

COS 429: COMPUTER VISION

STEREO (1 lecture)

- Stereo Reconstruction
 - The Stereo Fusion Problem
 - Random Dot Stereograms
 - Binocular Fusion Algorithms
-
- **Reading:** Chapters 11

Stereopsis

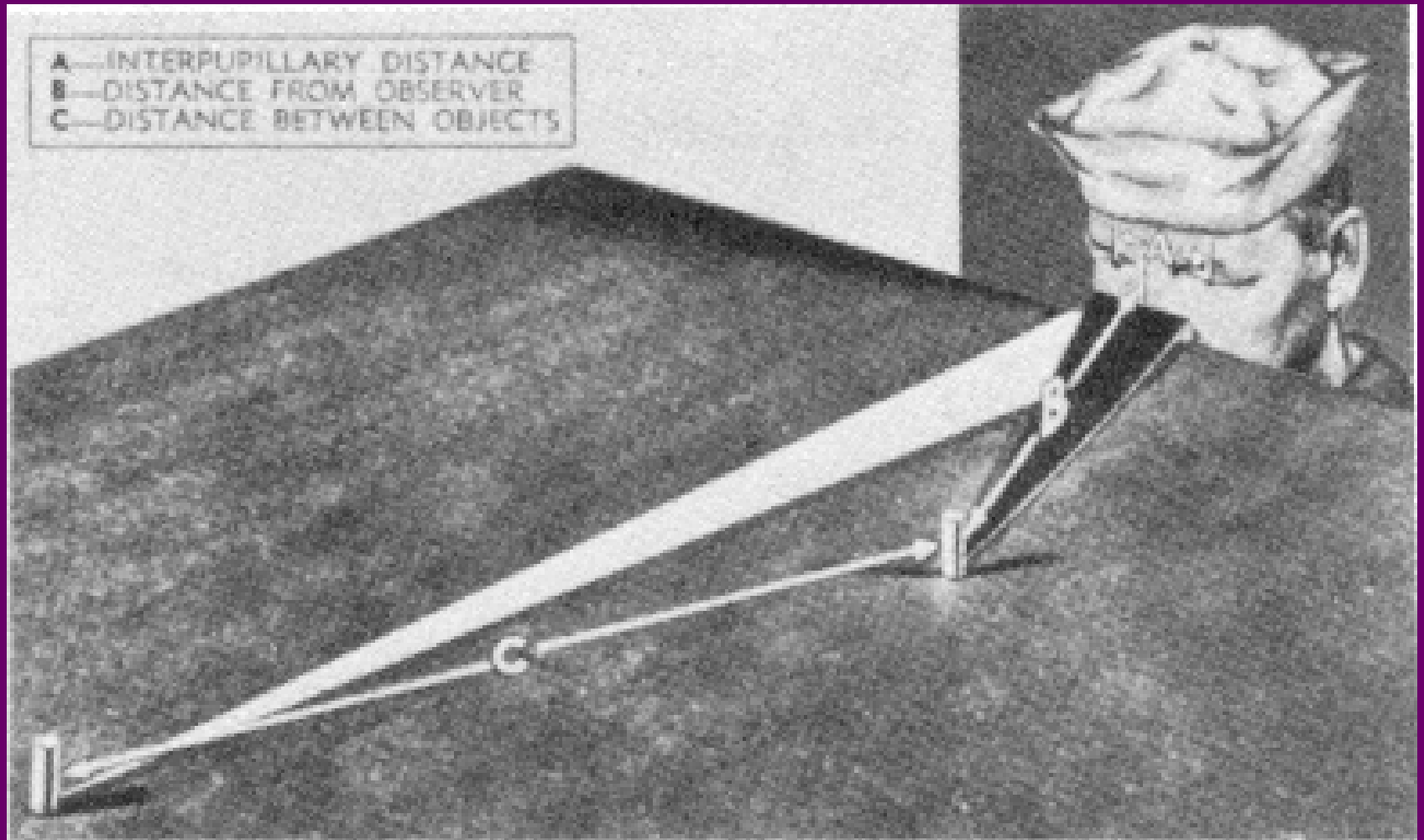


Figure from US Navy Manual of Basic Optics and Optical Instruments, prepared by Bureau of Naval Personnel. Reprinted by Dover Publications, Inc., 1969.



