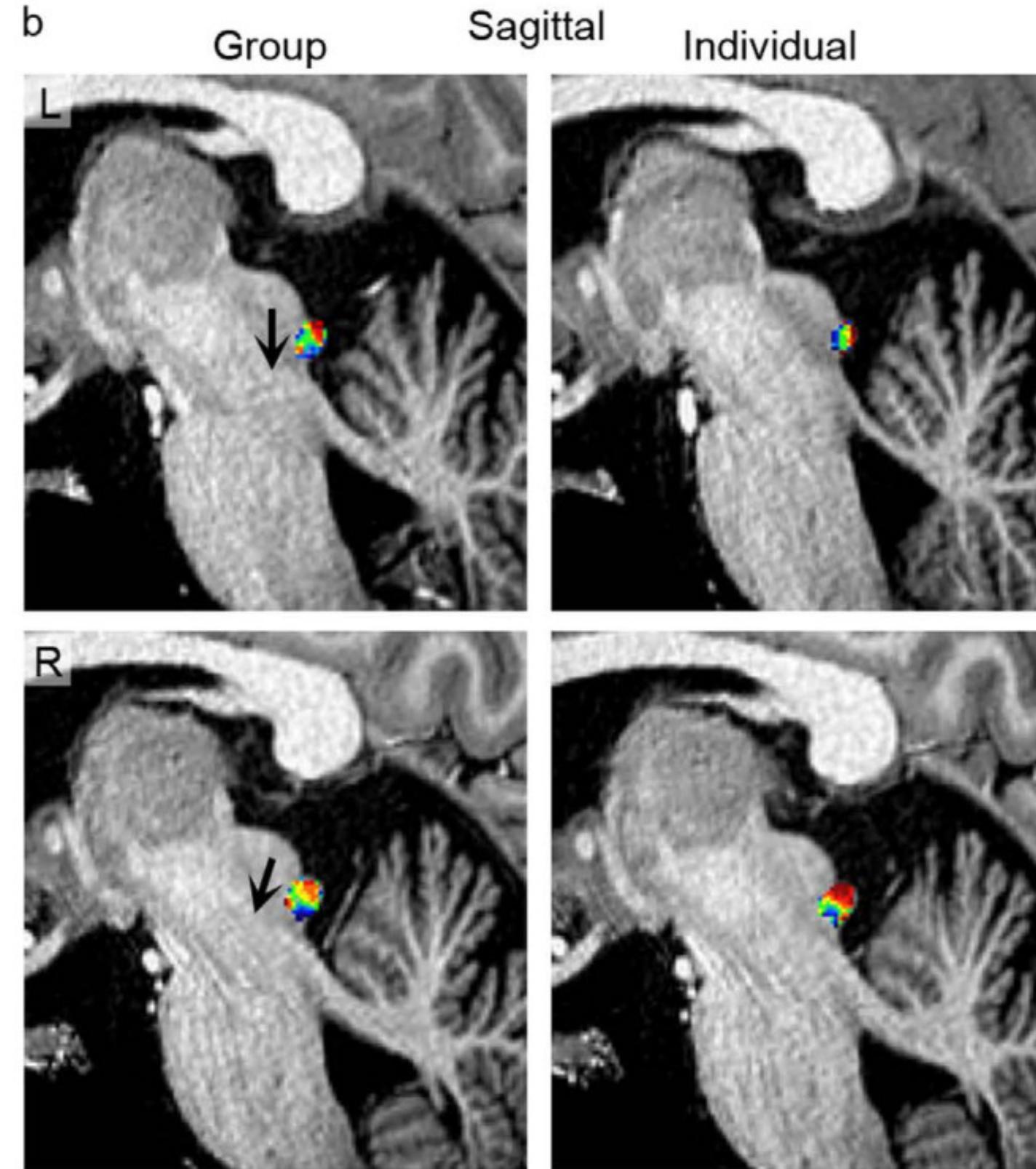
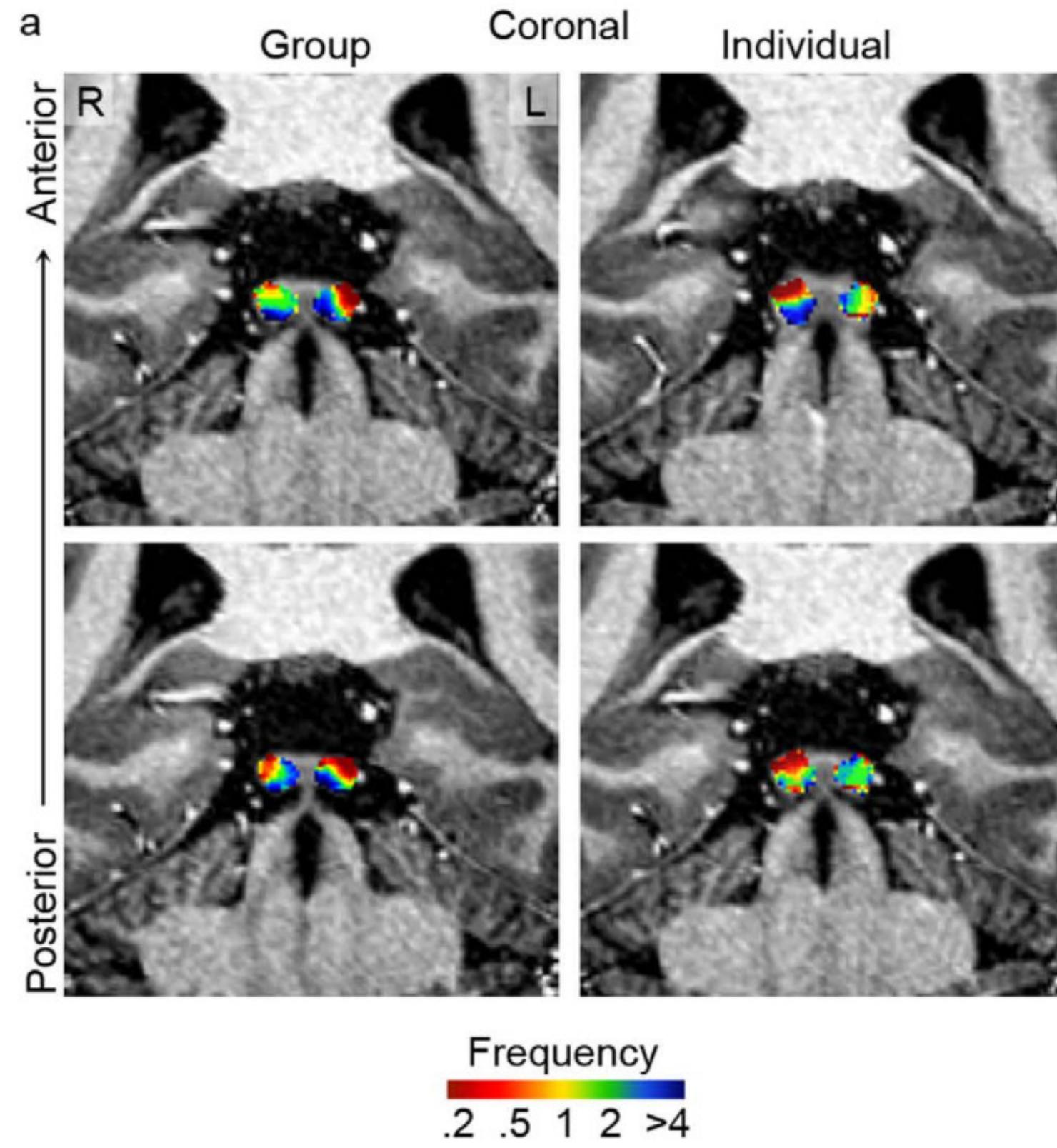


Measuring auditory RFs with fMRI

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Measuring auditory RFs with fMRI



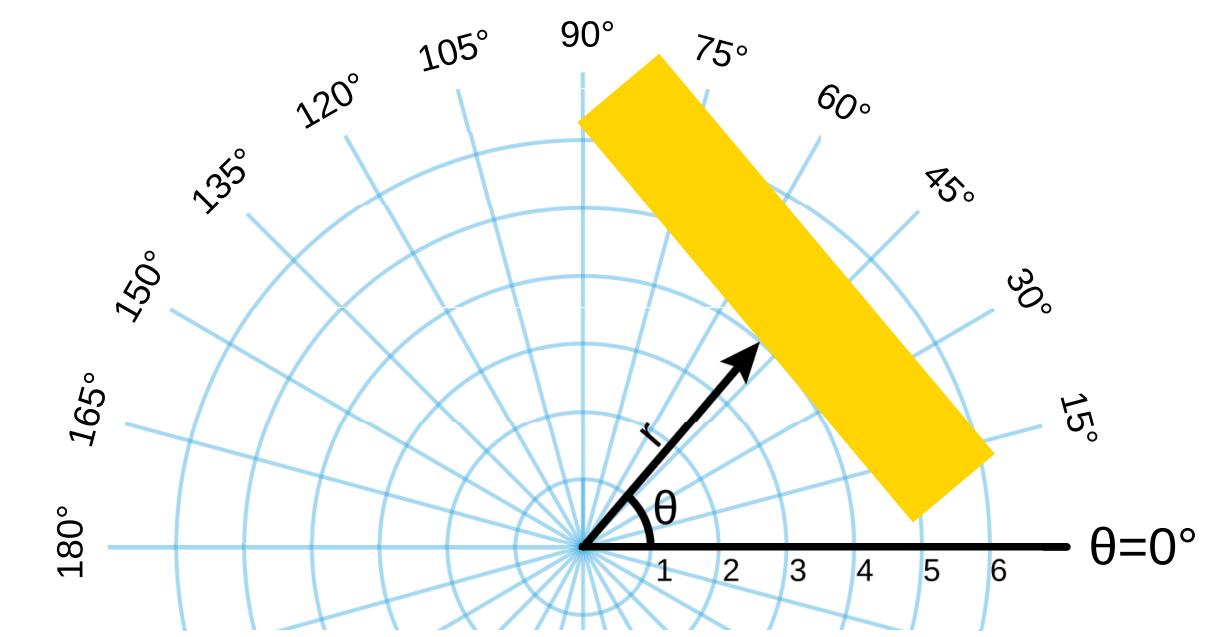
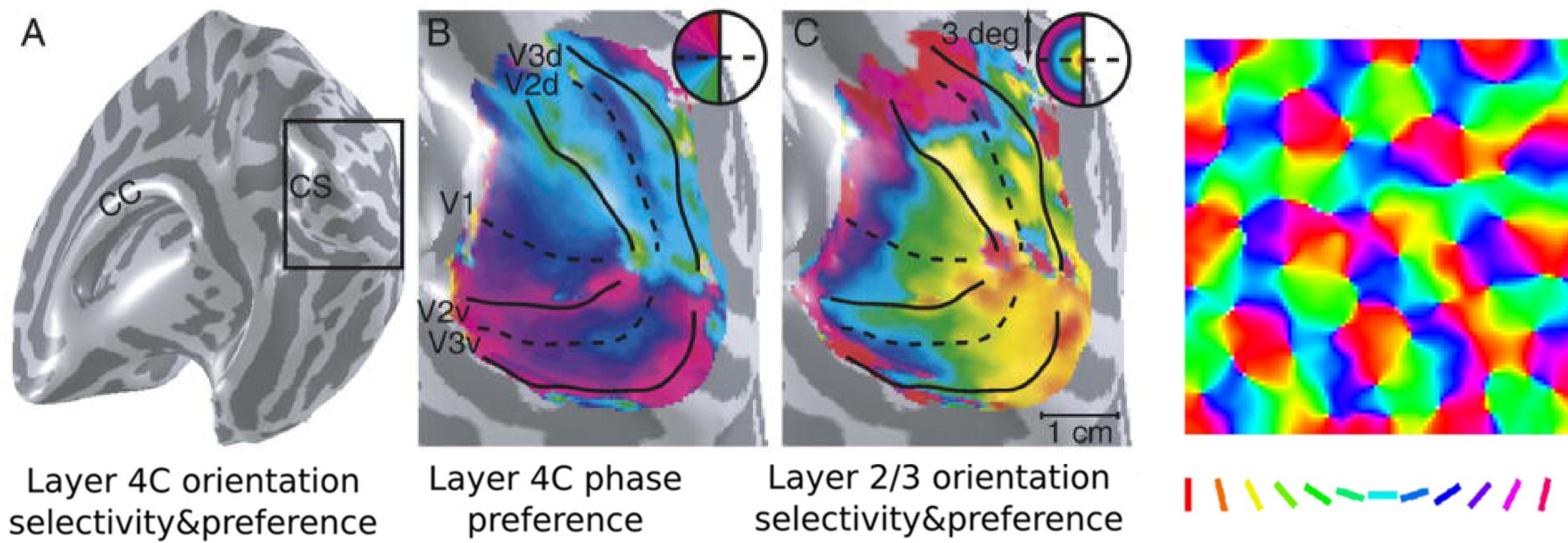
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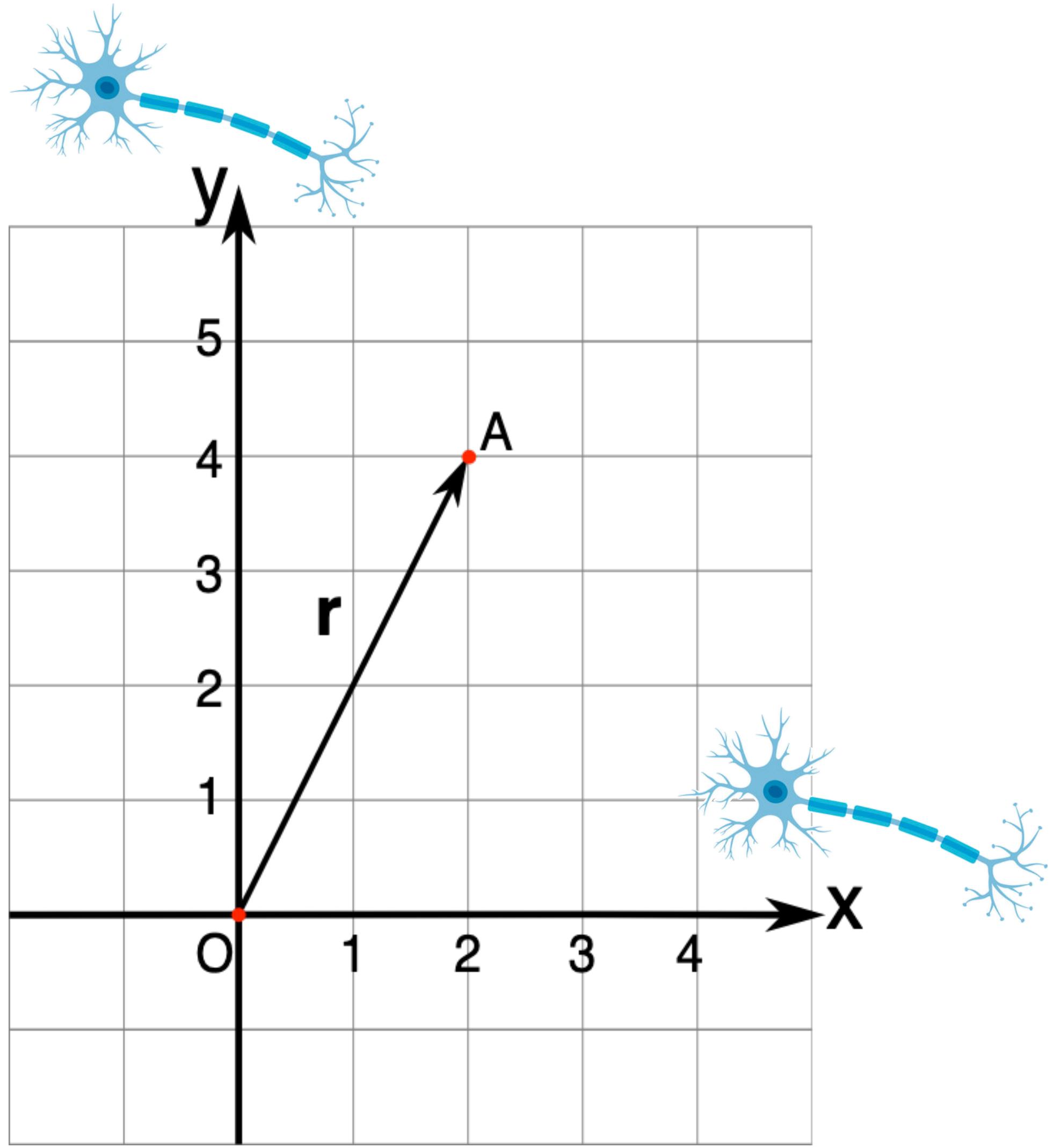
Multi-dimensional codes

So far we have showed how single units selectively encode a stimulus feature

- How would this code be used to encode this bar of light?

