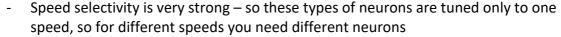
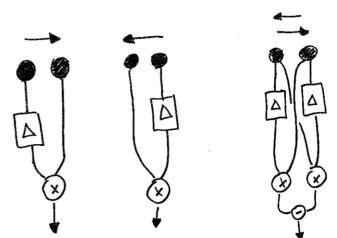
## Computational Vision – Lecture 9

## **Visual Motion: 1D**

#### Reichardt detector:

- Hypothetical neural circuits that can track motion
- A cell (x) will get two inputs, one from eye
   A and one from eye B
- Input from A is delayed and if the cell responds to that then it will fire
  - → Cells respond only to one of the three example cases and then generates the motion perception
  - → Can also combine reichardt detectors to higher level hierarchies





## New way of representation of receptive field of cell

- → takes into account the time aspect of the responses of the cell
  - Logic to take this approach, because cells fire over longer timespan and patterns (e.g. biphasic responses) and not static
  - Instead, dynamic and individual responses of cells

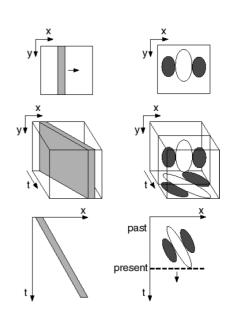
#### Space time representation $\rightarrow$ on the right

#### Top left situation:

- moving bar across field → could be shown as 3D representation (middle) = space time cube, but complex to draw for more complex stimuli
- alternative: represent the moving bar in an x/t diagram →
   slope of the line gives you the speed of the stimulus (lower)

## Top right situation:

- representation of receptive fields in the same scheme
- ON / OFF responses not static anymore, but dependent on the actual response → so e.g. above baseline = ON response, below baseline = OFF response
  - → receptive field shape changes with time!
  - → present line and receptive field above shows us that If we are at the present line and see a response at a certain location, then this will be due to a stimulation that has occurred in the timespan of the space receptive field. I.e. a response at the point of the arrow at present will be due to an ON response in the column of this x value up to the last point in the past.

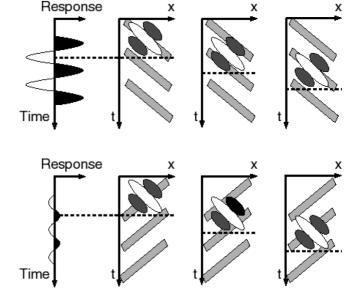


## Upper image:

- First peak: bar matches exactly the ON region of the cell
- Negative response after: bar does not match the ON region, so cell will not respond

## Lower image:

- Receptive field is also oriented in space time and that means that it is orientation selective
- That means that the same setup with bars that are in opposite orientation, the responses will be much smaller.

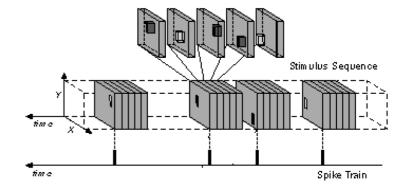


## Measuring space-time receptive fields with reverse correlation

Time consuming, painful method to measure: for each location and each timepoint, measure response

## BUT other method more efficient : **Reverse** correlation

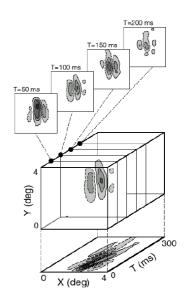
- Want to record from simple cell, whose orientation you already know
- Present to cell a movie of a bar at different locations and different polarities (on/off)
- Then record spike train of the neurons and analyse



Analysis: Creation of spike recordings and you measure what was on the screen 5 ms before the spike  $\rightarrow$  repeat for all spikes

- Then average all the images → result: on average, 5 ms before the image, there was an e.g. white bar → develop tendency, that everytime there is a white bar, you see a spike
- Repeat with 10 ms, 15 ms, 20 ms, 25 ms, etc. etc.
- Then you get series of pictures that represent the receptive field at different times
  - → represent the receptive field

! if you go back **too far back**, then the information tendencies will be **uniformly grey** because of the averaging process



! at **point 0 ms or 1 ms before**, the picture will also be completely **grey**, because it has no relevance for the response.

! only works for linear receptive fields

Video 1: Cat visual system → ON center cell (ganglion cell in retina / LGN) and non-direction selective

Video 2: biphasic response of a simple cell that is selective for vertical orientation  $\rightarrow$  not changing much over time

Video 3: simple cell: ON/OFF regions that also move over time and oriented in space time and direction selective!

