

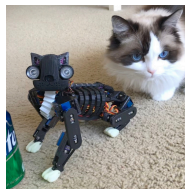
# Binocular Stereo Vision

Neural processing of stereo disparity

Marr-Poggio-Grimson

multi-resolution stereo algorithm

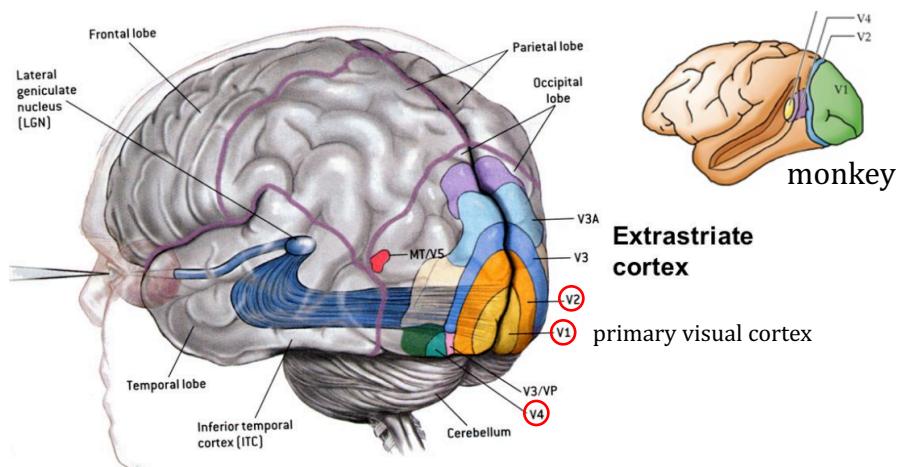
Stereo vision applications



**CS332 Visual Processing**

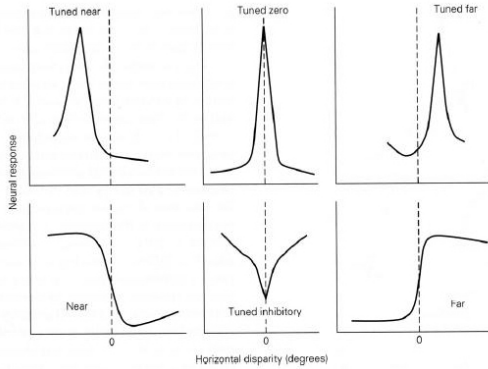
Department of Computer Science  
Wellesley College

## Neural processing of stereo disparity



1-2

## Neural mechanisms for stereo processing



From G. Poggio & others:

- neural recordings from monkey (area V1)
- viewing random-dot stereograms

**zero disparity:** at fixation distance

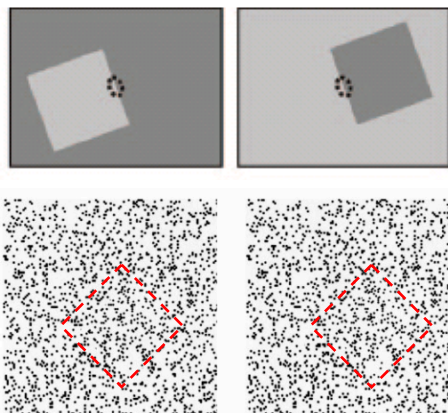
**near:** in front of fixation distance

**far:** behind fixation distance

- (some) simple & complex cells in **area V1** are *selective for stereo disparity*
  - neurons with large receptive fields are selective for a larger range of disparity
- ... but the stereo correspondence problem is **not solved** in V1!!