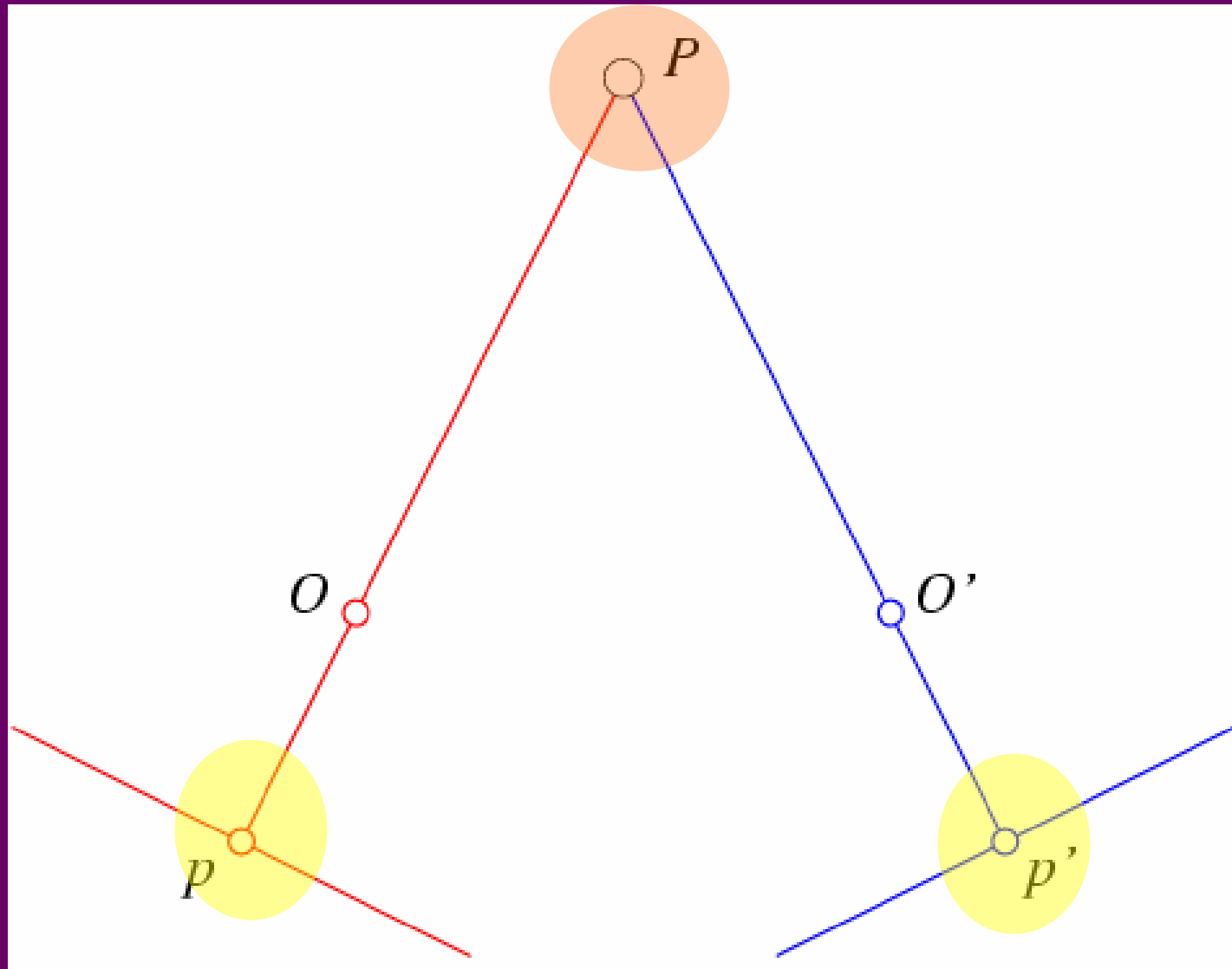
A spider