Video 4: stimulation of complex cells receptive field with reverse correlation technique only with a light emitting bar. If you were to measure with light/dark bars at the same time then they would cancel out → not so useful for complex cells because of nonlinear behavior and it cancels out!

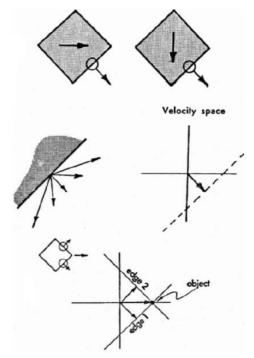
#### **Visual Motion: 2D**

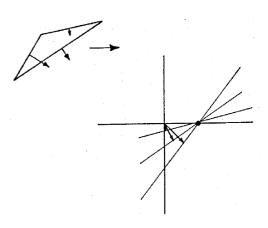
## **The Aperture Problem:**

- E.g. square that is moving from left to right and one cell's receptive field is the circle. This receptive field will perceive the edge moving down
- If the square is moving down, it will see the same.
- That means that there are many possible movements of the square that all result in exactly the same expression in the receptive field → PROBLEM
- → visual system can still solve this problem via:

## Intersection of constraints:

- combine outputs of two cells and then combine more and more to relieve the ambiguity (picture on the right)
- vector set creates complete impression
- line of constraints = line at which all these possible motion vectors end → edge 1 and edge 2
- → combining the lines of constraints at the intersection tells you the only possible motion of the square
  - length of the vector = speed of the motion
  - not just vector addition! See in the example on the right  $\rightarrow$

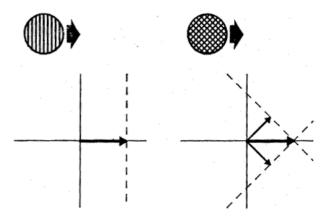




# What happens if you look at more complex stimuli such as a grating?

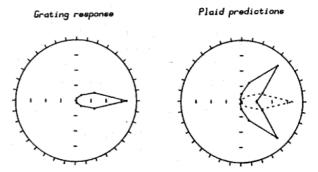
Experiments performed and recorded responses in area MT

- most cells in MT are selective for orientation
- → also for grids, you only perceive a fixed grid moving to the right and not the single lines



Polar plot that shows one cell's response first to grating and then to plaid

- angle = represents direction of motion
  - here cell responsive to orientation of motion and orientation in space
  - e.g. 0 deg. Grating is vertical and moving from left to right
  - 180 deg. Grating vertical and moving from right to left



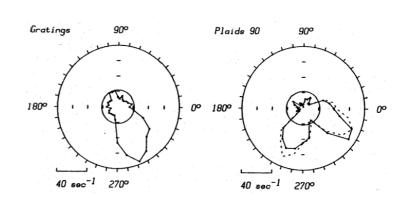
- o 90 deg. Grating horizontal and moving from to bottom to top? vice versa?
- → revisit the polar plots and readouts.. ( check out the paper of the author Movshon)
  - for a grating, the polar plot will just have a circle
  - an orientation selective cell will have the grating response replicated on both side

#### Plaid response:

- picture: response of plaid moving to the right
  - components are not vertical and that's why you see two directions on polar plot
  - o but together the perception is that the plaid is moving to the right

#### Responses of a V1 cell:

- left: cell likes grating that moves to one direction
- right: same cell stimulated with a plaid and responds with a different polar plot
- → like one component going in one motion

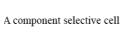


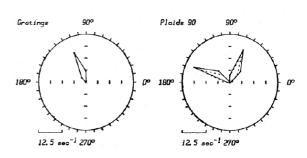
Same experiment done in MT:

#### Found cells:

1) component selective cell  $\rightarrow$  like cell responses above

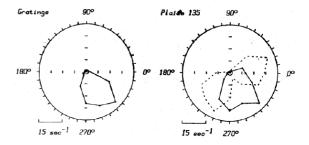
2) pattern selective cells →
response is like in the
grating for the plaid! cell
has combined the motion of
the two components to
compute the overall
direction pattern just like
you perceive it like a human
observer





→ shows that this pattern computation is not performed in primary visual cortex, but actually only in MT

A pattern selective cell



Summary of experiment:

Population analysis:

- first stimulation with grating, then stimulation with plaid

Q: How much can you predict response to the plaid when you know the response to the grating?

Answer: in V1, the correlation is much higher towards the component cell model than to pattern model

In MT, correlation more spread → must have both types of cells

