

Binocular Stereo



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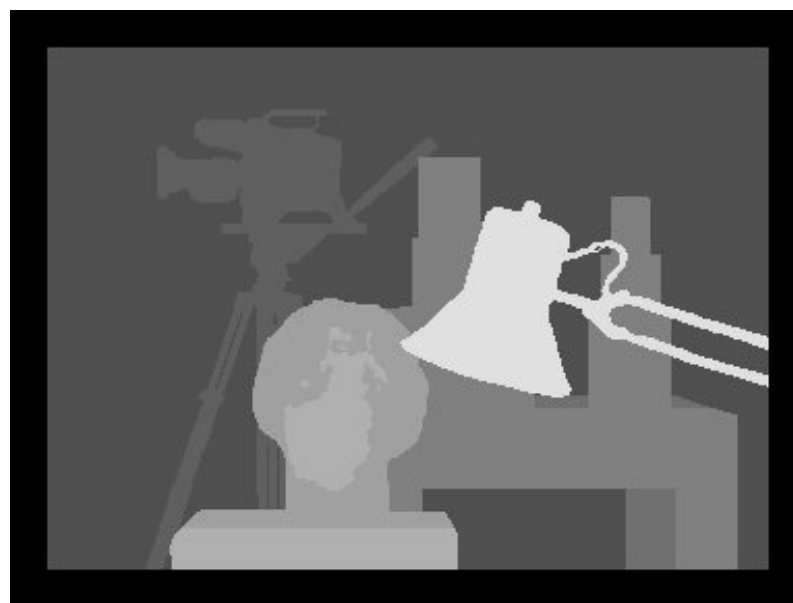
University of Erlangen-Nuremberg

Stereo Vision

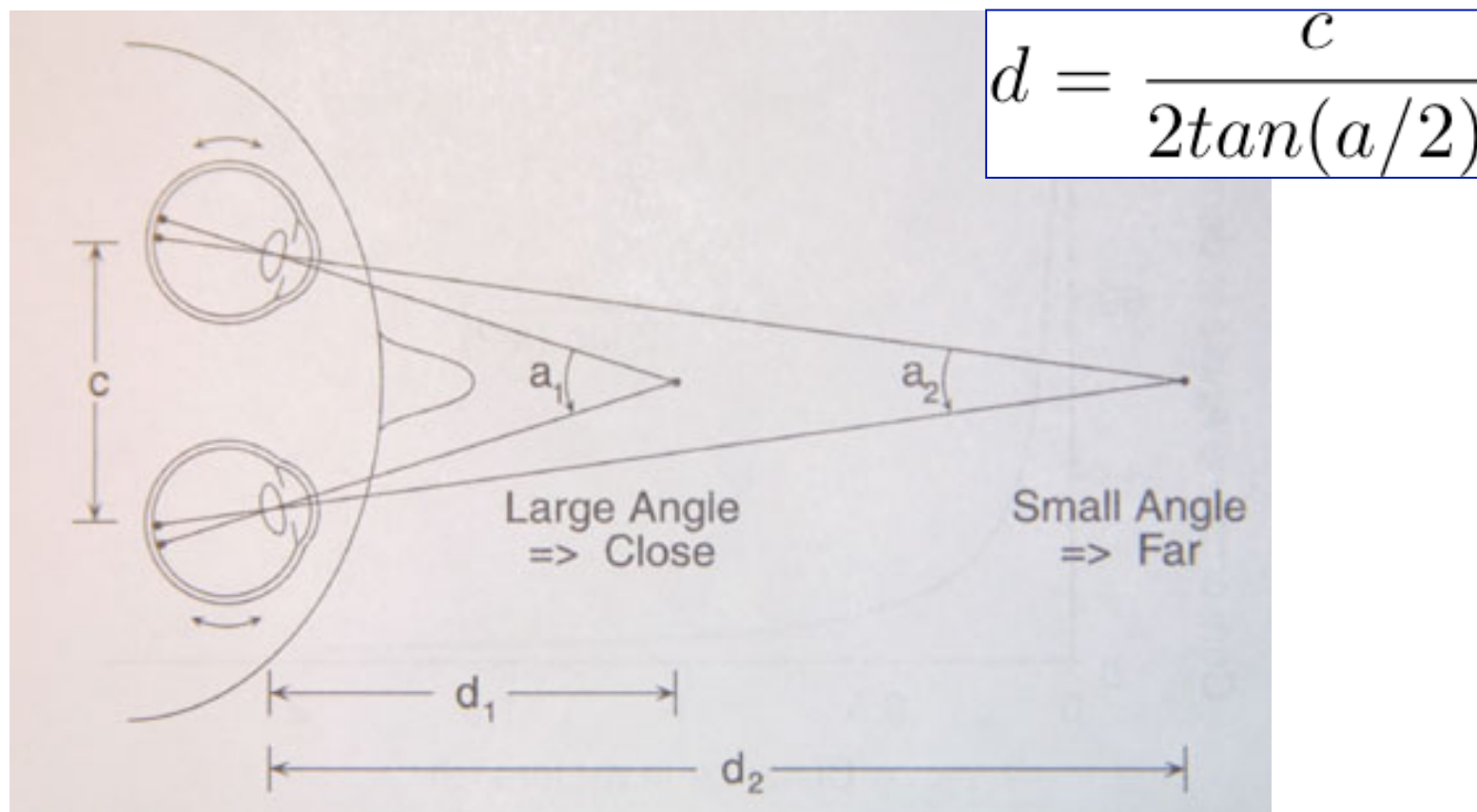


- Goal: Infer information about the 3-D structure and distances of a scene from two or more images taken from different viewpoints.
- A stereo system must solve two subproblems:
 - Correspondence problem
 - Reconstruction
- *Correspondence Problem*: which point on the left image and on the right image are projections of the same scene point.
- Once the point correspondence is established, we can compute the relative shift, the *disparity*, between the two projections.
- *Reconstruction*: The disparity data is then converted to a 3D map. In order to transform the disparity data to 3D measurements, we need some form of knowledge about the geometry of the stereo system.