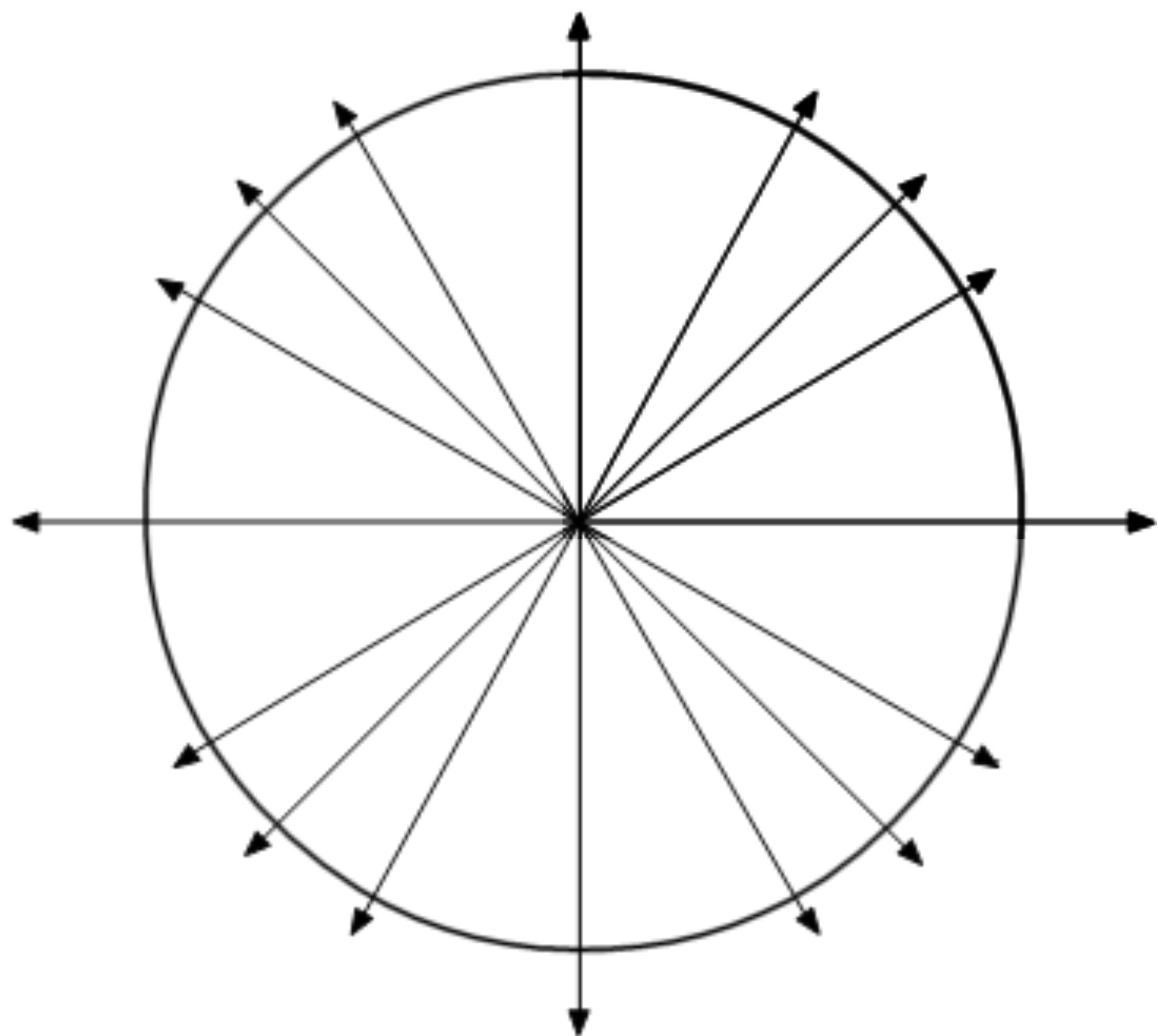


Vectors and bases



$$\begin{aligned} \left\{ \begin{array}{l} \uparrow \\ \rightarrow \end{array} \right. \\ \hookrightarrow = 1 \times \uparrow + \frac{1}{2} \times \rightarrow \end{aligned}$$

Vector spaces



{ ↑ { ↗, ↘, ↙, ↖ } }

A vector space for colour

$$RF = \{\textcolor{red}{\bullet}, \textcolor{green}{\bullet}, \textcolor{blue}{\bullet}\}$$

$$\textcolor{teal}{\bullet} = \frac{1}{2} \times \textcolor{red}{\bullet} + \frac{3}{4} \times \textcolor{green}{\bullet} + \frac{3}{4} \times \textcolor{blue}{\bullet}$$

A vector space for retinal activation

$$RF = \left\{ \begin{array}{c} \text{grid with 1 black square at top-left} \\ \text{grid with 1 black square at top-center} \\ \text{grid with 1 black square at top-right} \\ \vdots \\ \text{grid with 1 black square at bottom-right} \end{array} \right\}$$

$$\begin{array}{c} \text{grid with 1 black square at top-left, 1 gray square at center} \\ = 1 \times \text{grid with 1 black square at top-left} + 1 \times \text{grid with 1 black square at top-center} + 1 \times \text{grid with 1 black square at top-right} + \frac{1}{2} \times \text{grid with 1 black square at center} + \frac{1}{2} \times \text{grid with 1 black square at bottom-center} + \frac{1}{4} \times \text{grid with 1 black square at bottom-right} \end{array}$$

Another vector space for retinal activation

$$RF = \left\{ \text{ovals} \right\}$$

$$\text{grid} = 1 \times \text{oval} + \frac{1}{2} \times \text{oval} + \frac{1}{4} \times \text{oval}$$

Composition of vector spaces

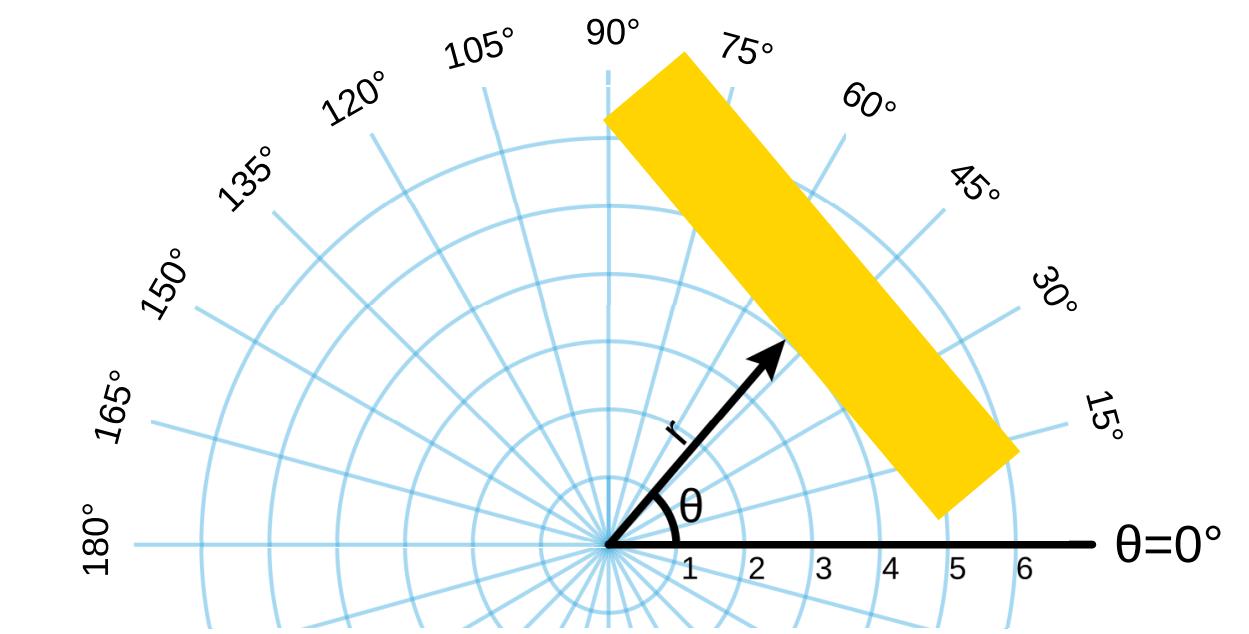
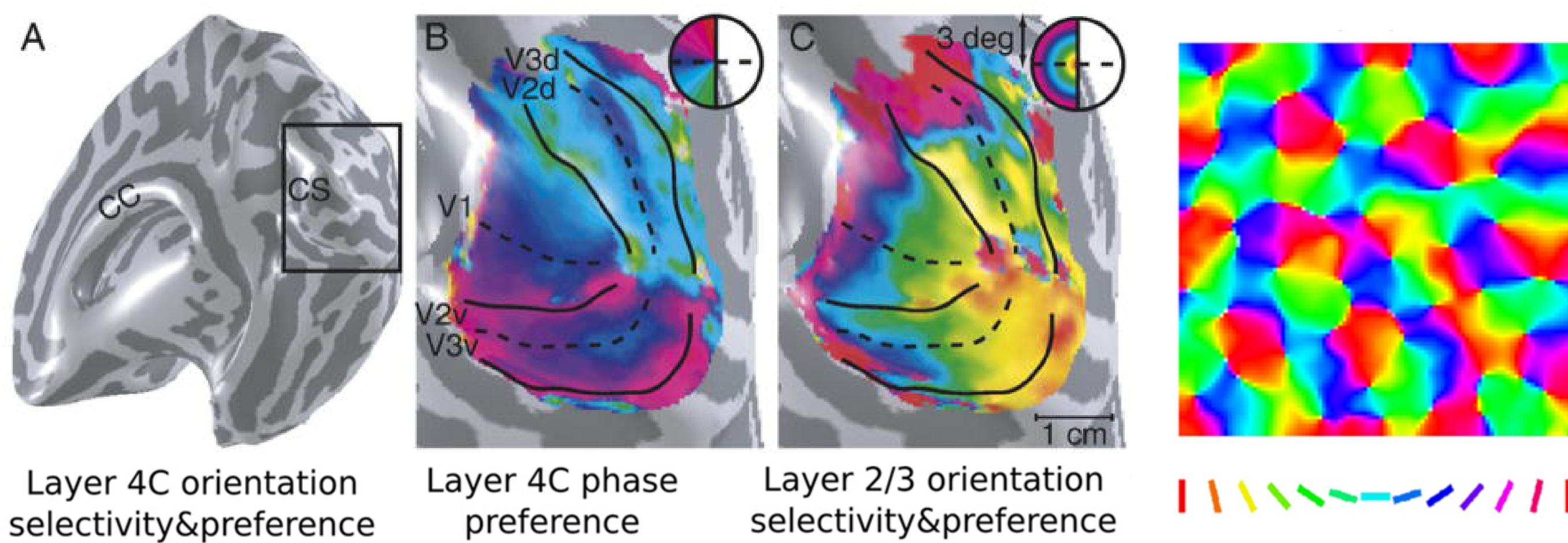
$$RF = \left\{ \begin{array}{c} \text{gray oval} \\ + + + + \\ - - - - \\ + + + + \end{array}, \begin{array}{c} \text{gray oval} \\ + + + + \\ - - - - \\ + + + + \end{array}, \begin{array}{c} \text{gray oval} \\ x x x x \\ x x x x \\ x x x x \end{array}, \dots, \begin{array}{c} \text{gray oval} \\ x x x x \\ x x x x \\ x x x x \end{array} \end{array} \right\} \otimes \left\{ \begin{array}{c} \text{red circle} \\ \times \times \times \times \\ \times \times \times \times \\ \times \times \times \times \end{array}, \begin{array}{c} \text{green circle} \\ \times \times \times \times \\ \times \times \times \times \\ \times \times \times \times \end{array}, \begin{array}{c} \text{blue circle} \\ \times \times \times \times \\ \times \times \times \times \\ \times \times \times \times \end{array} \end{array} \right\}$$

$$\begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline \end{array} = \left(1 \times \begin{array}{c} \text{gray oval} \\ + + + + \\ - - - - \\ + + + + \end{array} + \frac{1}{2} \times \begin{array}{c} \text{gray oval} \\ + + + + \\ - - - - \\ + + + + \end{array} + \frac{1}{4} \times \begin{array}{c} \text{gray oval} \\ x x x x \\ x x x x \\ x x x x \end{array} \right) \otimes \left(\frac{1}{2} \times \text{red circle} + \frac{3}{4} \times \text{green circle} + \frac{3}{4} \times \text{blue circle} \right)$$

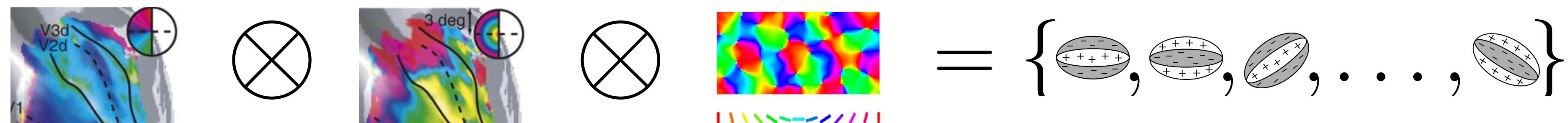
Multi-dimensional codes

So far we have showed how single units selectively encode a stimulus feature

- How would this code be used to encode this bar of light?



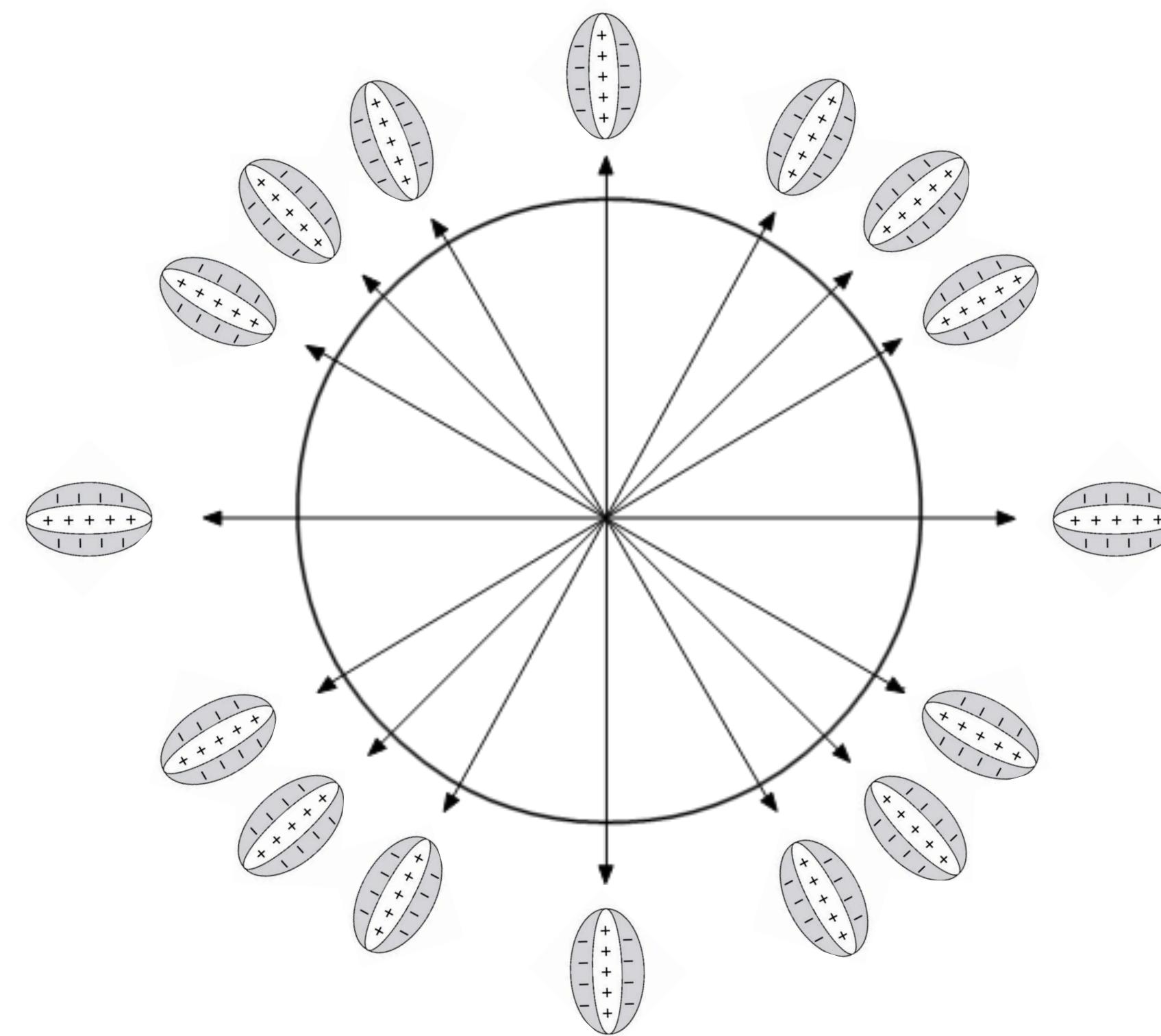
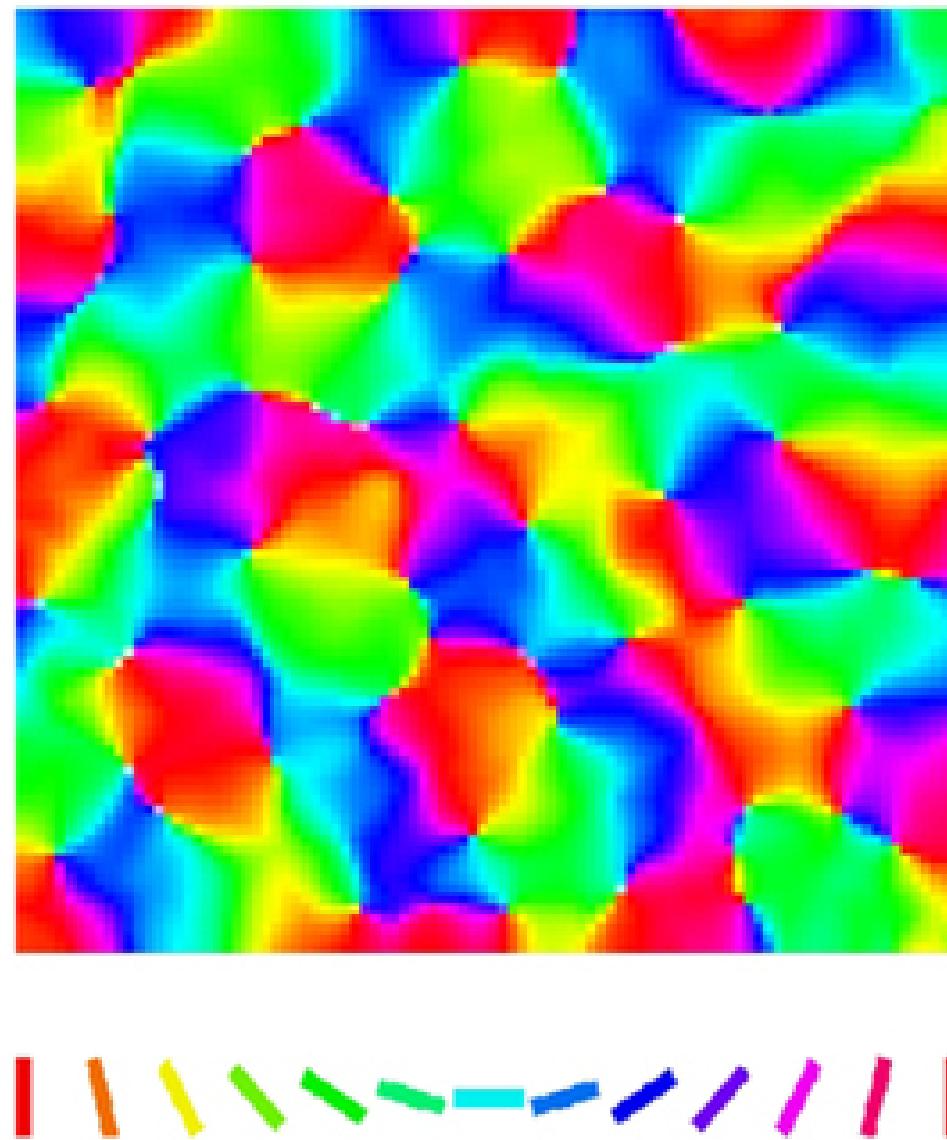
The bar or light can be encoded in the space spanned by the basis vectors:



Beyond single-unit codes

So far we have showed how single units selectively encode a stimulus feature

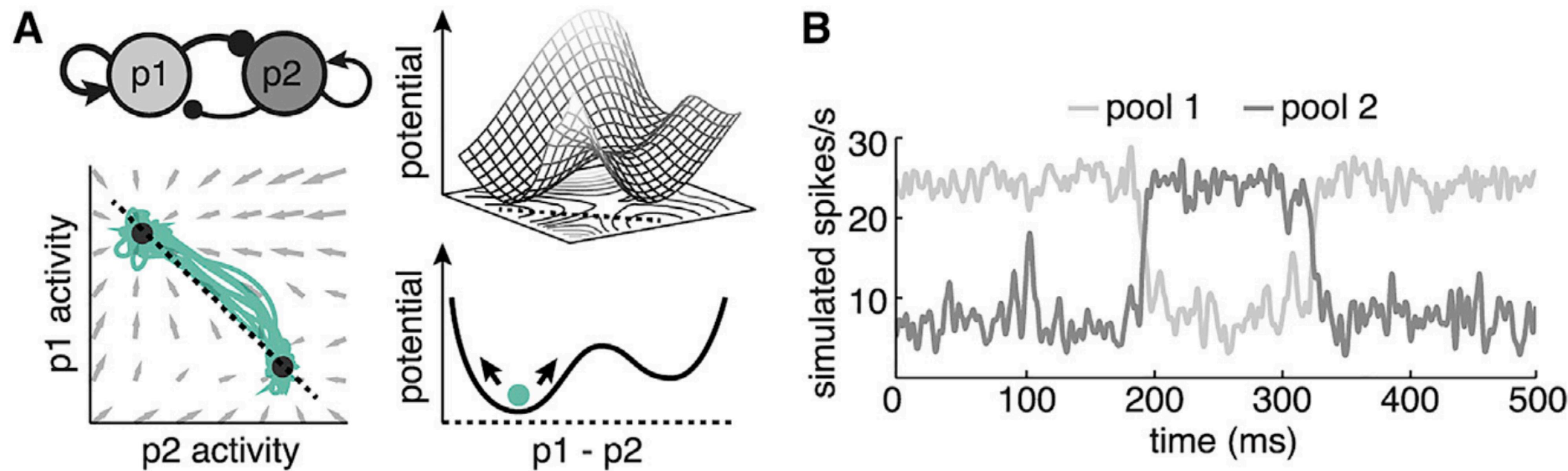
- In which other ways could neural activity encode information?



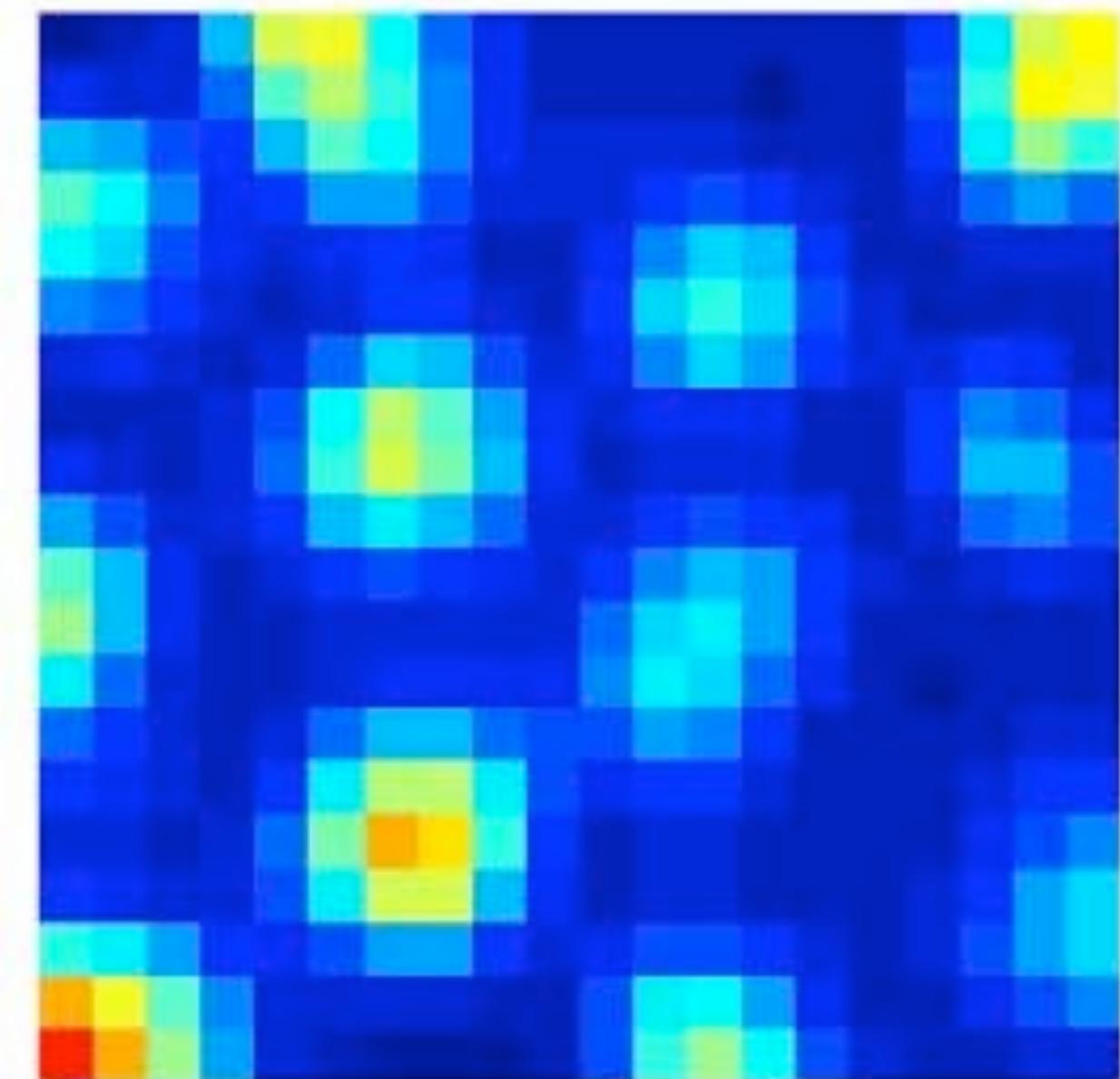
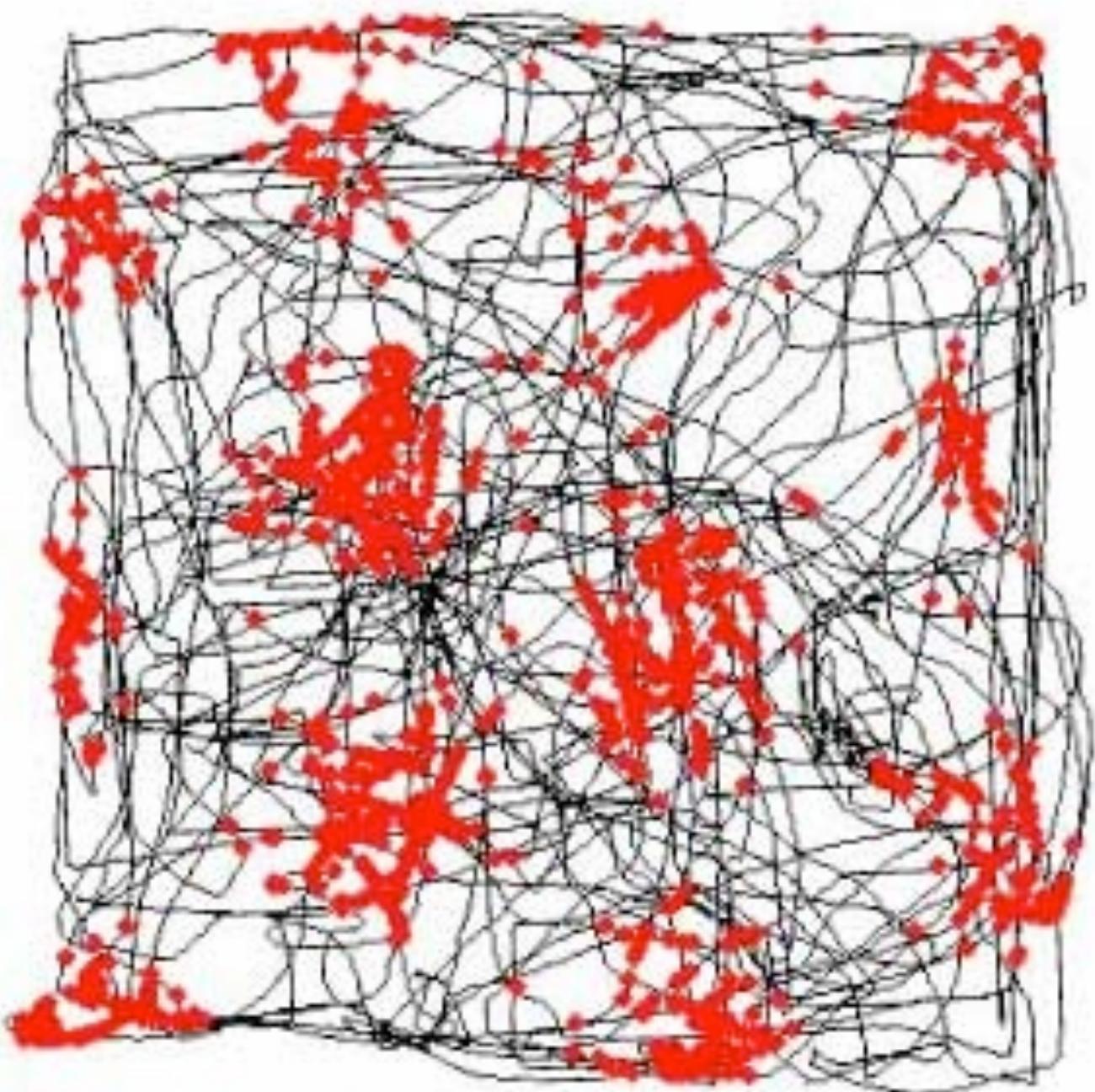
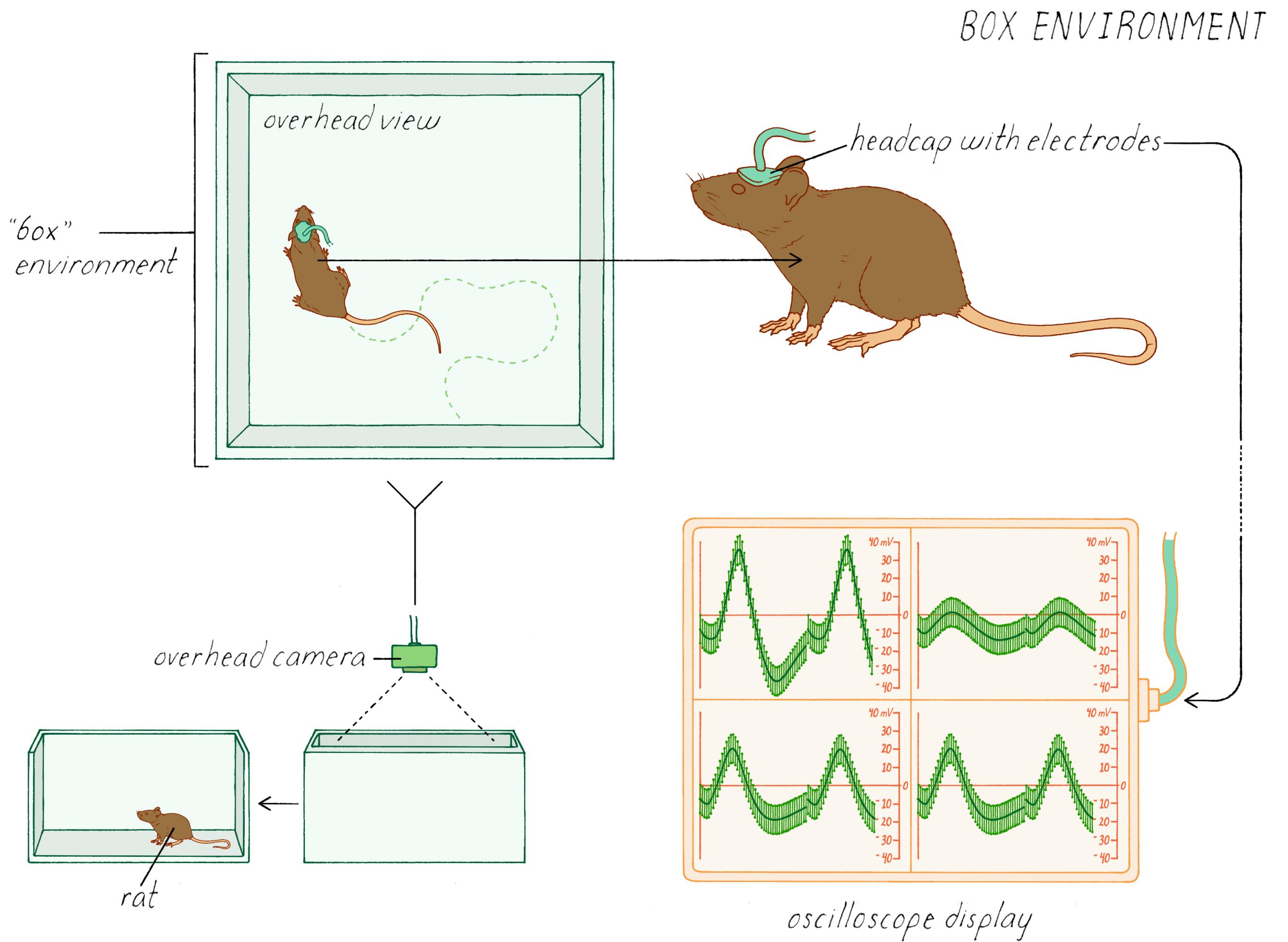
Population encoding