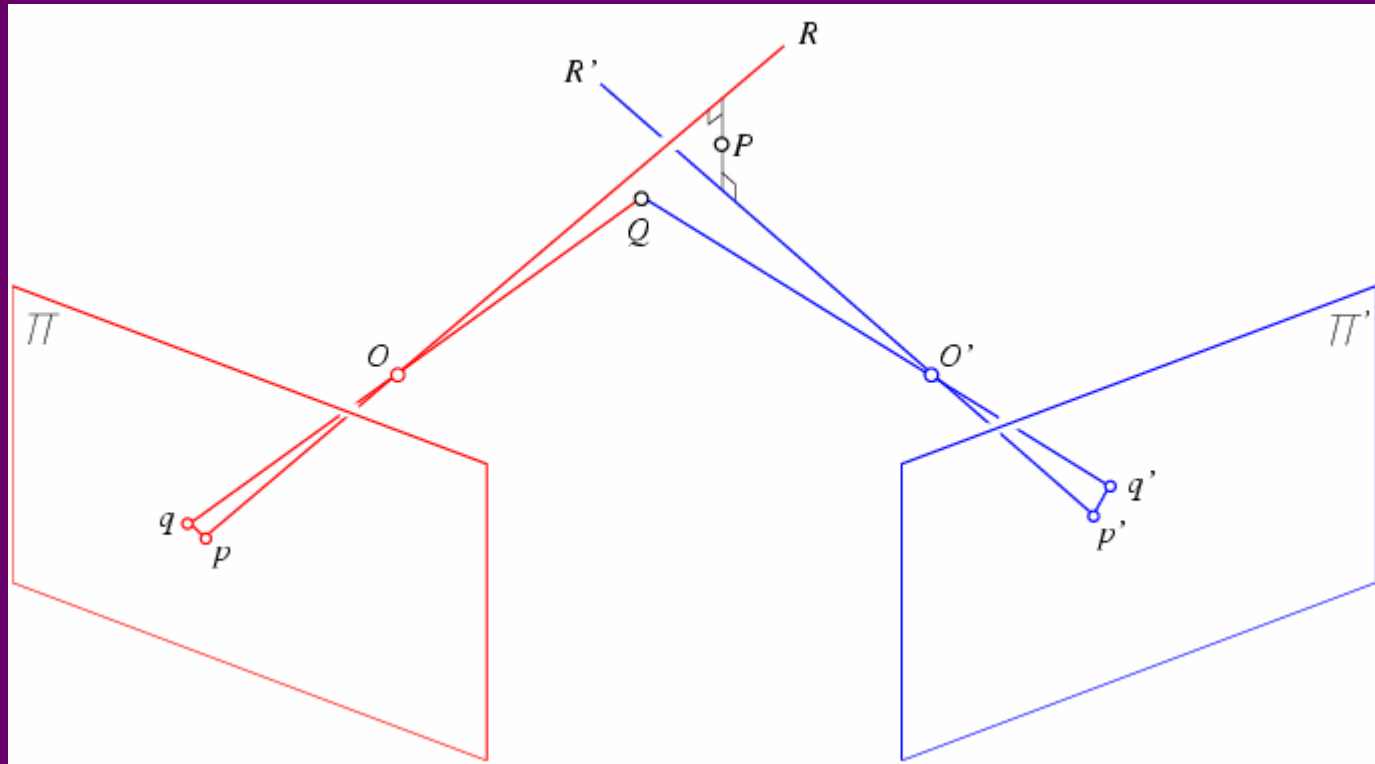
The INRIA Mobile Robot, 1990.