Binocular Stereo Example

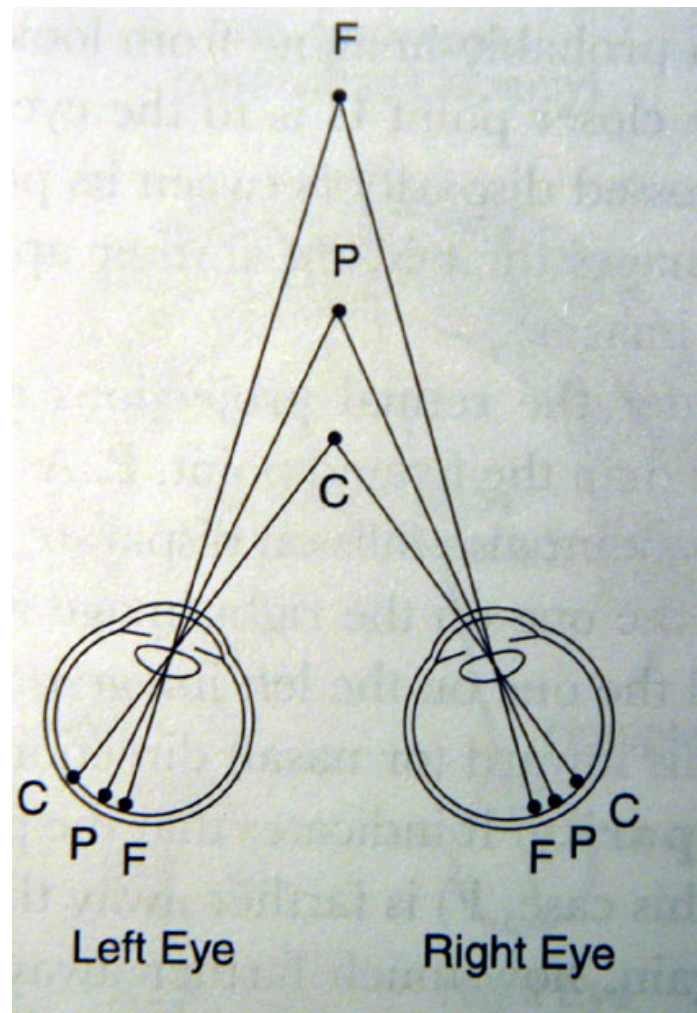


Depth from Convergence



Human performance: up to 2-2.5 meters

Depth from Binocular Disparity



P: converging point

C: object nearer
projects to the
outside of the P,
disparity = +

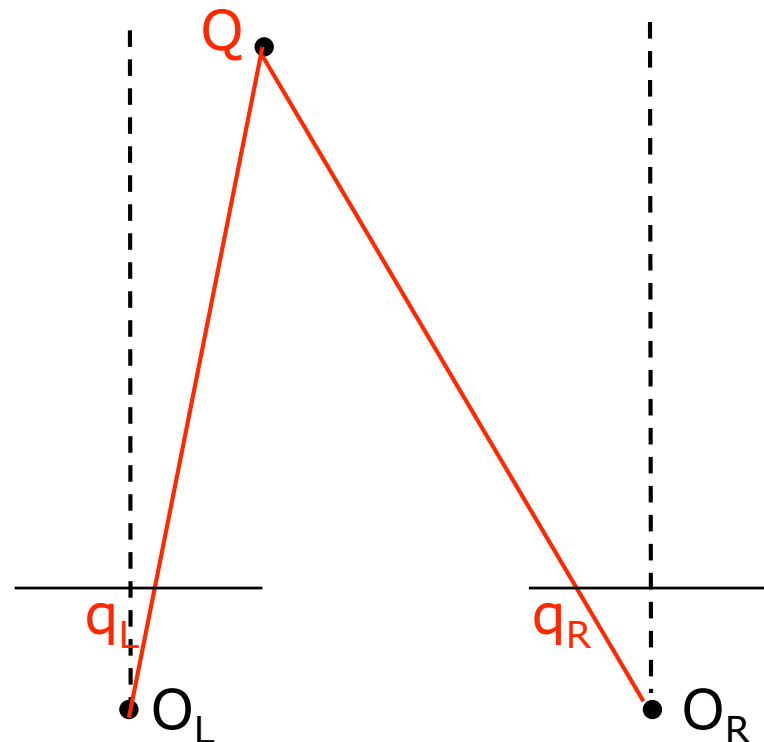
F: object farther
projects to the
inside of the P,
disparity = -

Sign and magnitude of disparity

Simple Binocular Stereo Setup



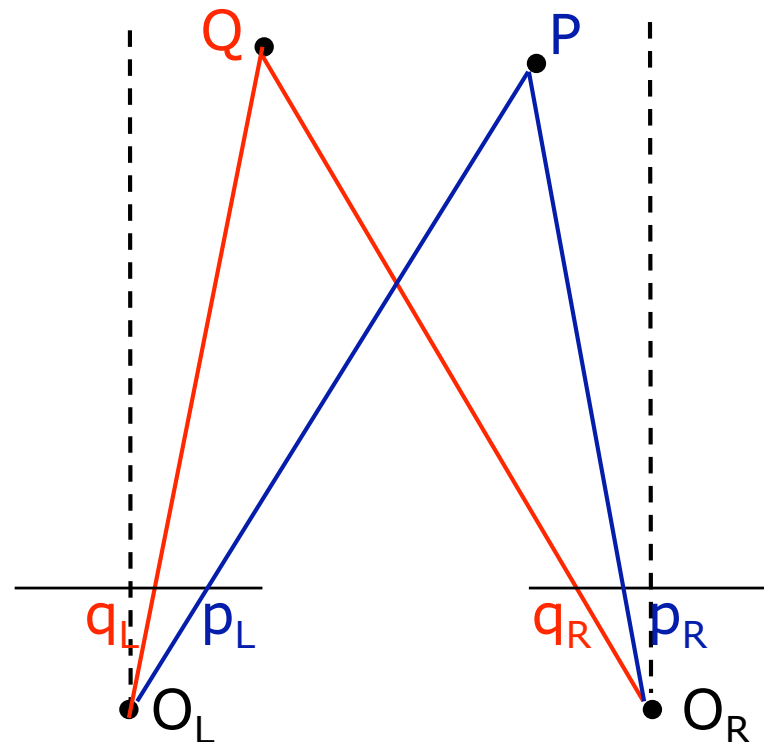
- Parallel optic axes, i.e. the fixation point (the point where the 2 optic axes intersect) is at infinity.
- Both image planes lie on the same plane.
- Their scan lines are aligned (scan-line coherence), i.e. $y = y'$.



Correspondence and Triangulation



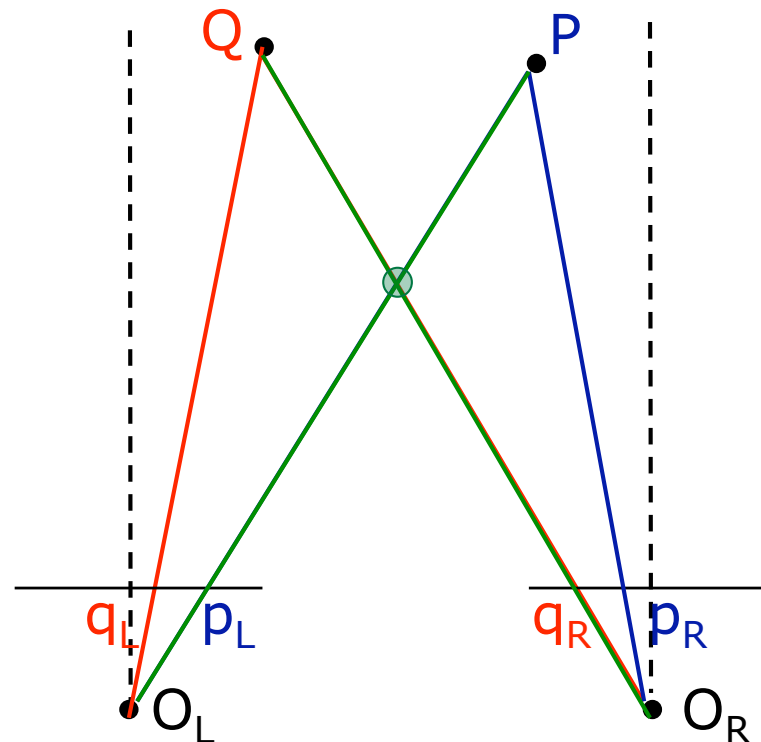
- When we correspond correctly (i.e. q_L with q_R and p_L with p_R), the intersection of the corresponding rays gives the 3D location of scene point that generated the projections (i.e. Q and P accordingly).