Neurons are not generally selective to a single stimuli:

- neurons span a vector space that multiplexes information (RFs are the exception, not the rule)

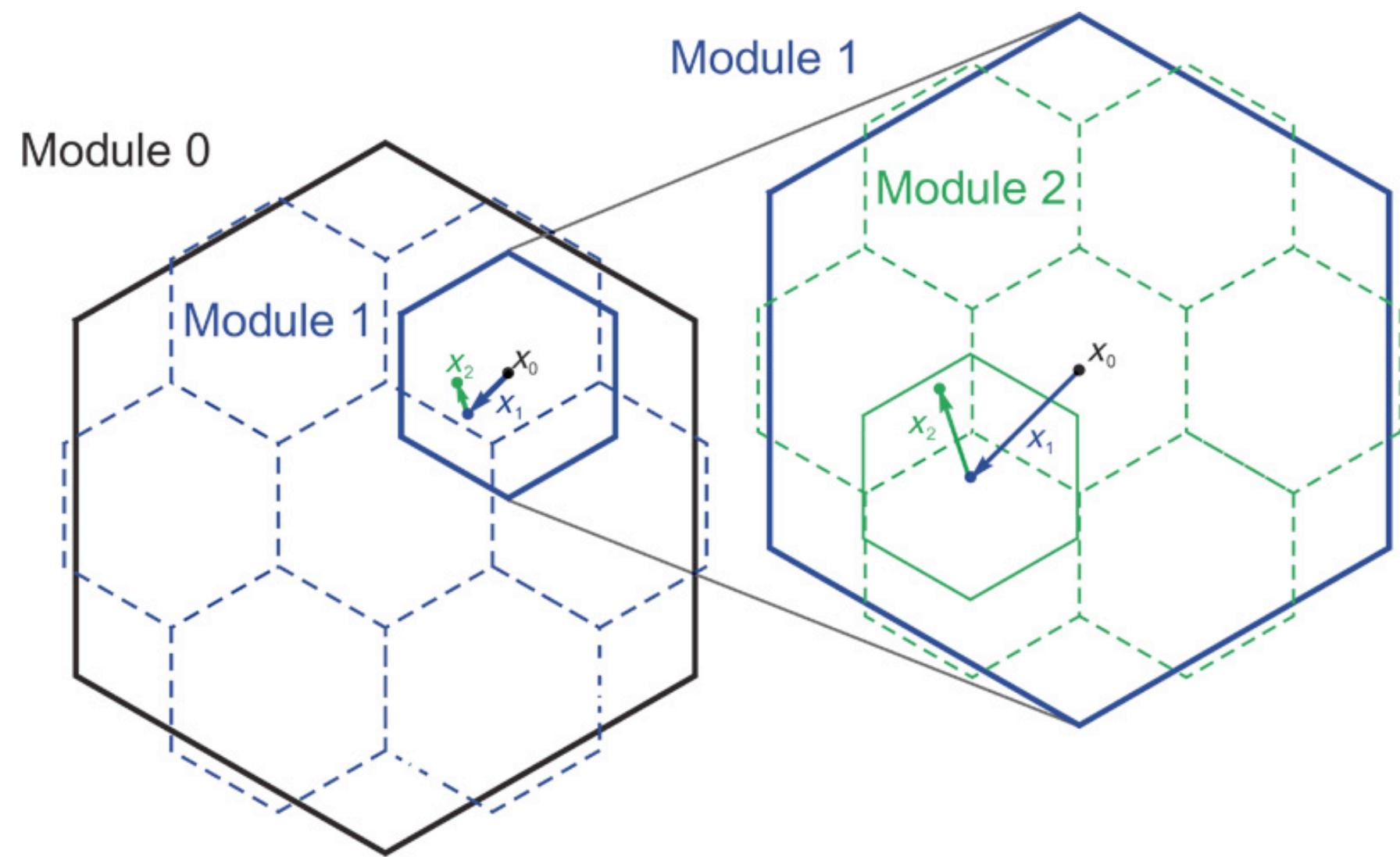


Example: Grid cells

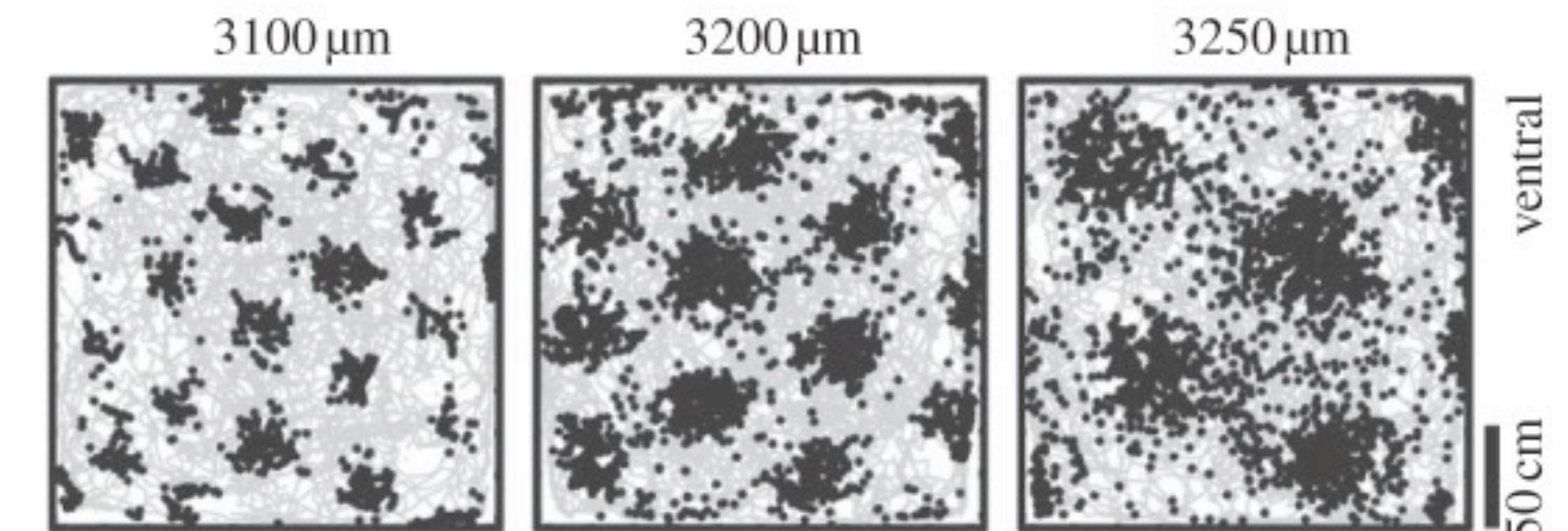
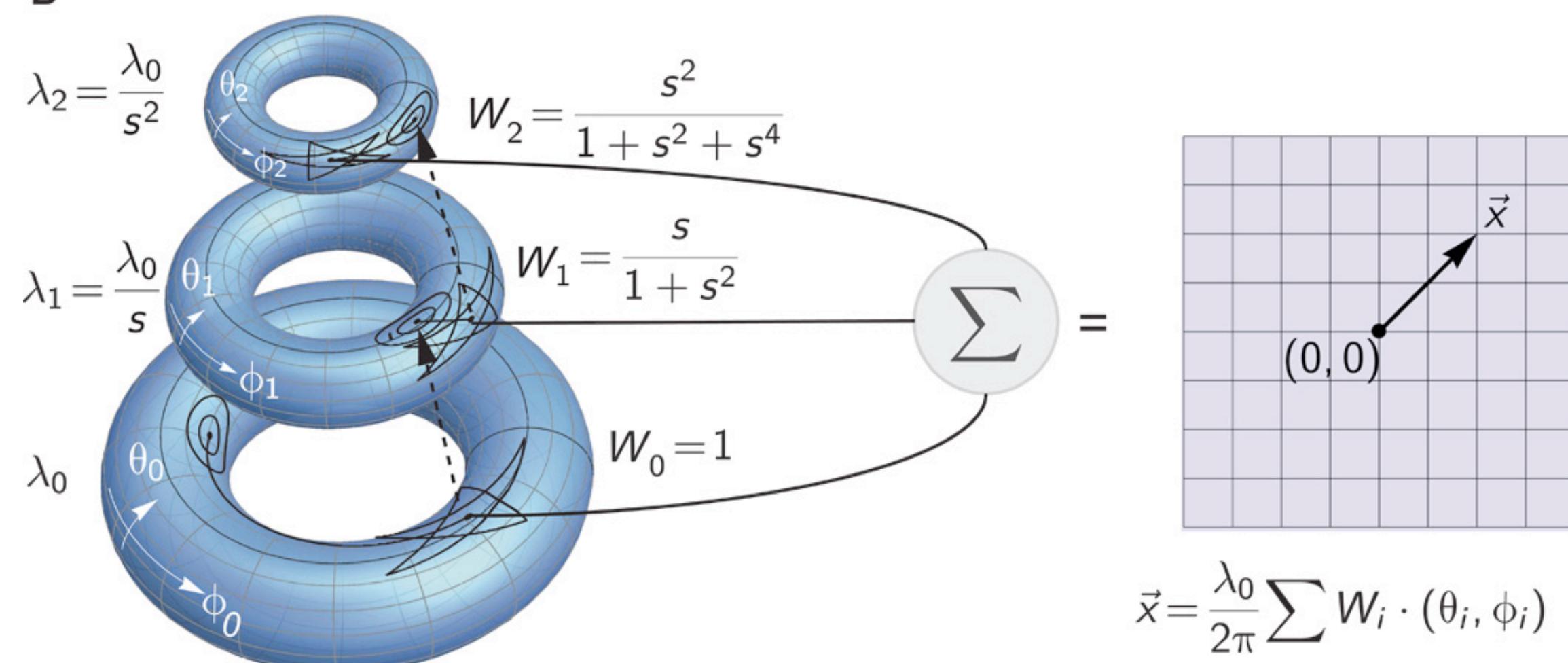