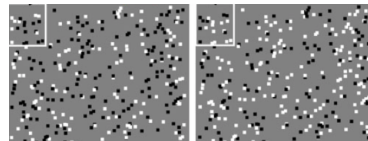
1-3

## Selectivity for *stereo boundaries* in V2



Von der Heydt & colleagues:

Some V2 cells are selective for the orientation, contrast, and *side of border ownership* of an edge ... for edges defined by luminance *or stereo disparity*



"anti-correlated" stereogram

Later, in area V4, neural responses to stereo disparity appear to correspond more closely to perceived depth

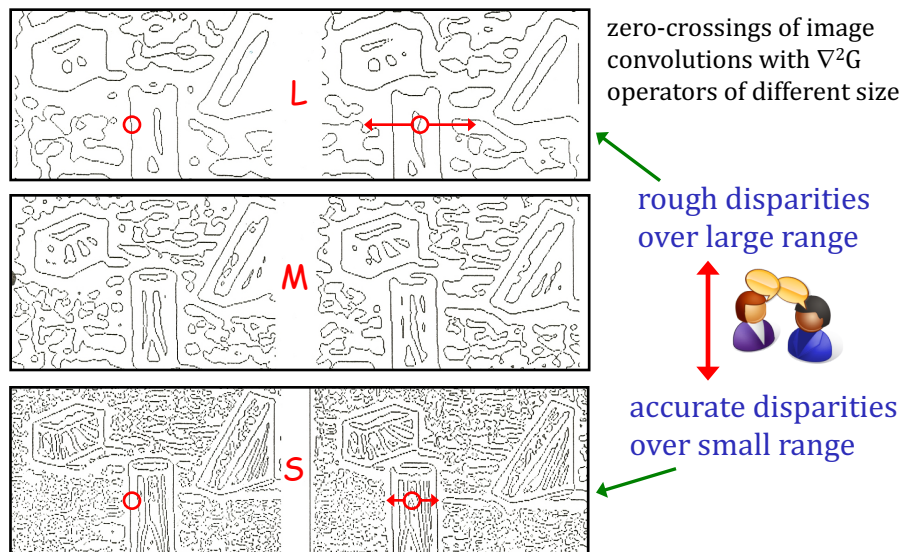
1-4

## In summary, some key points...

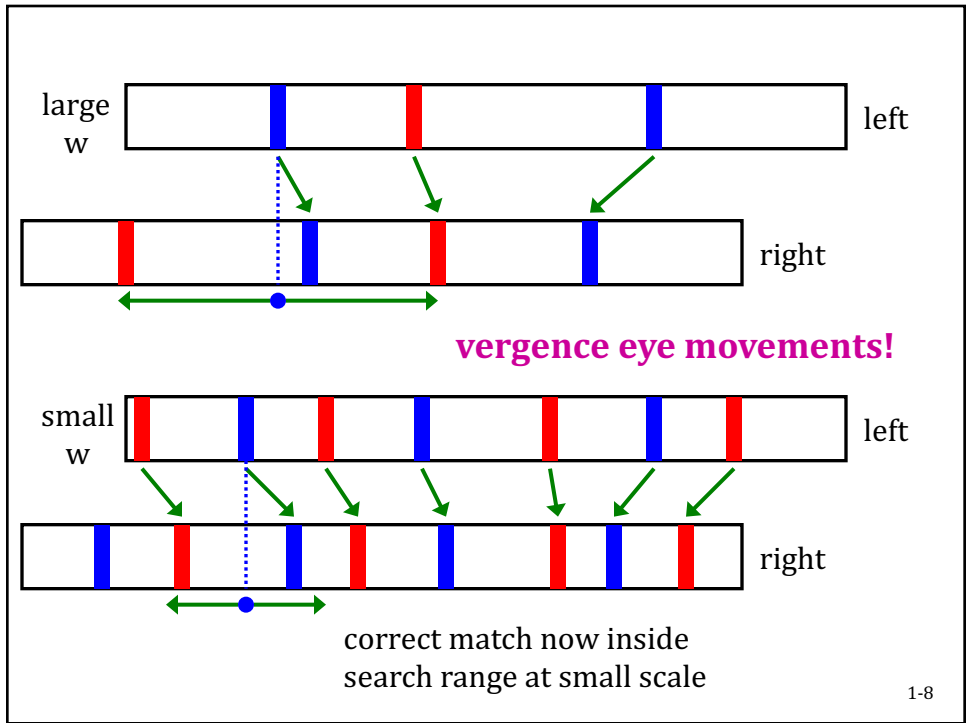