Impact of Correspondence



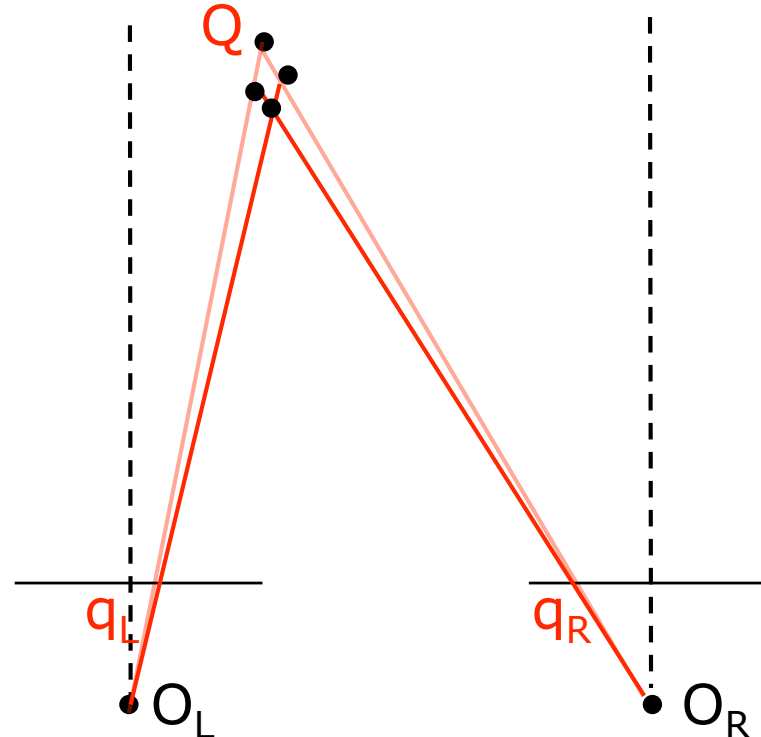
- A mistake in correspondence, e.g. q_R is matched with p_L , will result in the intersection of rays that correspond to projection of distinct points (Q and P). As a result the wrong 3D location is recovered.



Noise and Correspondence

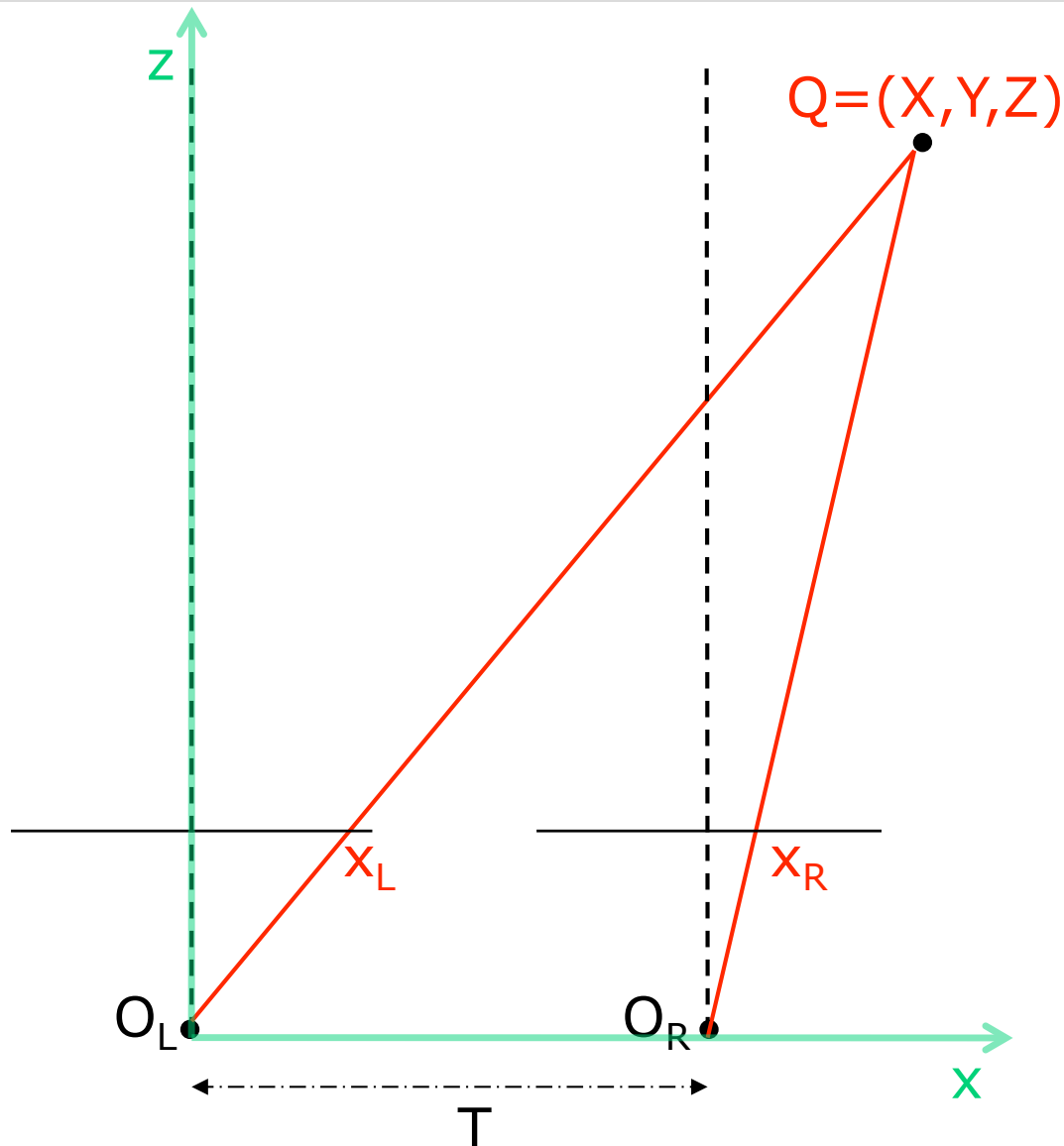


- The noise in the image capture process (sensor noise, quantization, discretization) introduce inaccuracies in the projection rays that directly affect the triangulation process.