Example: Grid cells

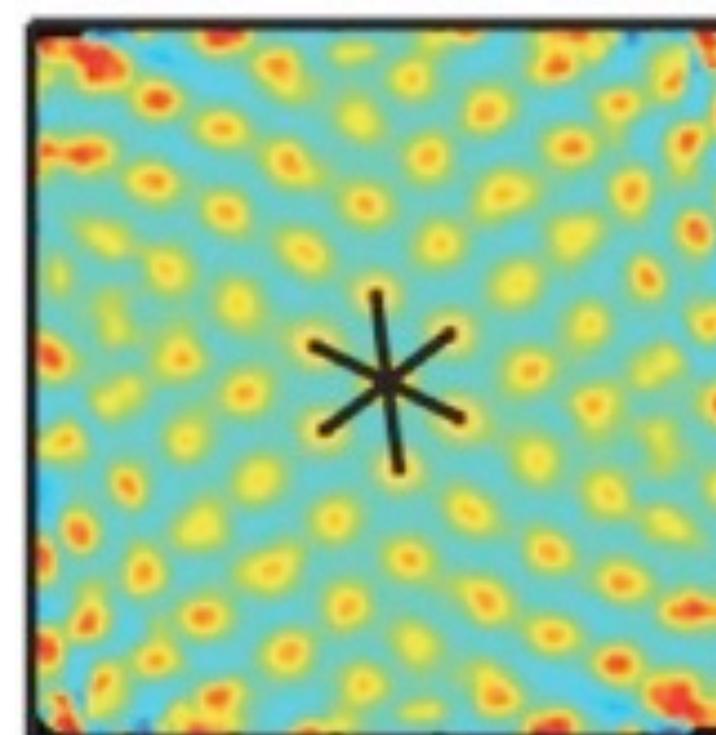
A



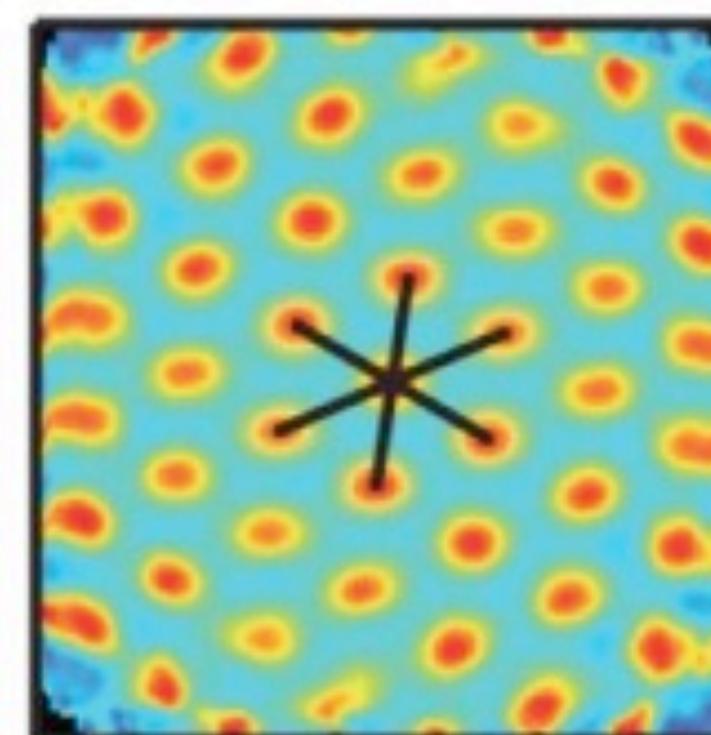
B



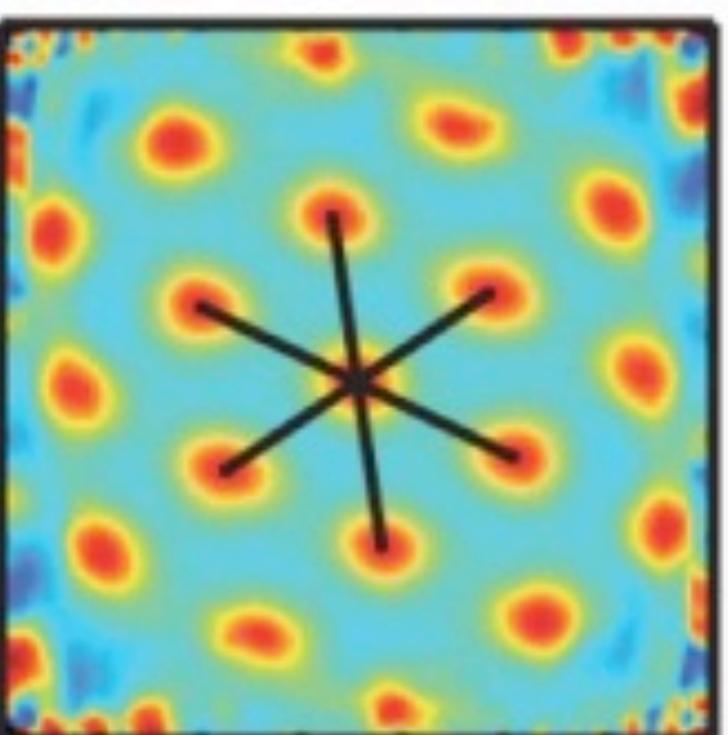
cell: 150610_T3C3



cell: 170610_T3C2



cell: 180610_T3C2



Computational models in Neuroscience research

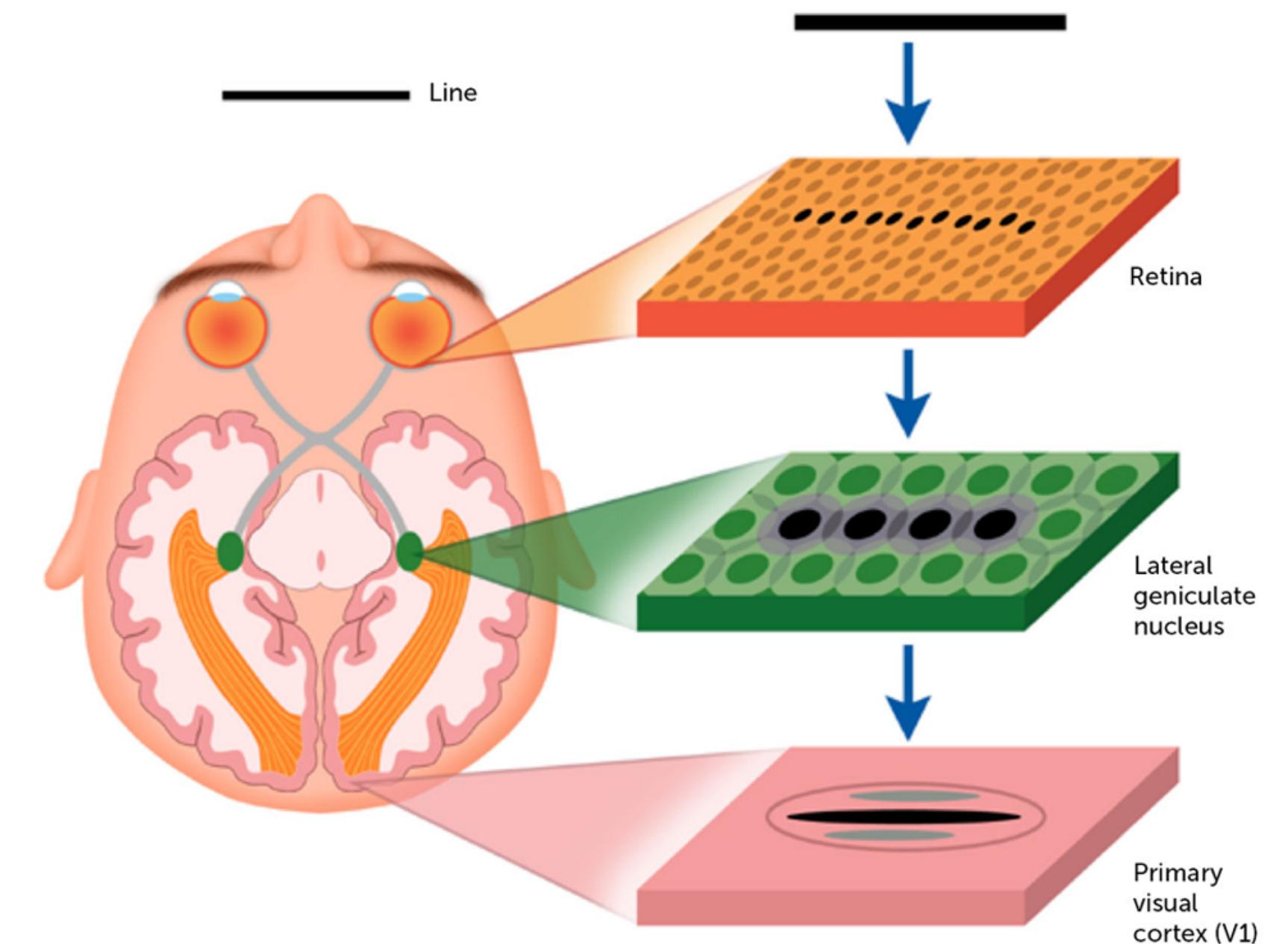
1. What is a representation?
2. Single-unit codes and receptive fields
3. Measuring receptive fields
4. Population codes
- 5. Transformation of information**
6. Conclusions
7. Mind reading

Transformations between RFs

$$RF = \left\{ \begin{matrix} \text{grid 1} \\ \text{grid 2} \\ \text{grid 3} \\ \vdots \\ \text{grid n} \end{matrix} \right\}$$

$$\downarrow$$

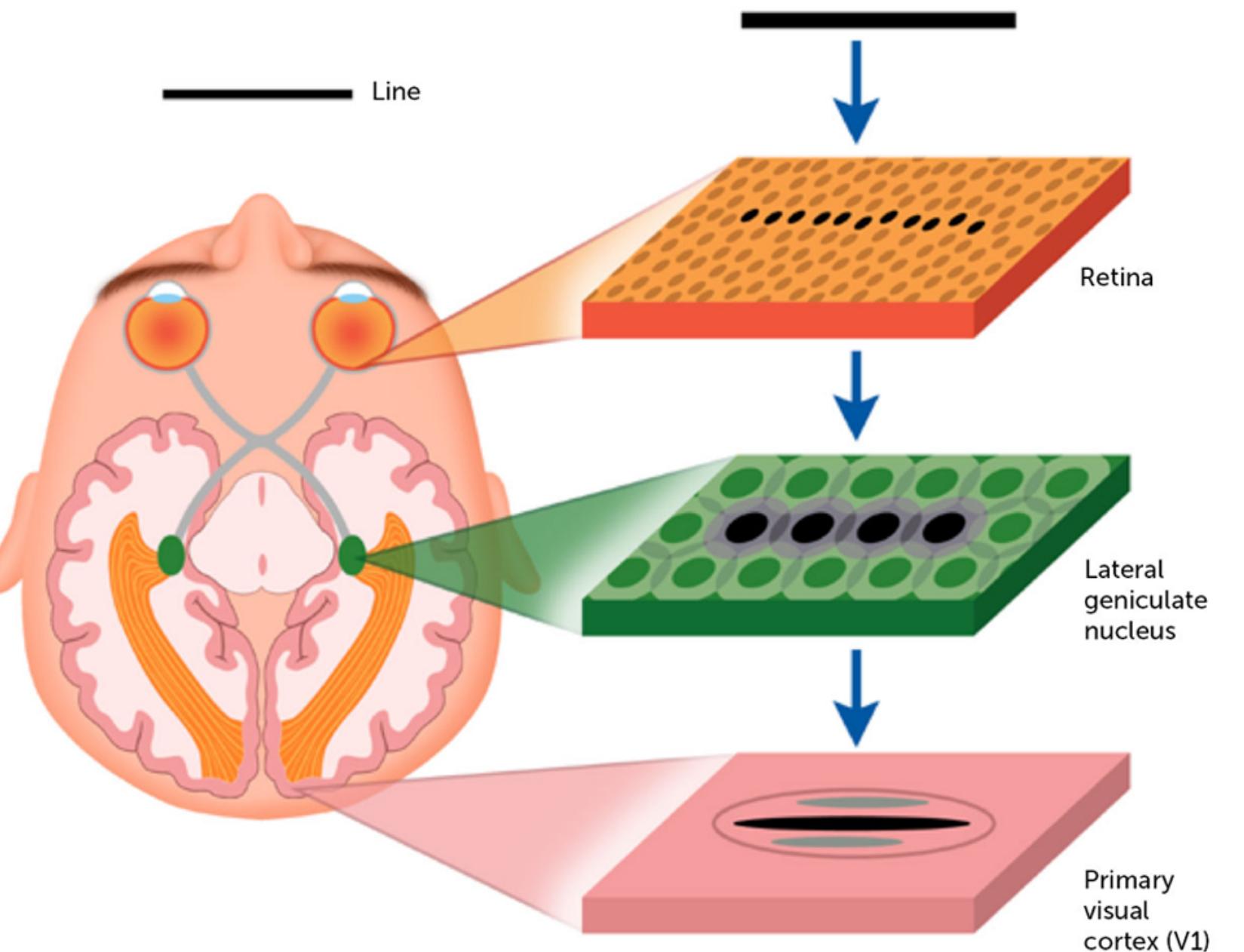
$$RF = \left\{ \begin{matrix} \text{blob 1} \\ \text{blob 2} \\ \text{blob 3} \\ \vdots \\ \text{blob n} \end{matrix} \right\}$$



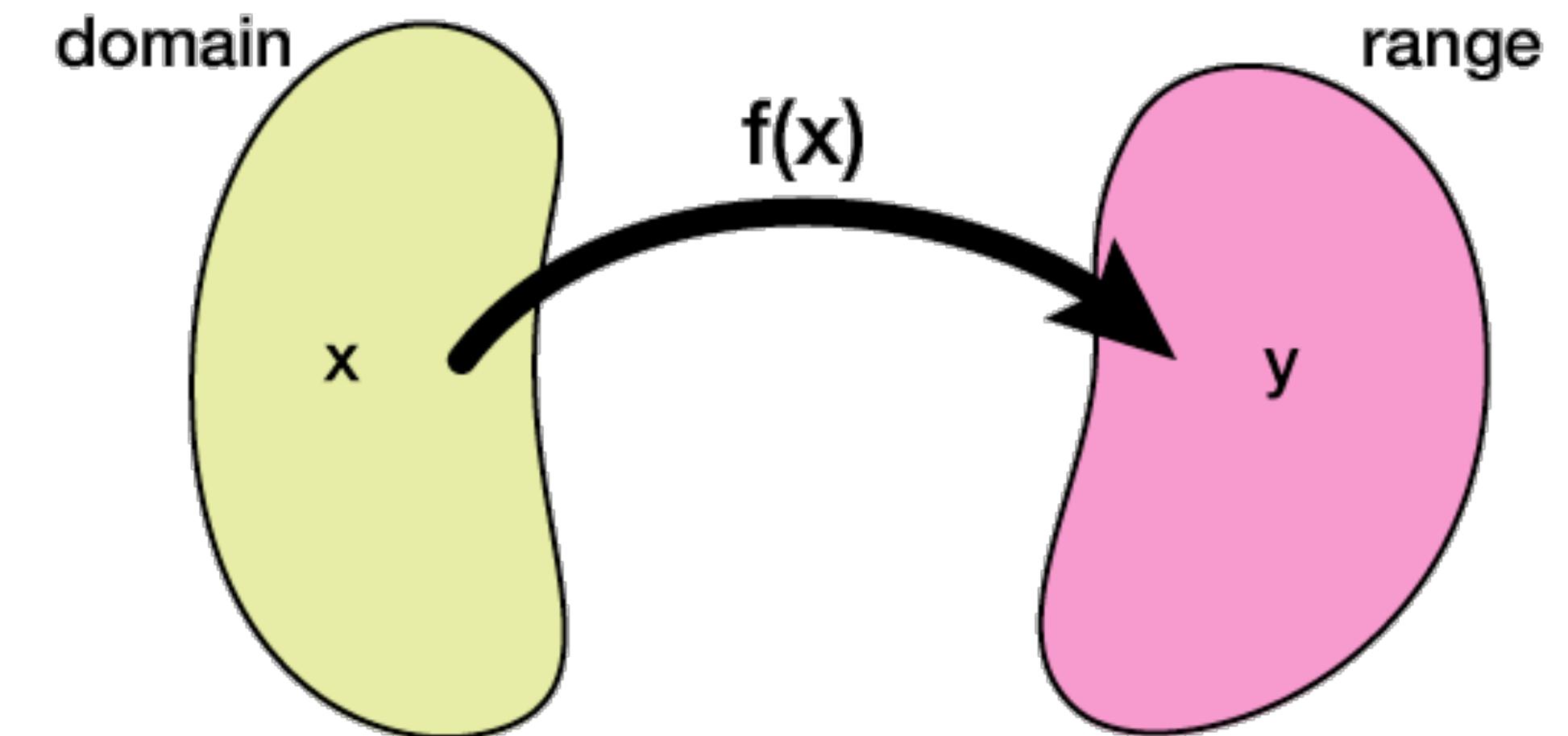
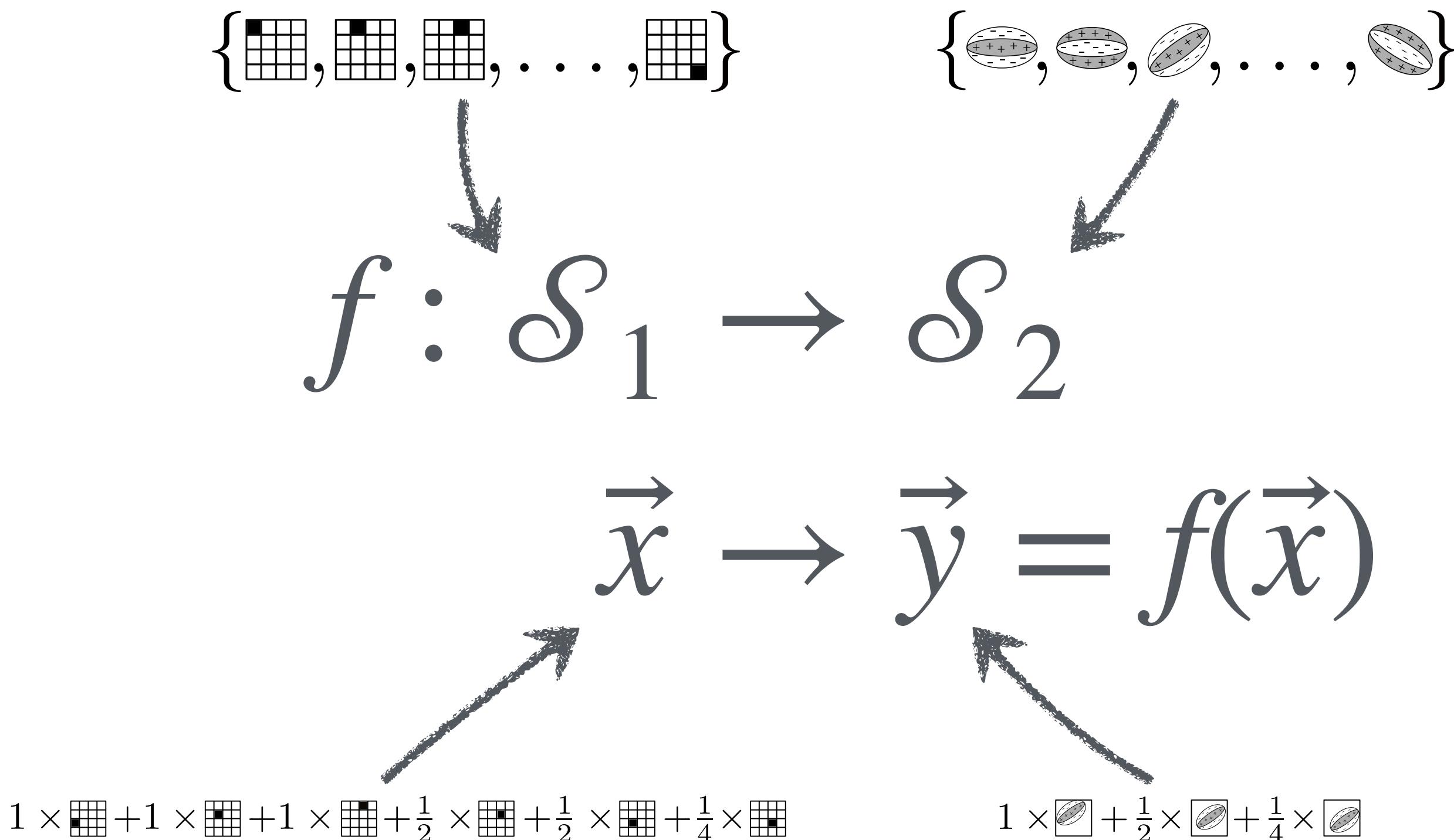
Transformations between RFs

$$\begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} = 1 \times \begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} + 1 \times \begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} + 1 \times \begin{bmatrix} \text{█} \\ \text{█} \\ \text{█} \end{bmatrix} + \frac{1}{2} \times \begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} + \frac{1}{2} \times \begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} + \frac{1}{4} \times \begin{bmatrix} \text{█} \\ \text{█} \\ \text{█} \end{bmatrix}$$