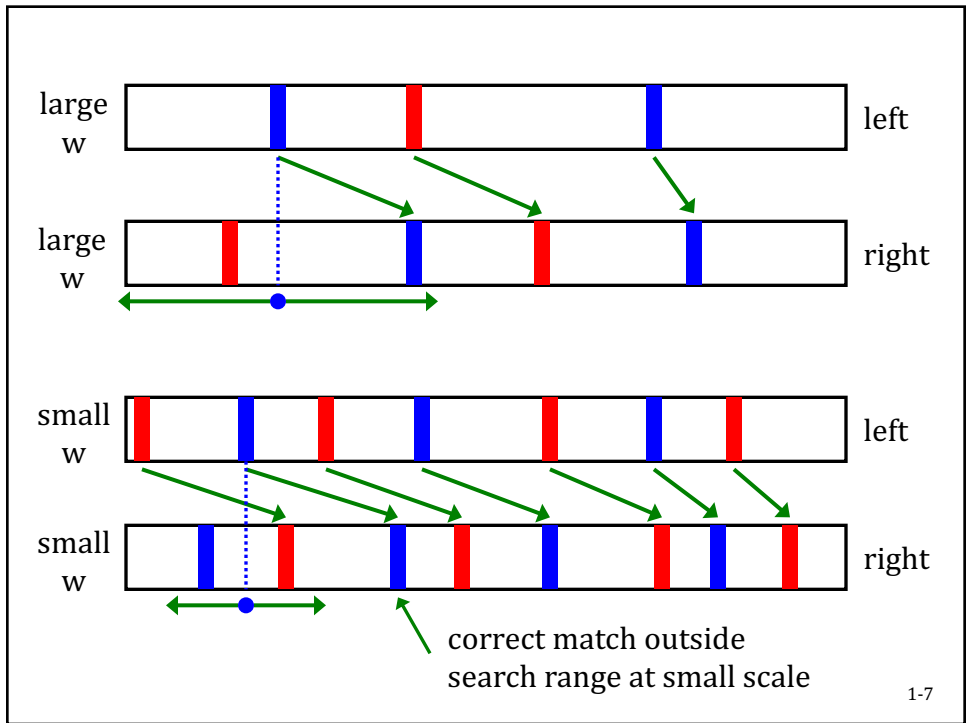
- Image features used for matching:  
~simple, precise locations, similar between left/right images
- At single fixation, match features over a limited range of horizontal & vertical disparity
- Eye movements used to match features over larger range of disparity
- Stereo matching performed at multiple scales  
~independently, disparity range depends on scale
- Neurons selective for different ranges of stereo disparity,  
multiple processing stages V1 → V2 → V4