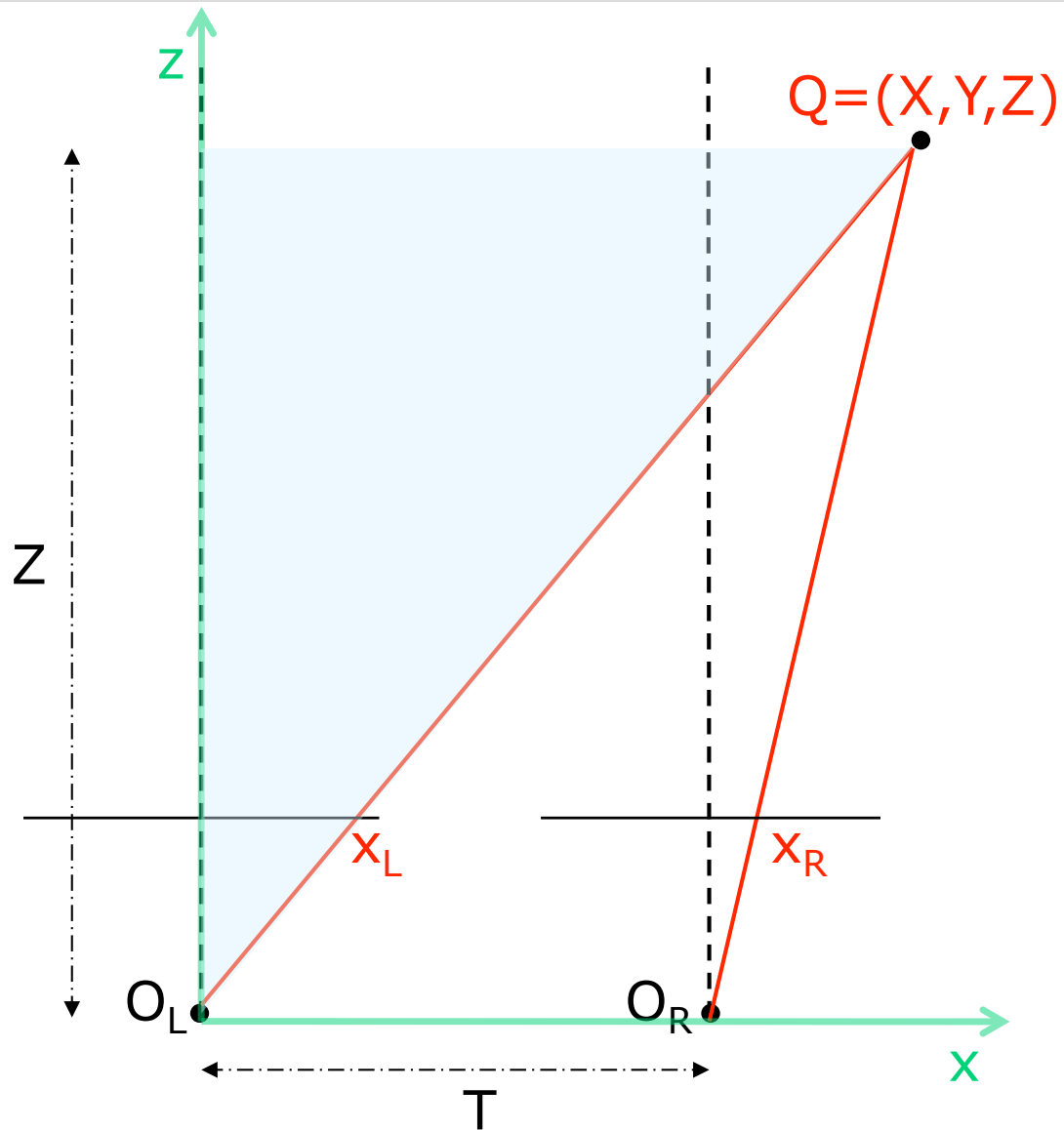
Triangulation



- Assume that the correspondence has been correctly established.
- Under the simple binocular setup (parallel optic axes and scan-line coherence), the only difference between the two projections q_L and q_R is in the x-component, i.e. x_L versus x_R .
- Let T be the **baseline**, i.e. the distance between the two COPs.

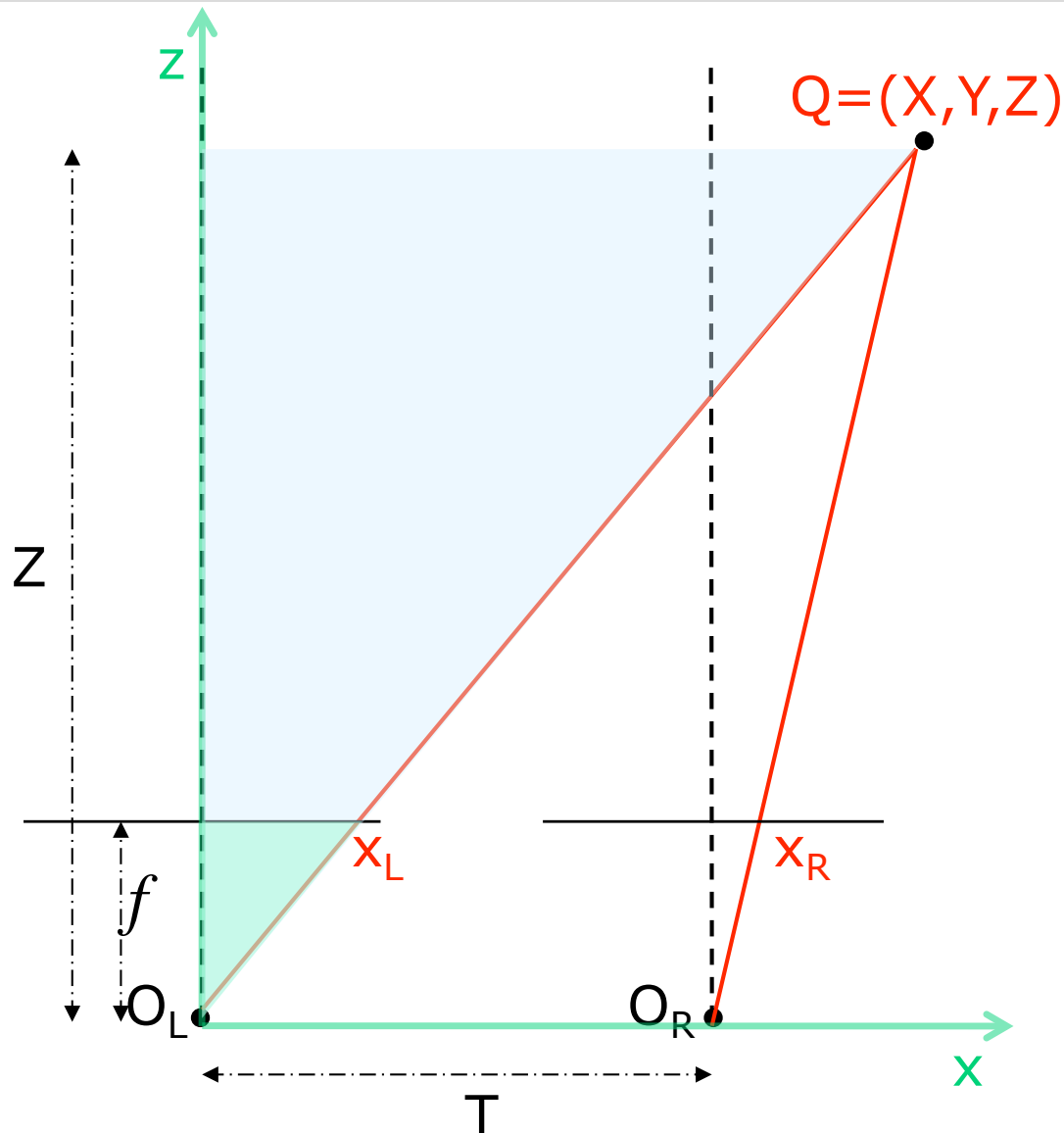


Triangulation





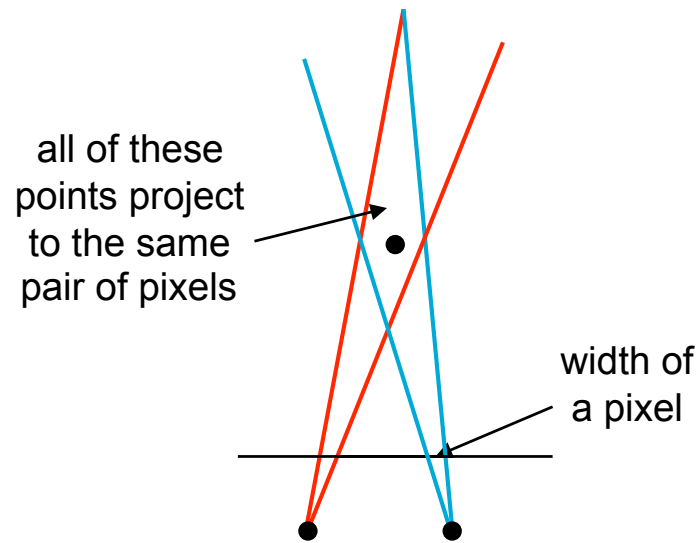
Triangulation



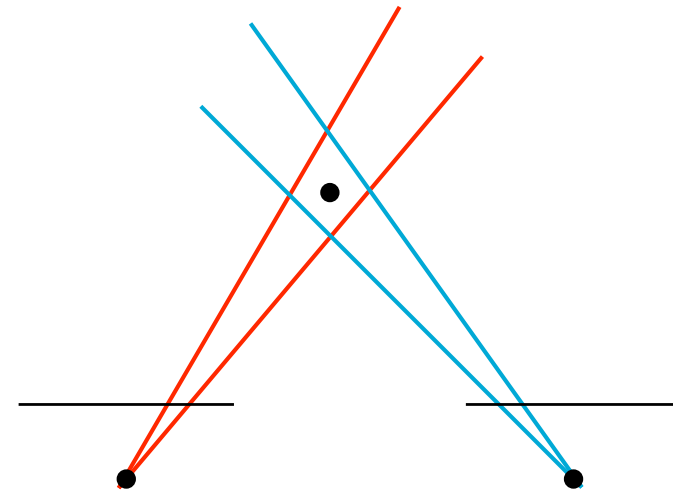
From the similar triangles:

$$\frac{x_L}{f} = \frac{X}{Z} \Rightarrow X = x_L \frac{Z}{f}$$

Impact of Baseline



Small Baseline



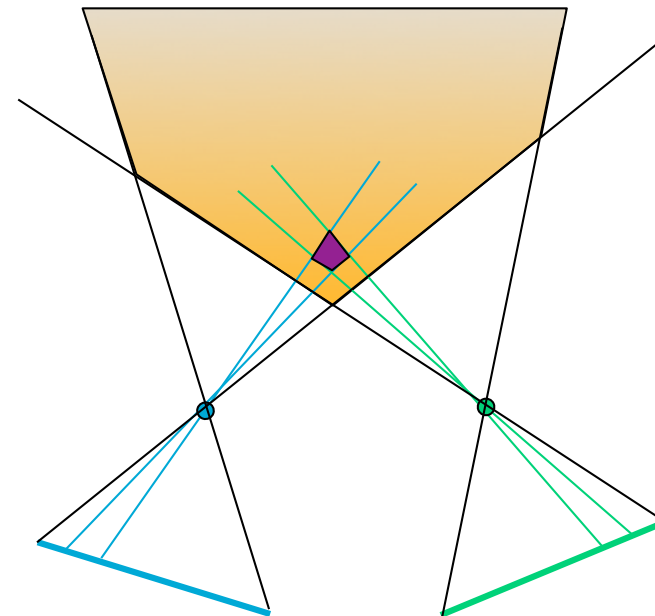
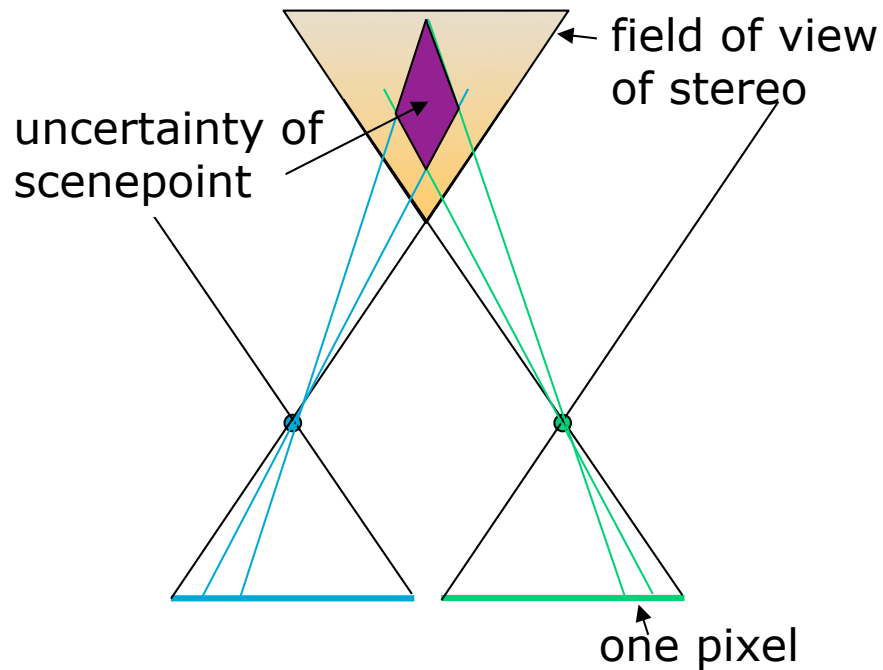
Large Baseline

■ What's the optimal baseline?

- Too small: large depth error
- Too large: difficult search problem
 - Appearance may change between the 2 viewpoints
 - Decrease in part of the scene that is mutually visible



Vergence



Optical axes of the two cameras need not be parallel

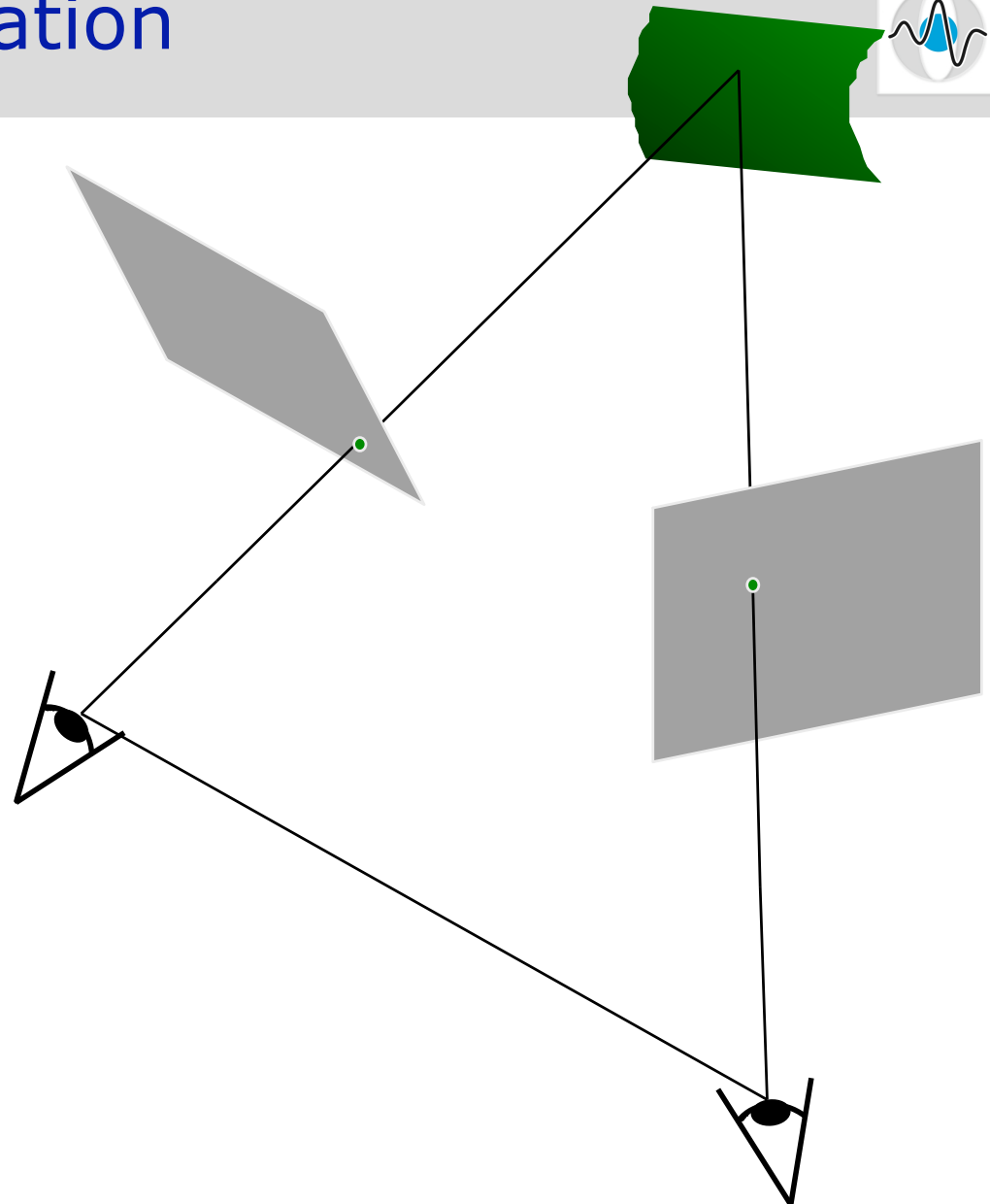
■ Solution: Vergence (turn cameras towards each other)