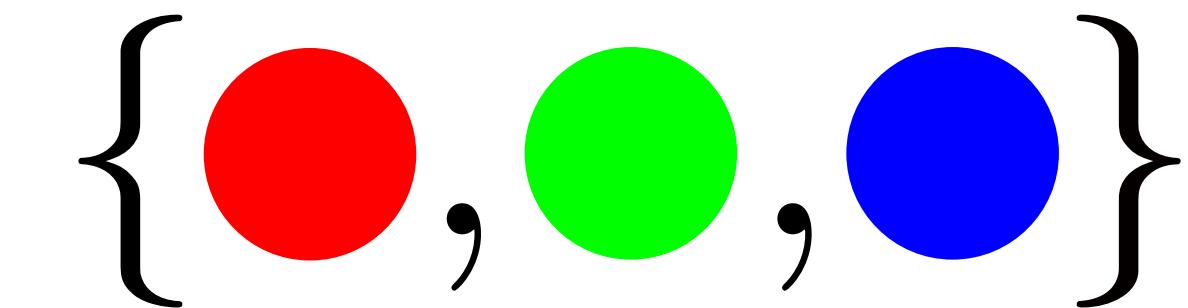
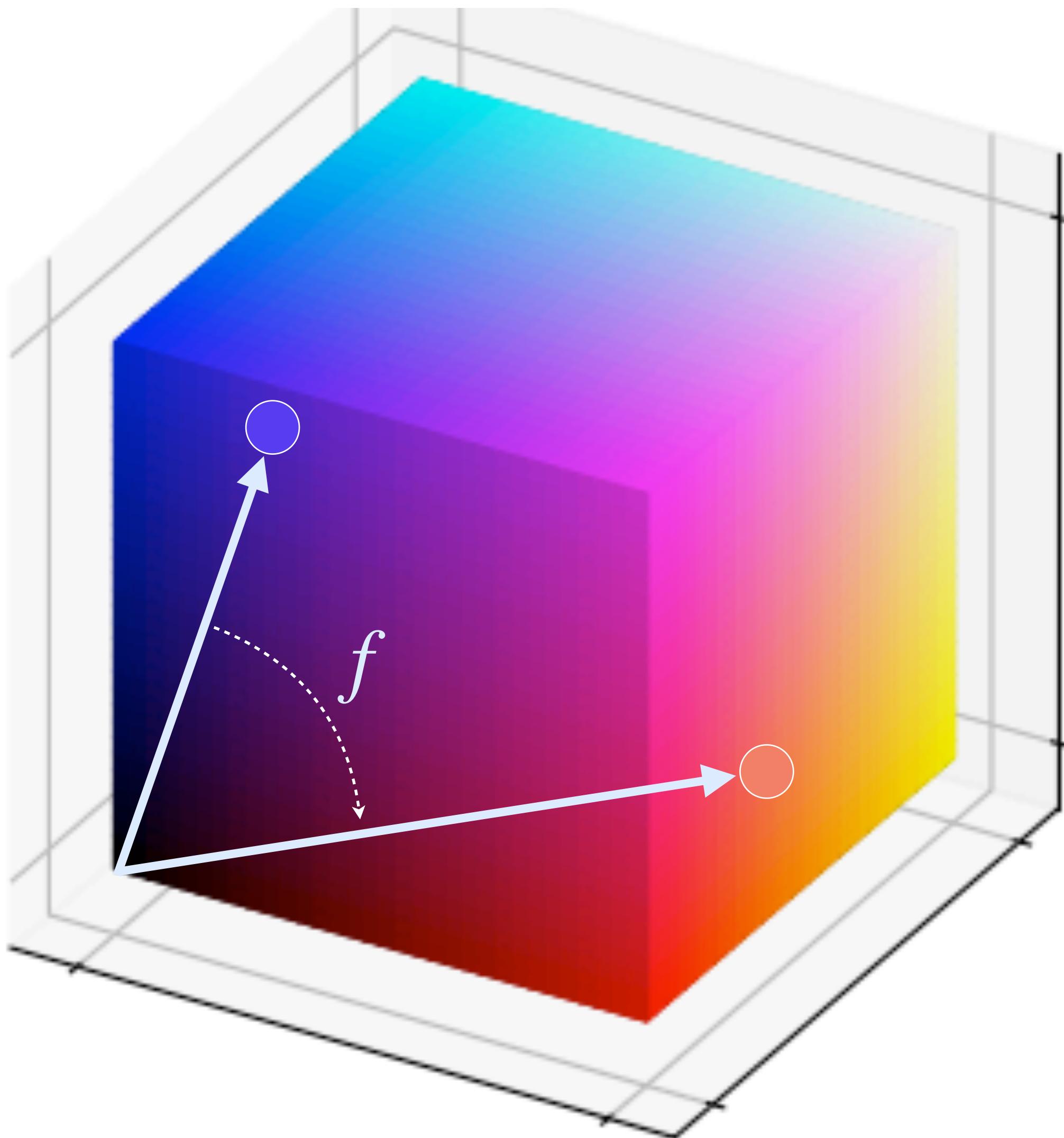
$$\begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} = 1 \times \begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} + \frac{1}{2} \times \begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix} + \frac{1}{4} \times \begin{bmatrix} \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \\ \text{█} & \text{█} & \text{█} \end{bmatrix}$$



Transfer = Mapping between vector spaces



Transformations = mappings between vector spaces



Transformations in multi-dimensional codes

- Rotations in the code vector space modify information
- Pieces of information can be added within the same space (blue + yellow = green)

Computational models in Neuroscience research

1. What is a representation?
2. Single-unit codes and receptive fields
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7. Mind reading

Conclusions

The brain can represent information through:

- Single neuron spikes
 - Spike timing
 - Firing rate
- Population codes
 - Single-unit coding (one-dimensional firing rate)
 - Multi-unit coding (multi-dimensional firing rate)
- ...

Conclusions

Information representation can be modelled using:

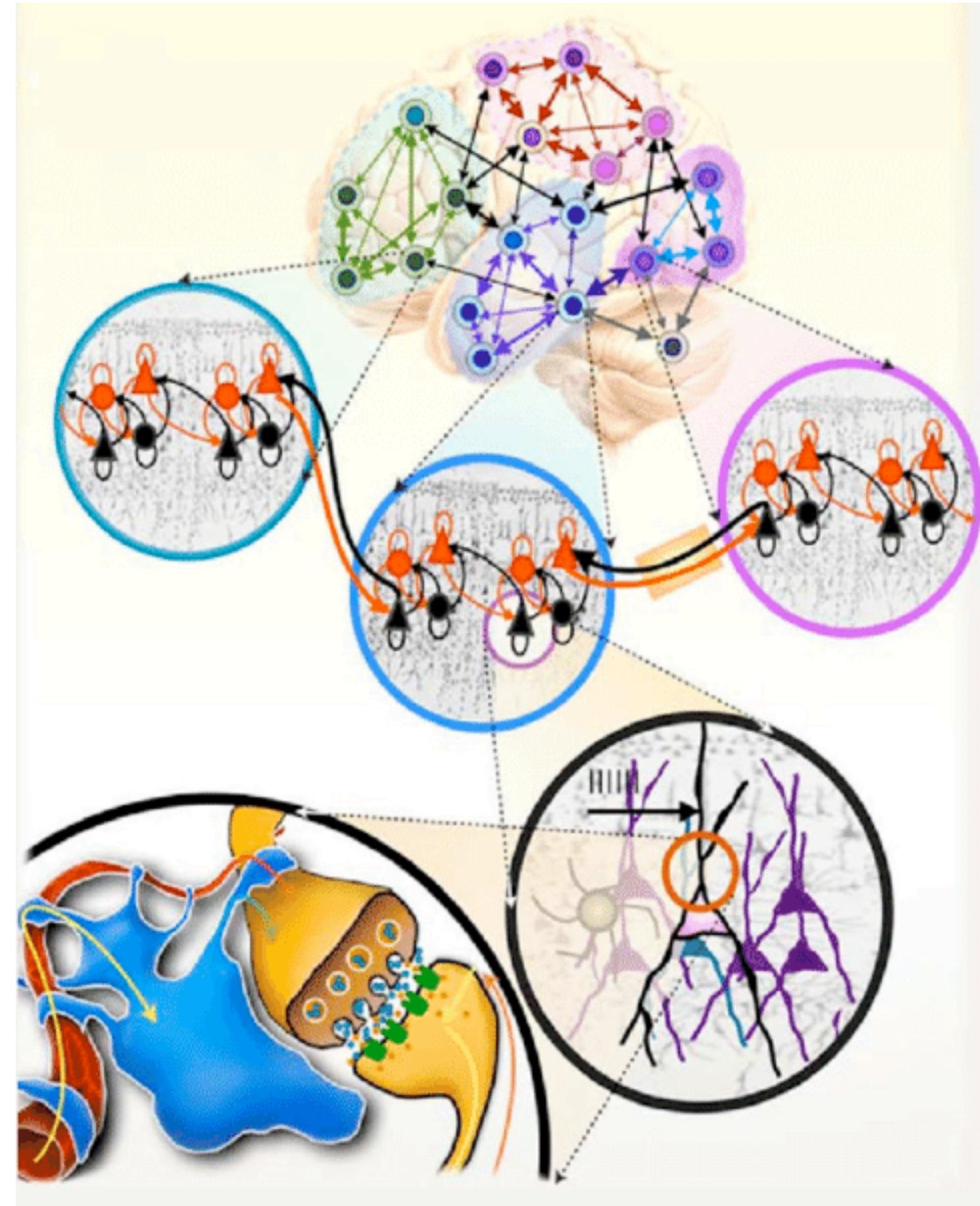
- Parametric receptive fields
- Vector spaces
- ...

Information transfer can be modelled as:

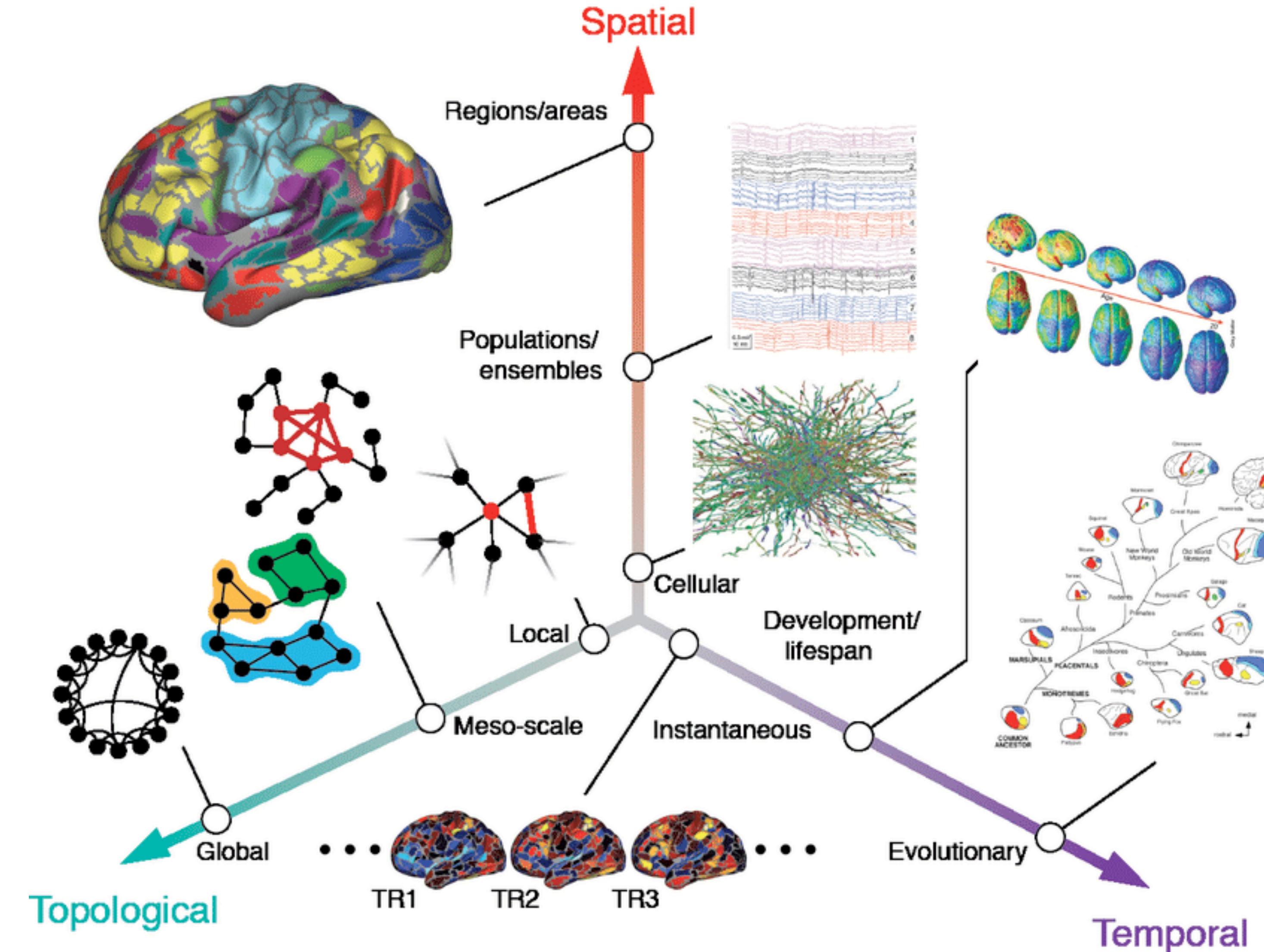
- Mappings between vector spaces (transfer)
- Rotations and other operations within the same vector space (transformation)
- ...

A note for caution: even more levels of analysis

macroscopic



microscopic

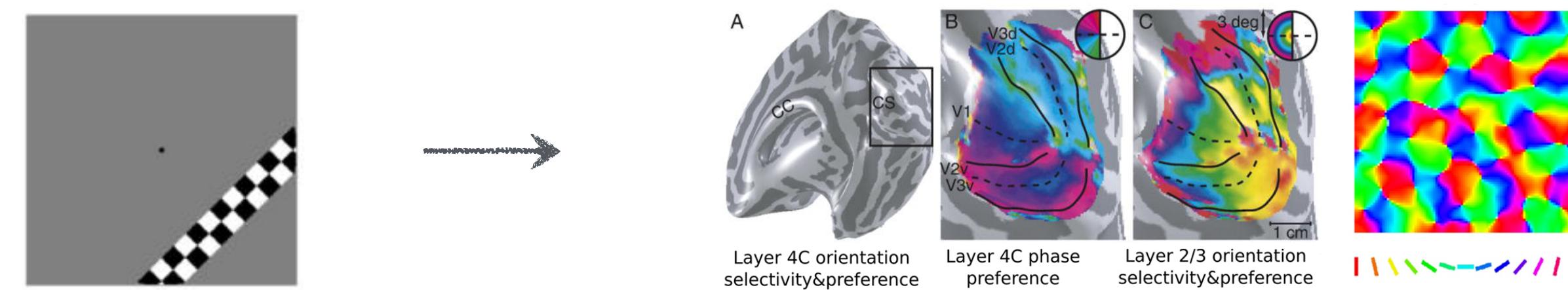


Computational models in Neuroscience research

1. What is a representation?
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- 7. Mind reading**

Encoding and decoding

Encoding: finding where and how stimulus features are represented in the brain

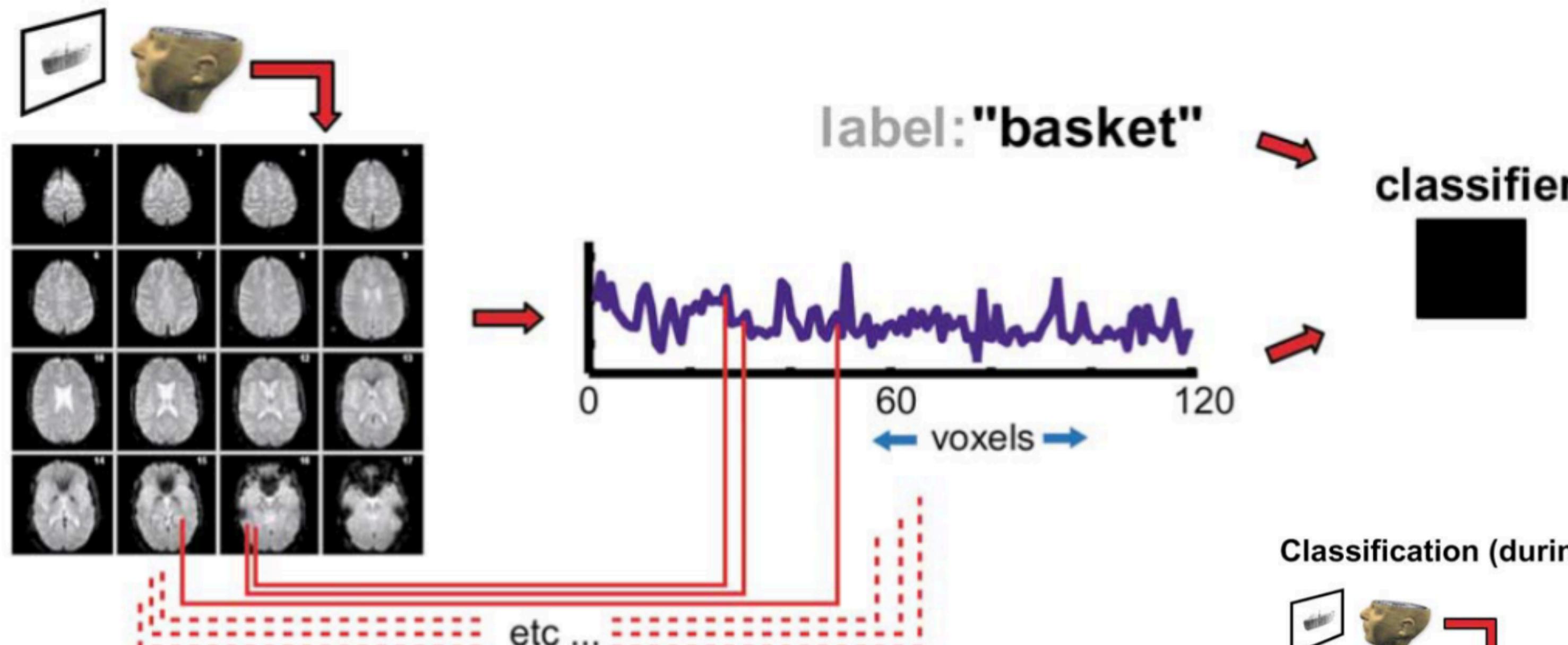


Decoding: infer the information represented in the brain from a measure of its activation