1-5

## Matching features for the MPG stereo algorithm



1-6

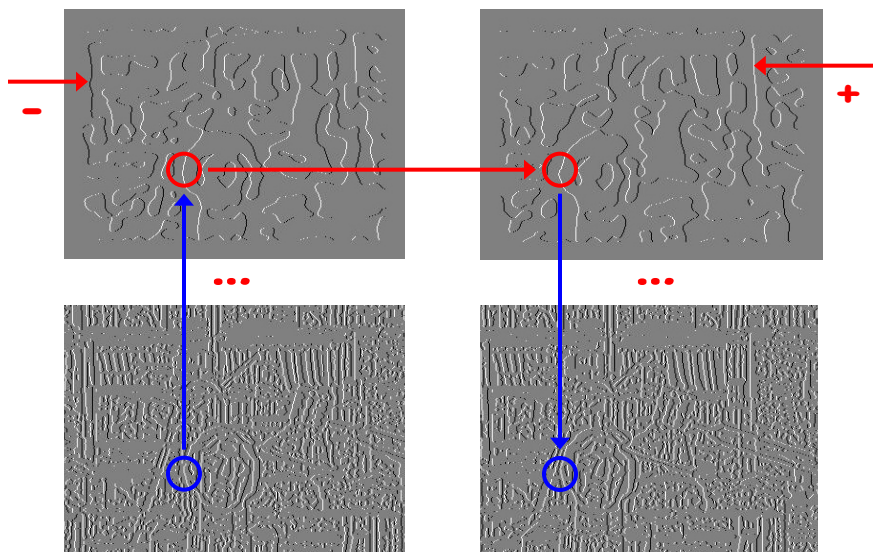


## Stereo images (Tsukuba, CMU)



1-9

## Zero-crossings for stereo matching



1-10

## Simplified MPG algorithm, Part 1

To determine initial correspondence:

- (1) Find zero-crossings using a  $\nabla^2G$  operator with central positive width  $w$
- (2) For each horizontal slice:
  - (2.1) Find the nearest neighbors in the right image for each zero-crossing fragment in the left image
  - (2.2) Find the nearest neighbors in the left image for each zero-crossing fragment in the right image
  - (2.3) For each pair of zero-crossing fragments that are closest neighbors of one another, let the right fragment be separated by  $\delta_{\text{initial}}$  from the left. Determine whether  $\delta_{\text{initial}}$  is within the matching tolerance,  $m$ . If so, consider the zero-crossing fragments matched with disparity  $\delta_{\text{initial}}$

$$m = w/2$$

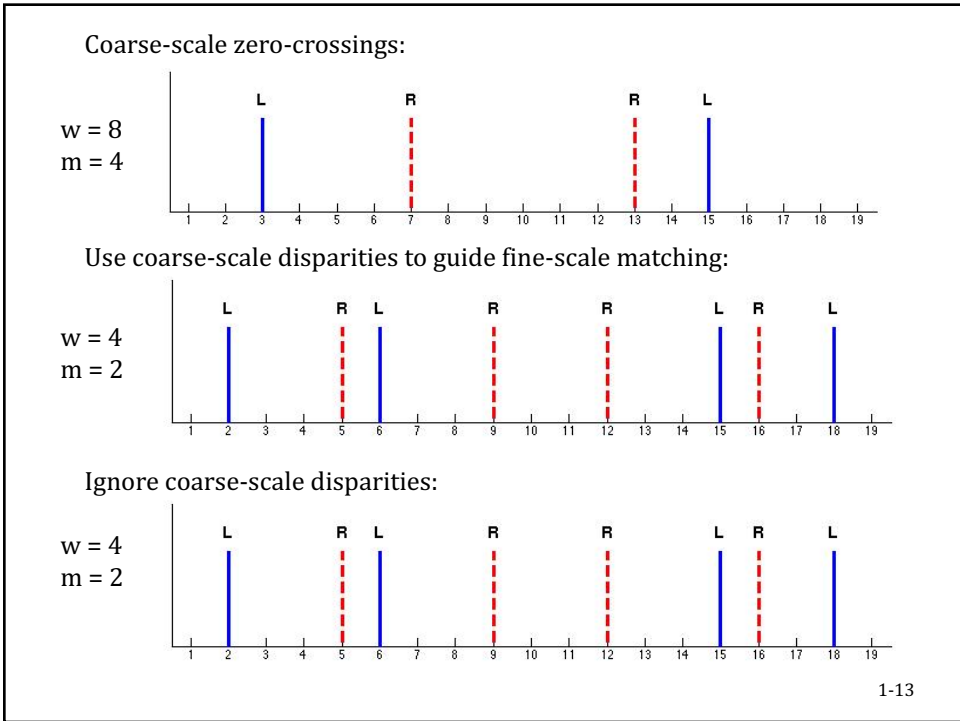
1-11

## Simplified MPG algorithm, Part 2

To determine final correspondence:

- (1) Find zero-crossings using a  $\nabla^2G$  operator with reduced width  $w/2$
- (2) For each horizontal slice:
  - (2.1) For each zero-crossing in the left image:
    - (2.1.1) Determine the nearest zero-crossing fragment in the left image that matched when the  $\nabla^2G$  operator width was  $w$
    - (2.1.2) Offset the zero-crossing fragment by a distance  $\delta_{\text{initial}}$ , the disparity of the nearest matching zero-crossing fragment found at the lower resolution with operator width  $w$
  - (2.2) Find the nearest neighbors in the right image for each zero-crossing fragment in the left image
  - (2.3) Find the nearest neighbors in the left image for each zero-crossing fragment in the right image
  - (2.4) For each pair of zero-crossing fragments that are closest neighbors of one another, let the right fragment be separated by  $\delta_{\text{new}}$  from the left. Determine whether  $\delta_{\text{new}}$  is within the reduced matching tolerance,  $m/2$ . If so, consider the zero-crossing fragments matched with disparity  $\delta_{\text{final}} = \delta_{\text{new}} + \delta_{\text{initial}}$

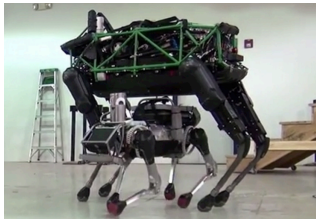
1-12



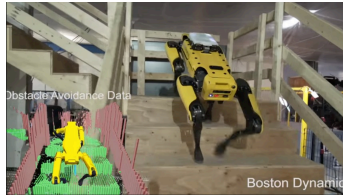
## Stereo vision applications - robots



Boston Dynamics Atlas Robot  
<https://www.youtube.com/watch?v=QQni8Uz3Q0I>



Boston Dynamics SpotMini  
[https://www.youtube.com/watch?v=Ve9kWX\\_KXus](https://www.youtube.com/watch?v=Ve9kWX_KXus)



Carnegie Robotics MultiSense S7  
<https://carnegierobotics.com/multisense-s7/>

Spirit Mars rover

ASIMO HONDA

humanoid robots

Robonaut to the rescue!

security robots

OpenCat - stay tuned!  
<https://www.hackster.io/petoi/open-cat-845129>

SMP Robotics

1-15

## Stereo vision applications - driving

**Large-Scale Direct SLAM with Stereo Cameras**  
 Jakob Engel, Jörg Stückler, Daniel Cremers  
 IROS 2015, Hamburg  
 Computer Vision Group  
 Technical University Munich

KITTI 00 (Full SLAM)  
 00:06:01.000 (154 speed)  
 left video keyframe depth

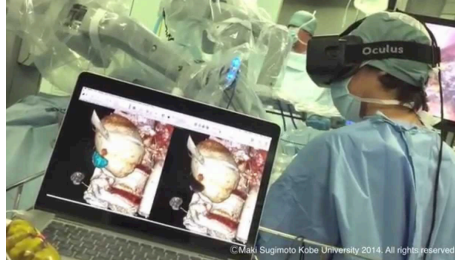
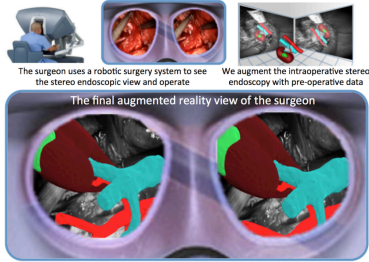
**SLAM: Simultaneous Localization and Mapping**  
<https://www.youtube.com/watch?v=oJt3Ln8H03s>

Automated Driving  
 Prof. Amnon Shashua  
 In Robotics, Deep Learning, CV

1-16



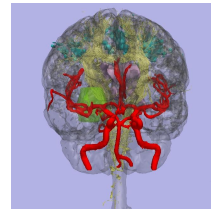
## Stereo visualization for image guided surgery



Da Vinci surgical robot



CMU 3D overlay system



MIT/BWH

1-17