- Increases the field of view
- Increases accuracy in the correspondence

Stereo Image Rectification

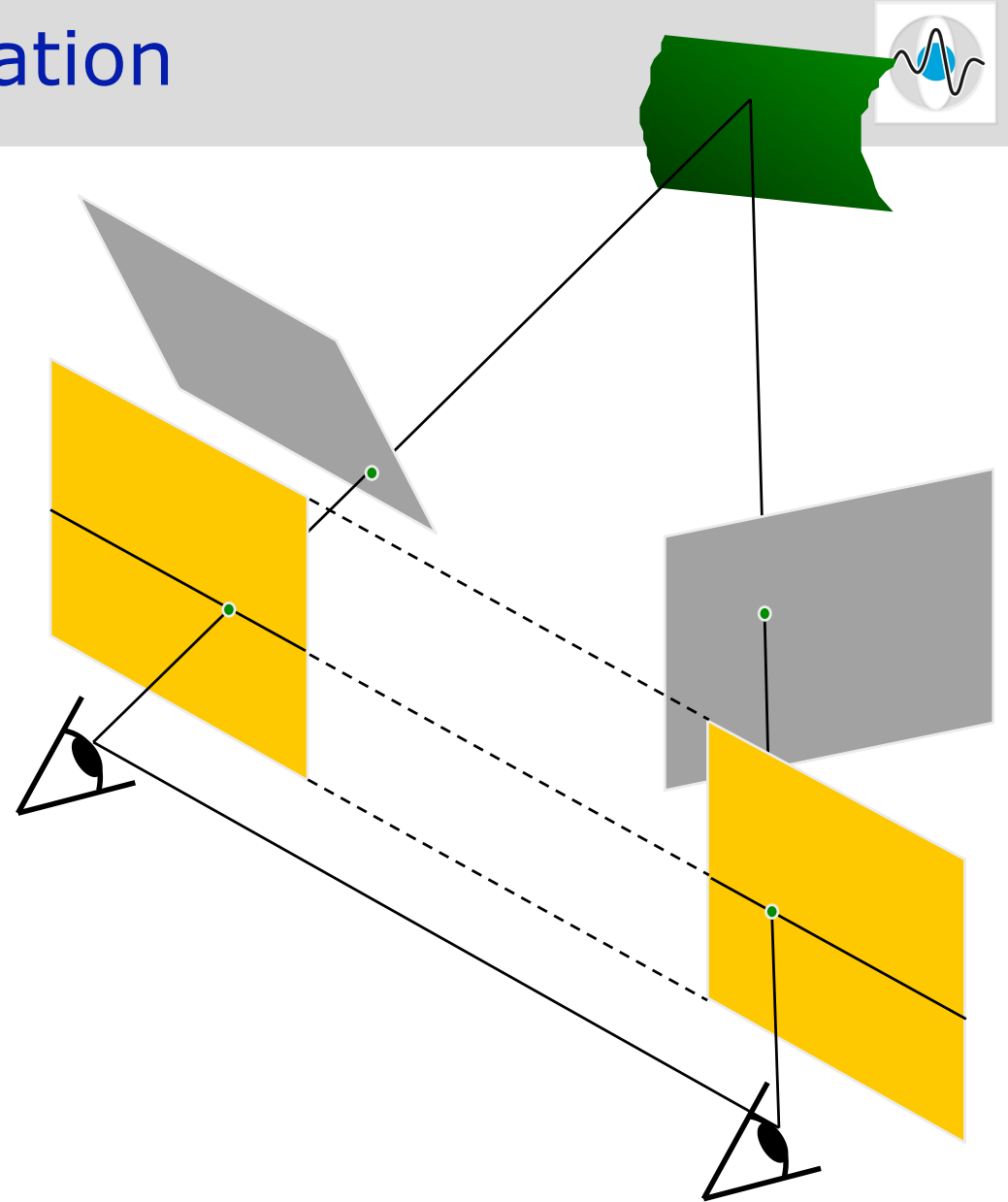


- So far we have assumed:
 - parallel optic axes
 - scan-line coherence
- Such a setup can lead to inaccuracies.
- More commonly cameras are *verged*, i.e. the 2 optic axes intersect each other.
- Can we use the same math?



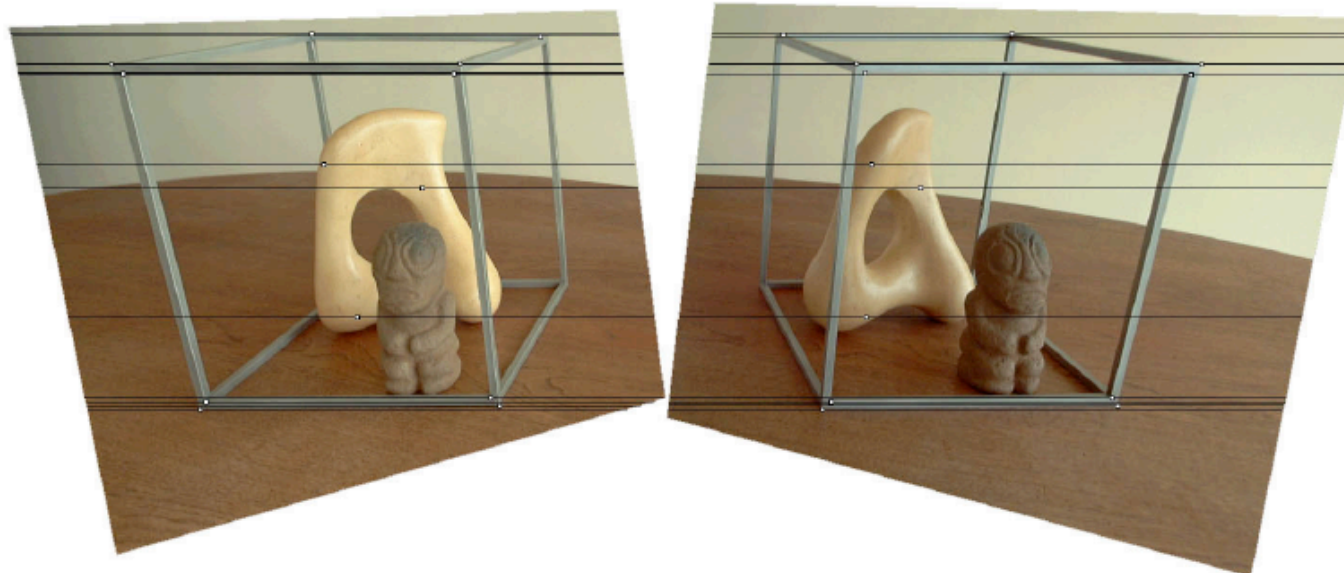
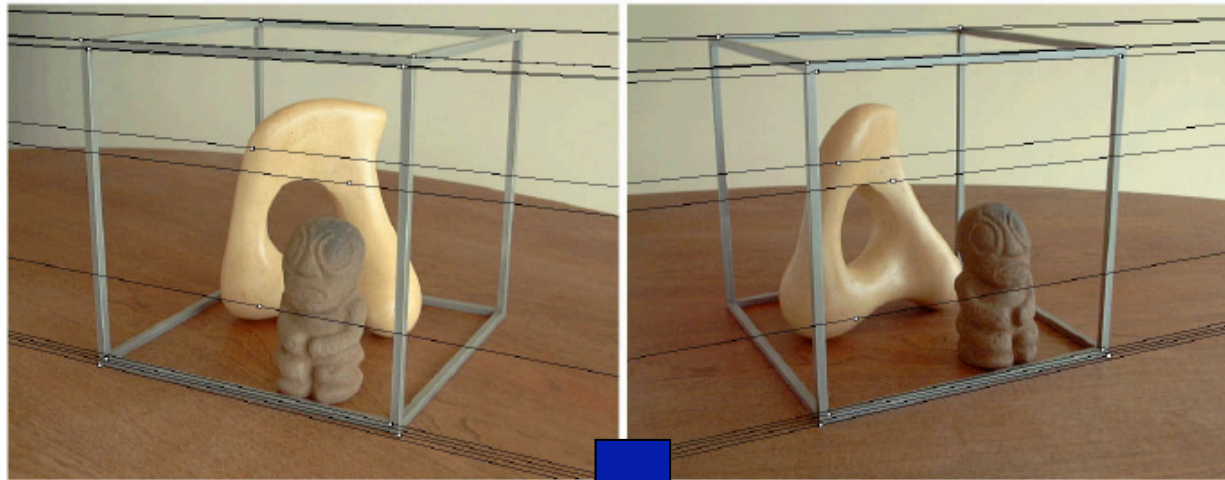
Stereo Image Rectification