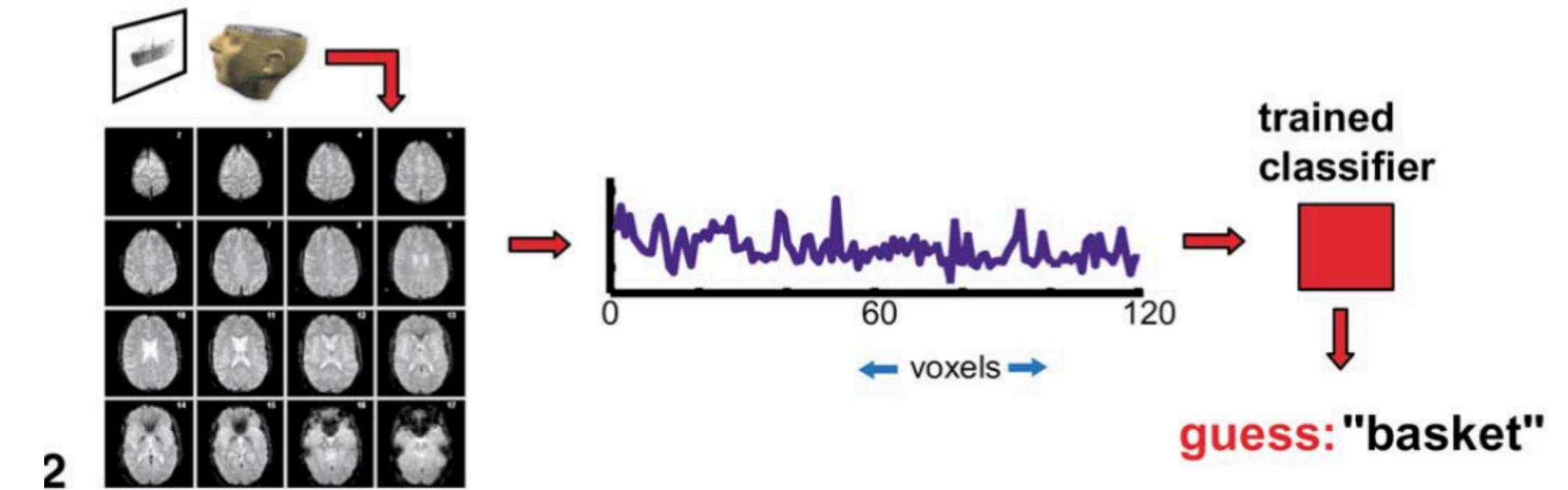


The essentials of decoding

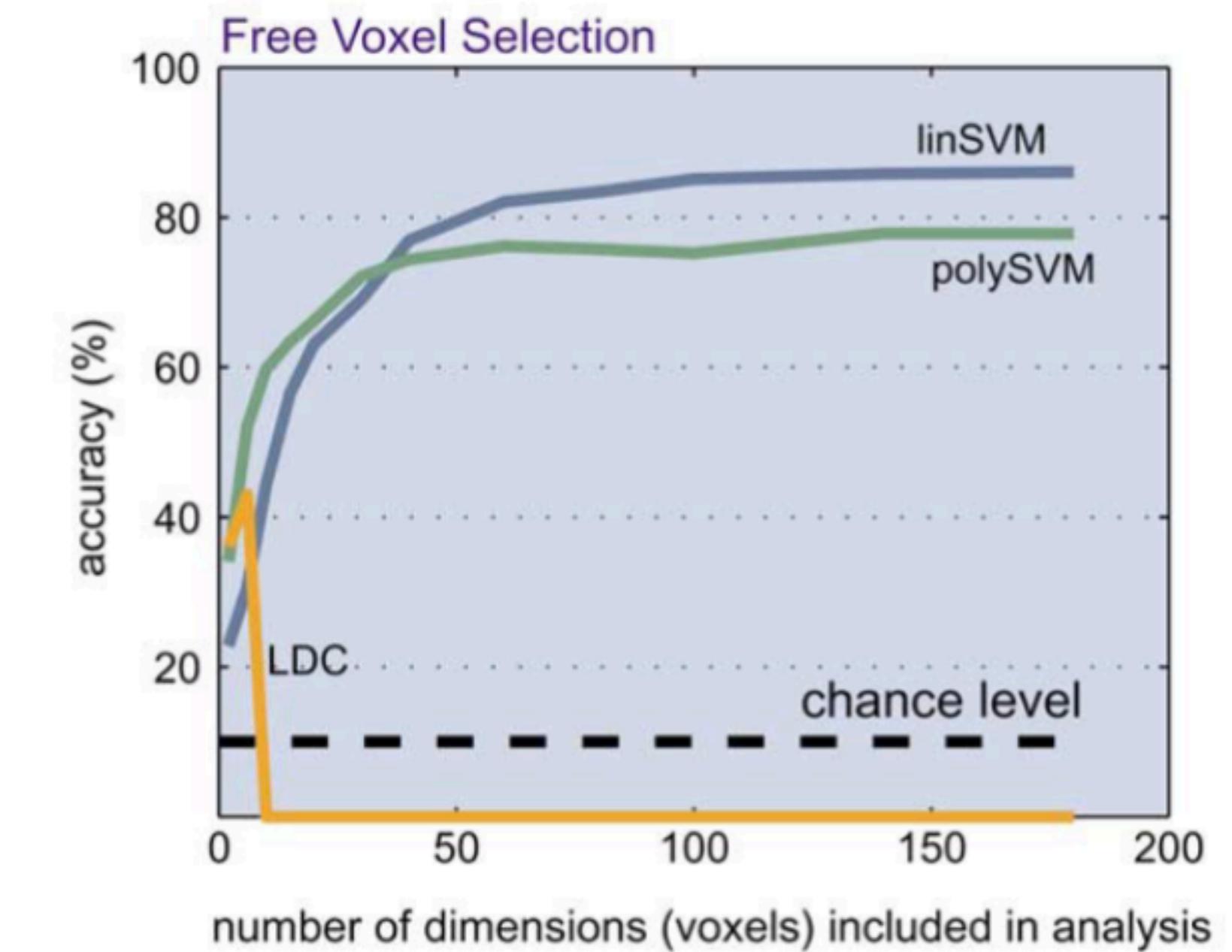
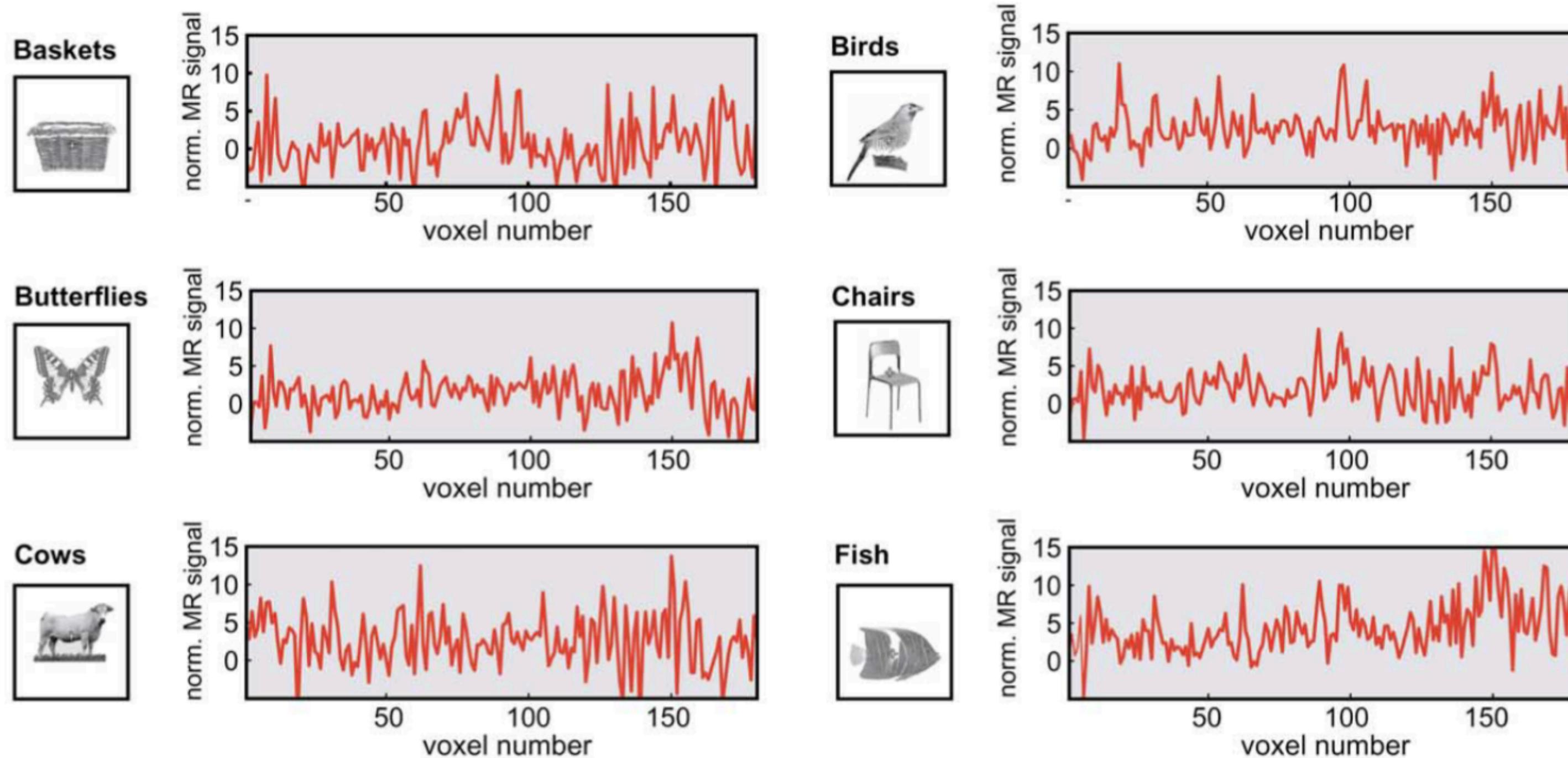
Training



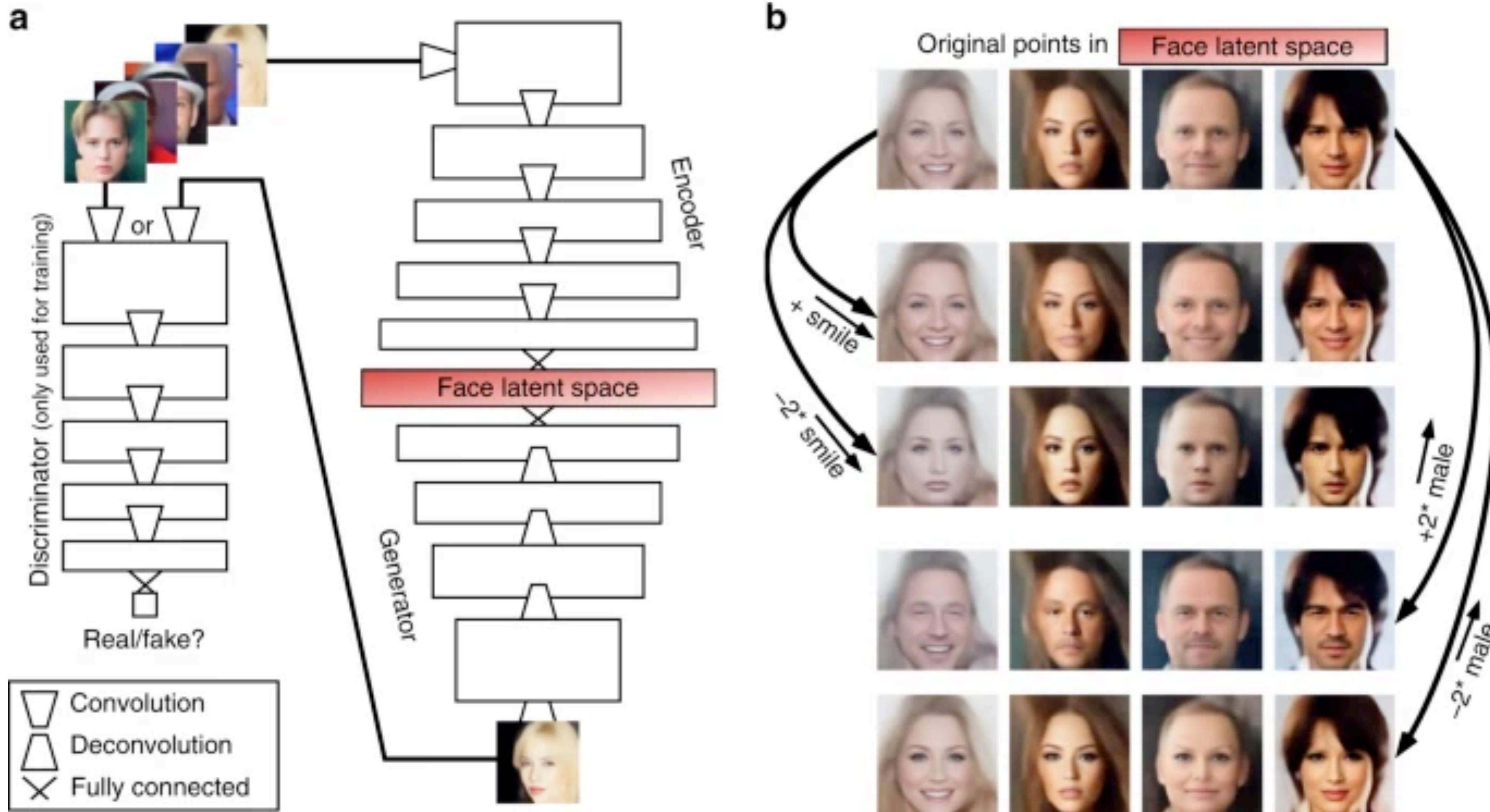
Classification (during a subsequent session)



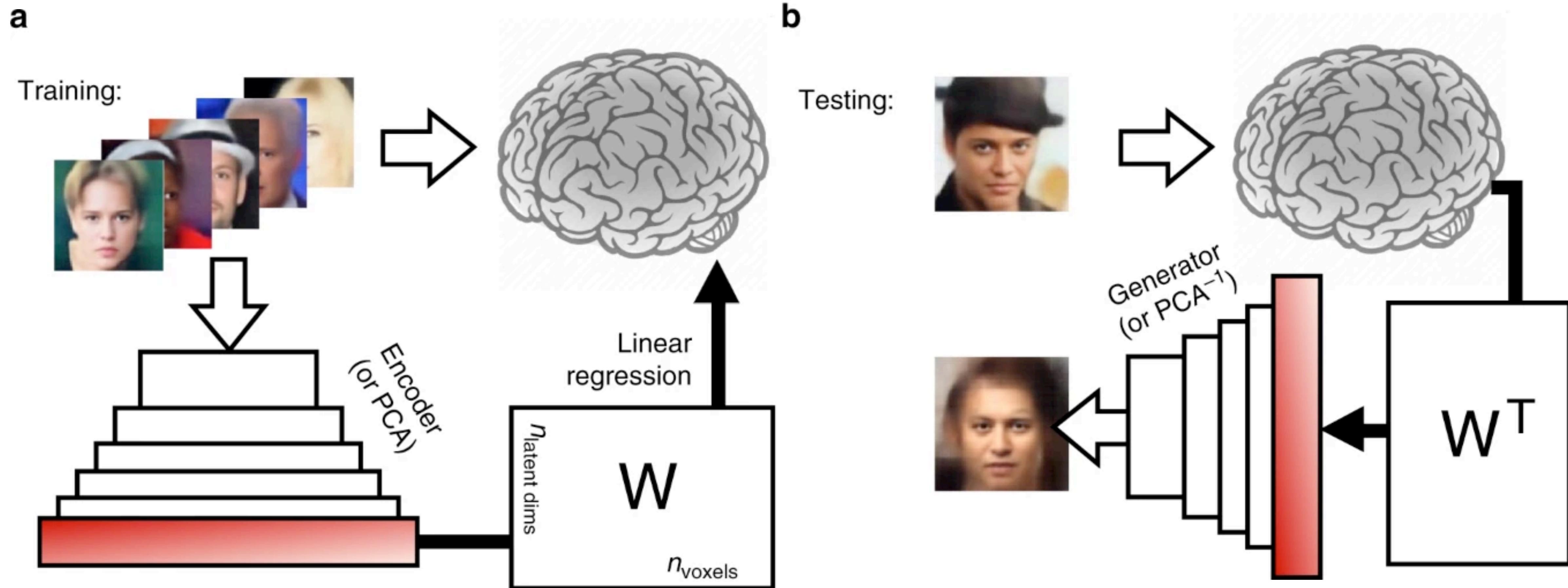
Decoding in 2003



Decoding in 2019

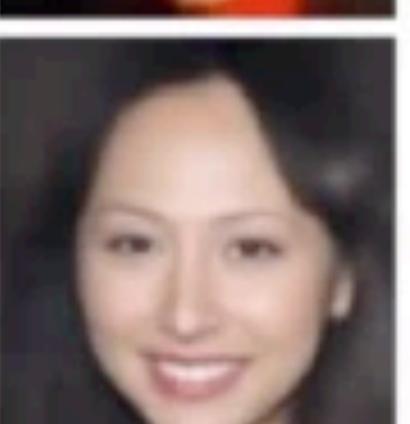
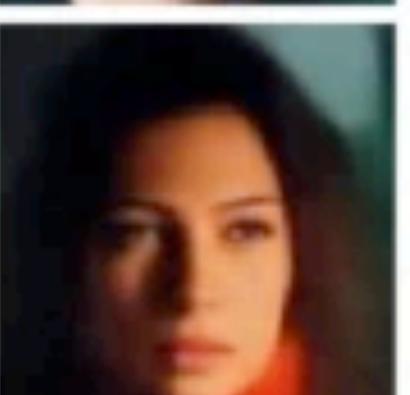
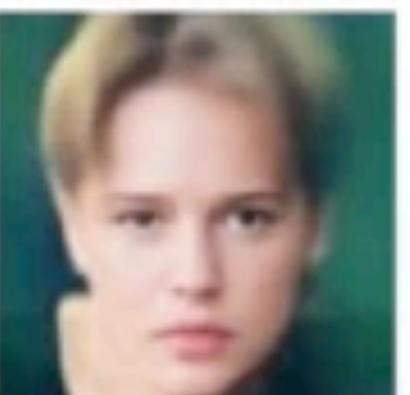


Decoding in 2019

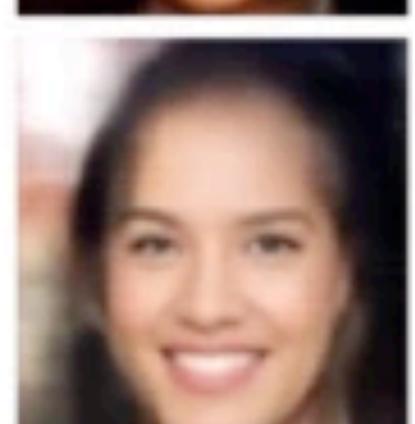
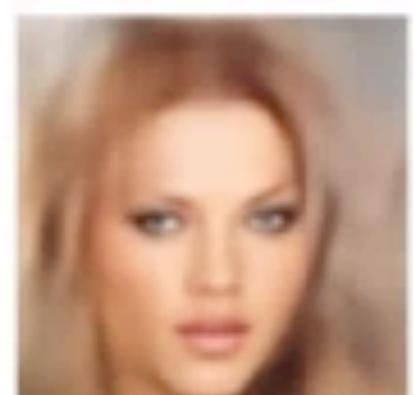


Decoding in 2019

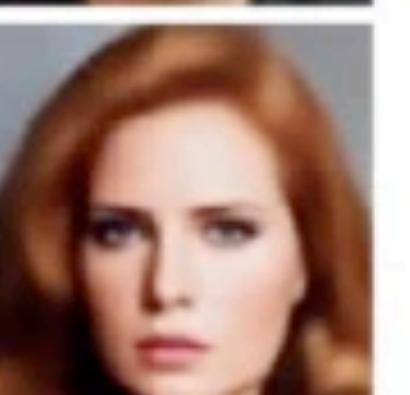
Shown



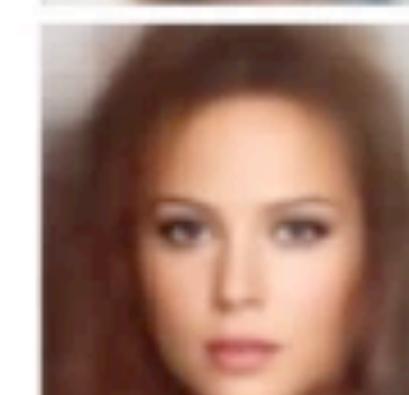
VAE-GAN



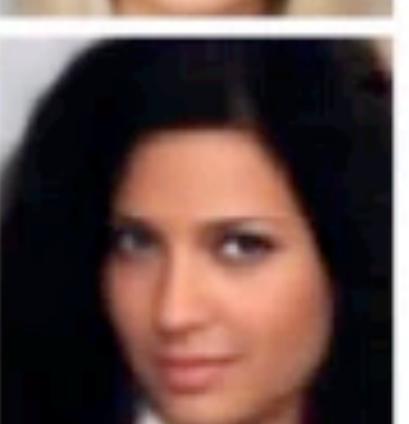
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VAE-GAN



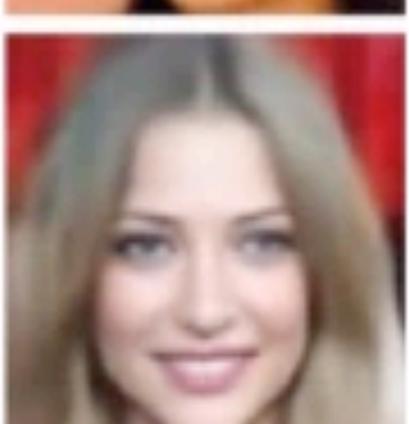
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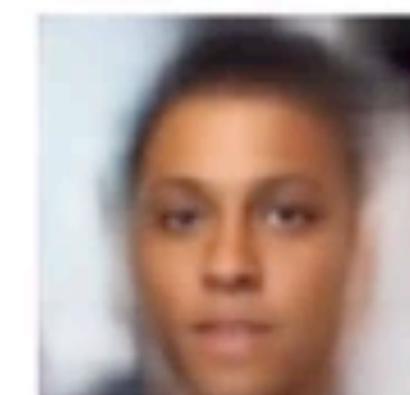
VAE-GAN



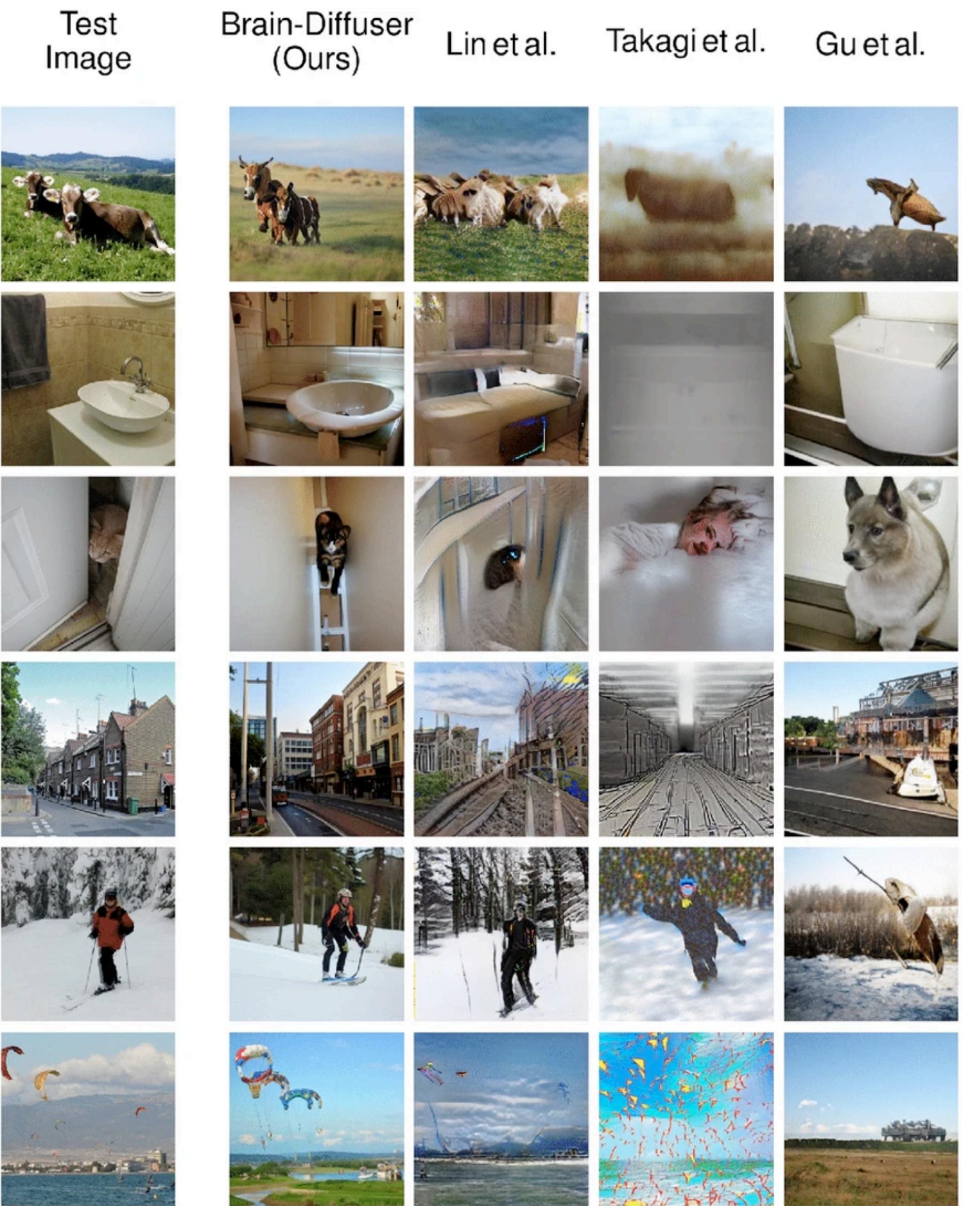
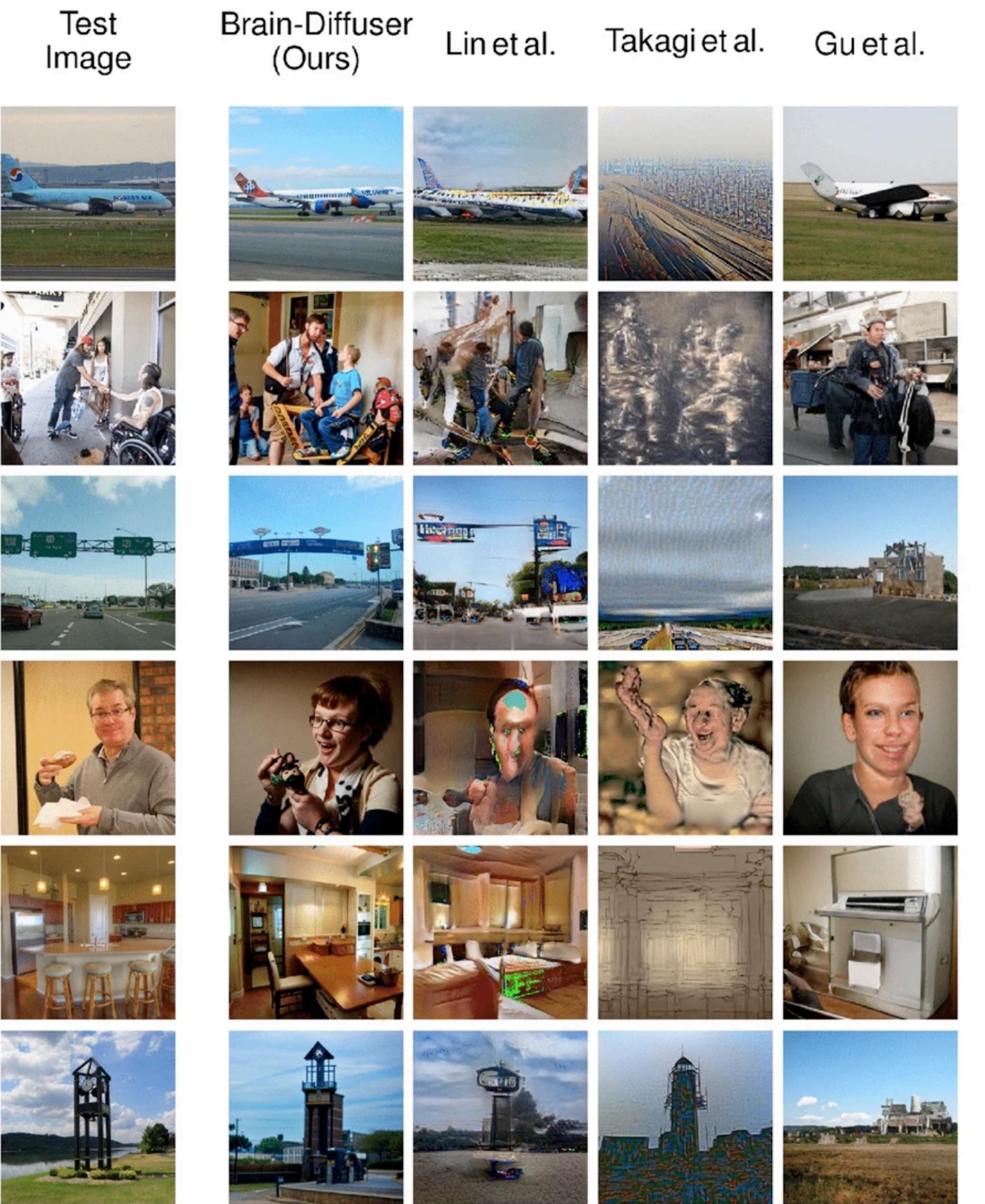
Shown



VAE-GAN



Decoding in 2023



Questions for next session

Questions for next session: Q2.1

A neuron's receptive field describes how it responds to the world. Is this an objective property of the stimulus or a subjective model constructed by the brain? If different organisms have different RFs for the same sensory input, whose model is correct? What does this tell us about the nature of representation?

Questions for next session: Q2.2

Why do some receptive fields, like the orientation selectivity of V1, have finite width? Why don't they, instead, respond selectively only to a single orientation value? What constraints or principles force the brain to use broader tuning?

Questions for next session: Q2.3

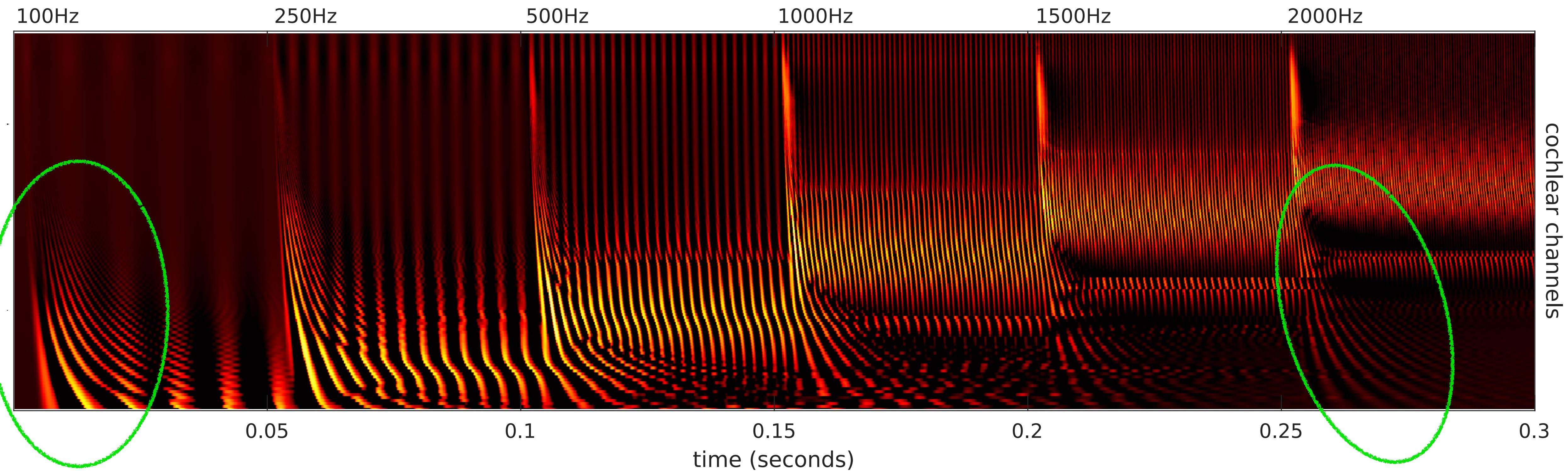
Receptive fields in the epidermis show considerable variability in size and overlap. What might be the normative principles underlying this variability?

Questions for next session: Q2.4

Imagine you're designing a brain from scratch with no evolutionary constraints. Which properties would be desirable in the brain's internal representations of the visual input? For a few of the properties you list, try to formulate specific normative or phenomenological hypotheses on the kind of receptive fields we would expect in the visual pathways. How would you test such hypotheses?

Questions for next session: Q2.5

Take a close look at the beginning of the responses across cochlear frequency channels at the beginning of each tone. What are the first valleys in the response pattern encoding? Why do lower frequencies show longer delays and extended responses compared to higher frequencies?



Questions for next session: Q2.6

Imagine a neural population represents a stimulus feature (e.g., on-off RF responses in the visual thalamus) in its activity pattern but this information is not transmitted to any downstream area. What can you say about whether this information is represented in the brain?

Questions for next session: Q2.7

Different measurement techniques (single-unit recordings, fMRI, EEG) reveal different aspects of neural representation. How does our choice of measurement tool shape our theories about how the brain represents information? What biases might this introduce?

Questions for next session: Q2.8

Modern techniques can reconstruct faces or images from brain activity. What does successful decoding prove or help us understand about the representation?

All course materials:

github.com/qtabs/compneuro4cogneuros