- Yes!
- Re-project the image planes onto a common plane parallel to the baseline (the line between the two centers of projection).
- Two virtual image planes are created, which are now scan-line coherent.
- Do all the computations on these rectified (virtual) image planes.





Stereo Rectification Example



Correspondence Problem



■ Assumptions:

- Most scene points are visible from both viewpoints
- Corresponding image regions look similar

■ It is a search problem: Given an element in the left image, search the right image to find the corresponding element.

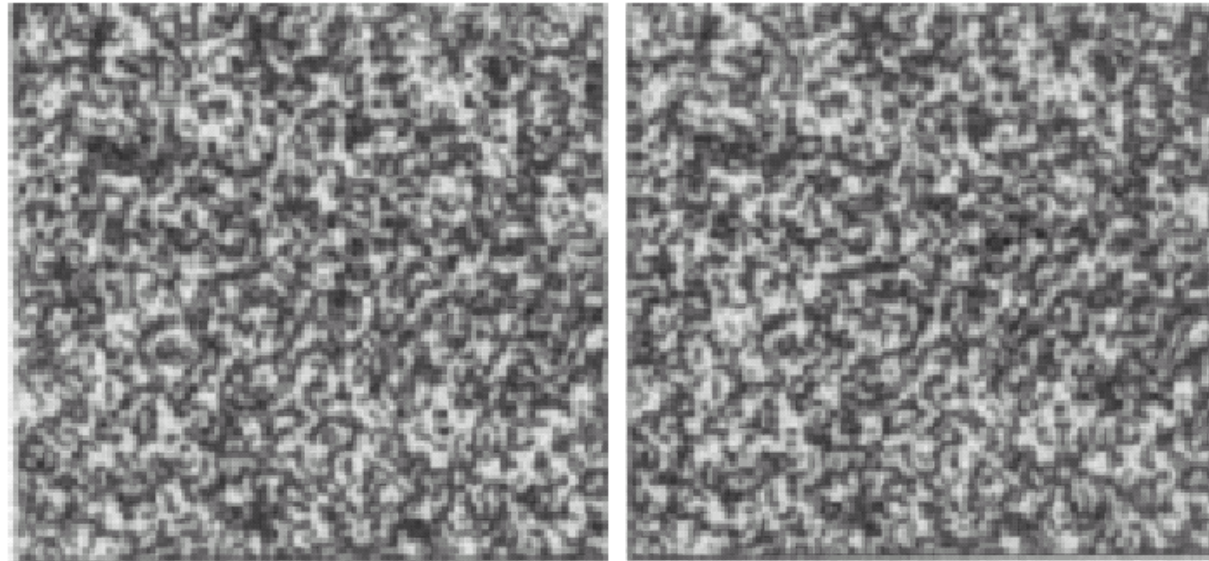
■ Three underlying questions:

1. What do we match between the two images? (objects, edges, pixels, sets of pixels?)
2. What measure of similarity do we use?
3. Can we search in a systematic way?

Point Correspondence



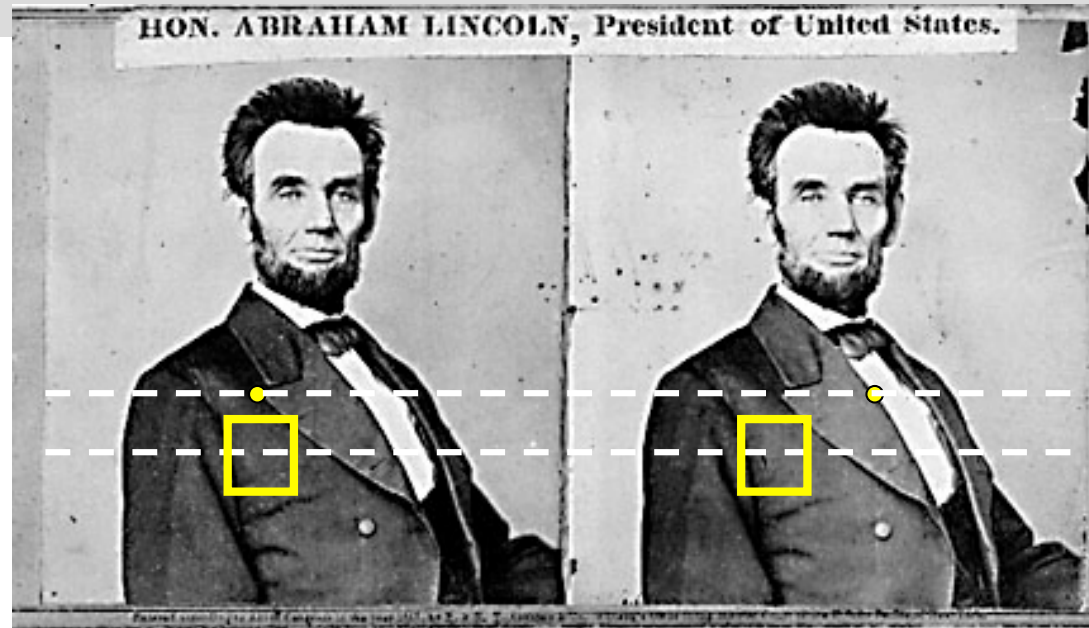
Random dot stereograms



Julesz: had huge impact because it showed that recognition not needed for stereo.



Point Correspondence in Practice



For each scan-line (more properly epipolar line)

For each pixel in the left image

- compare with every pixel on same epipolar line in right image
- pick pixel with closest intensity value (or more general minimum match cost).
- This will never work, so:

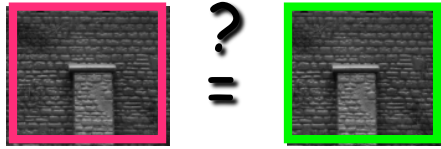
Improvement: match ***windows***

Compare Regions around Points



- Idea: Compare intensity profiles around neighborhoods of potential points.
- Elements to be matched are now image windows of fixed size.
- The similarity measure is the correlation between windows in the two images.

Similarity Metrics



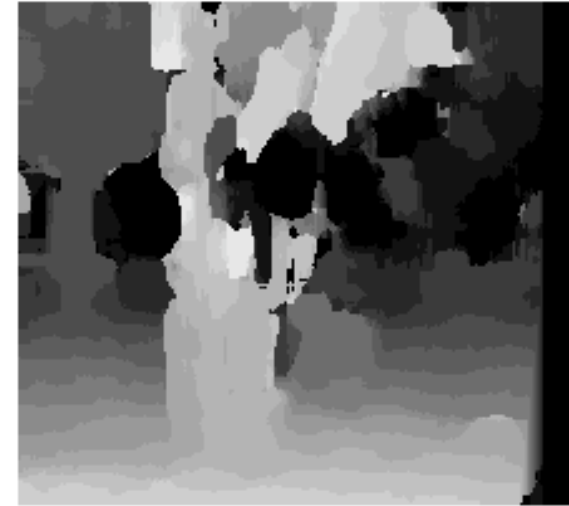
$$\begin{aligned}
 SSD &= \sum_{[i,j] \in R} (f(i,j) - g(i,j))^2 \\
 C_{fg} &= \sum_{[i,j] \in R} f(i,j)g(i,j) \\
 NC_{fg} &= \frac{1}{n-1} \sum_{[i,j] \in R} \frac{(f(i,j) - \bar{f})(g(i,j) - \bar{g})}{\sigma_f \sigma_g}
 \end{aligned}
 \left. \vphantom{\begin{aligned} SSD \\ C_{fg} \\ NC_{fg} \end{aligned}} \right\} \text{Most popular}$$

For each window, match to the closest window on the horizontal (epipolar) line in the other image.

Window Size



$W = 3$



$W = 20$

- Smaller window: more detail, more noise.
- Larger window: less noise, less detail
- Better results with adaptive window size

Compare Features

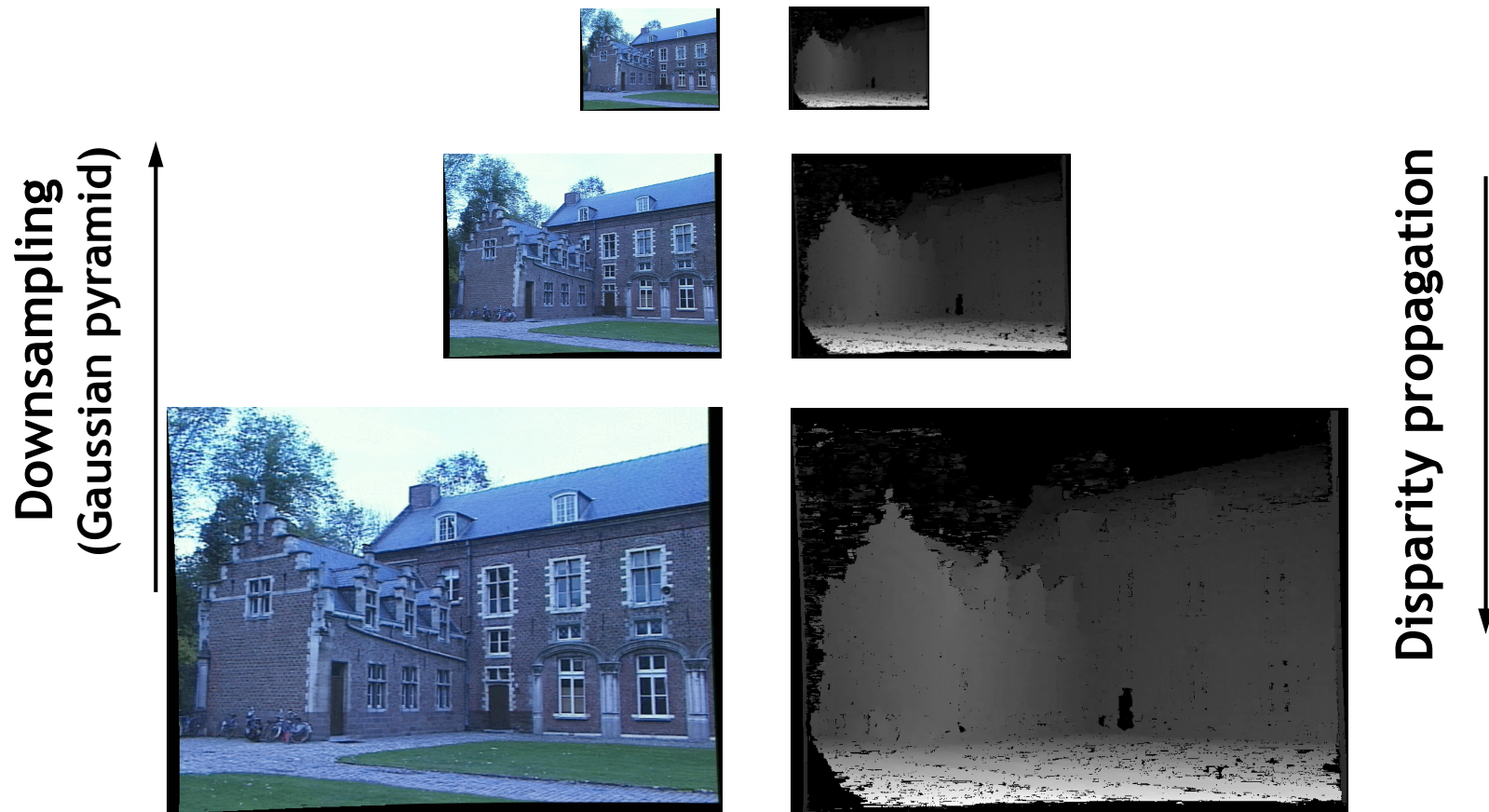


- Another Idea: Compute features and match only pixels based on their feature values.
- Possible features:
 - Edges
 - Lines...
- Pros: Possibly more unique values => easier correspondence
- Cons: Not all the pixels have a feature value => sparse correspondence; need for interpolation
- Often used in combination with hierarchical correspondence.

Hierarchical Correspondence



- Allows faster computation
- Can handle large disparity ranges





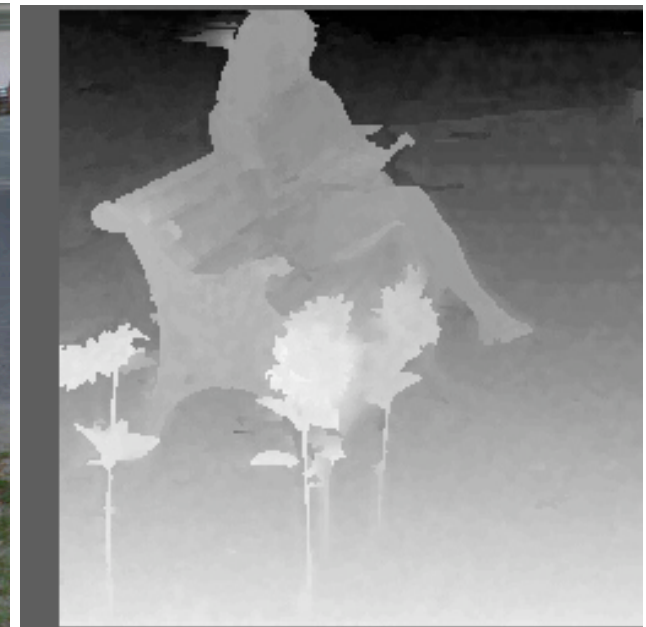
Stereo Example



left image



right image



depth map

H. Tao et al. "[Global matching criterion and color segmentation based stereo](#)"

Reconstruction



H. Tao et al. "[Global matching criterion and color segmentation based stereo](#)"
Elli Angelopoulou



Image Sources

1. The slides on image rectification are courtesy of J. Chai,
http://faculty.cs.tamu.edu/jchai/cpsc641_spring10/lectures/lecture9.ppt
2. A number of slides in this presentation have been adapted by the presentation of S. Narasimhan,
<http://www.cs.cmu.edu/afs/cs/academic/class/15385-s06/lectures/ppts/lec-14.ppt>
3. The Lincoln image is courtesy of S. Seitz.
4. The window-matching slide is courtesy of O. Camps.
5. The example slide on hierarchical correspondence algorithms is courtesy of ETH,
<http://www.inf.ethz.ch/personal/pomarc/courses/qcv/class07.ppt>