Stereo vision: Fusion and Reconstruction



Reconstruction



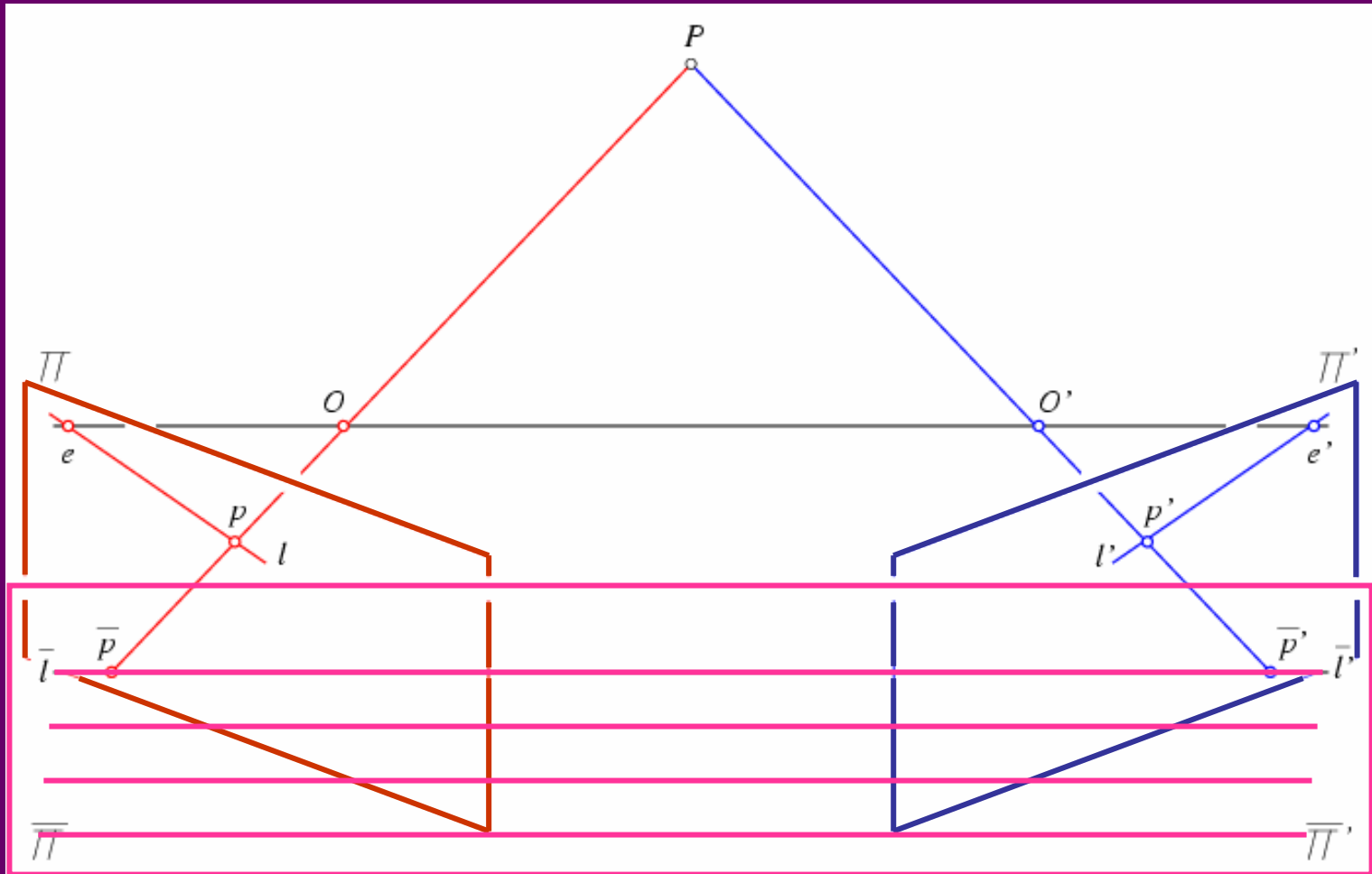
- Linear Method:
find P such that

$$\begin{cases} \mathbf{p} \times \mathcal{M}\mathbf{P} = 0 \\ \mathbf{p}' \times \mathcal{M}'\mathbf{P} = 0 \end{cases} \iff \begin{pmatrix} [\mathbf{p}_\times] \mathcal{M} \\ [\mathbf{p}'_\times] \mathcal{M}' \end{pmatrix} \mathbf{P} = 0$$

- Non-Linear Method: find Q minimizing

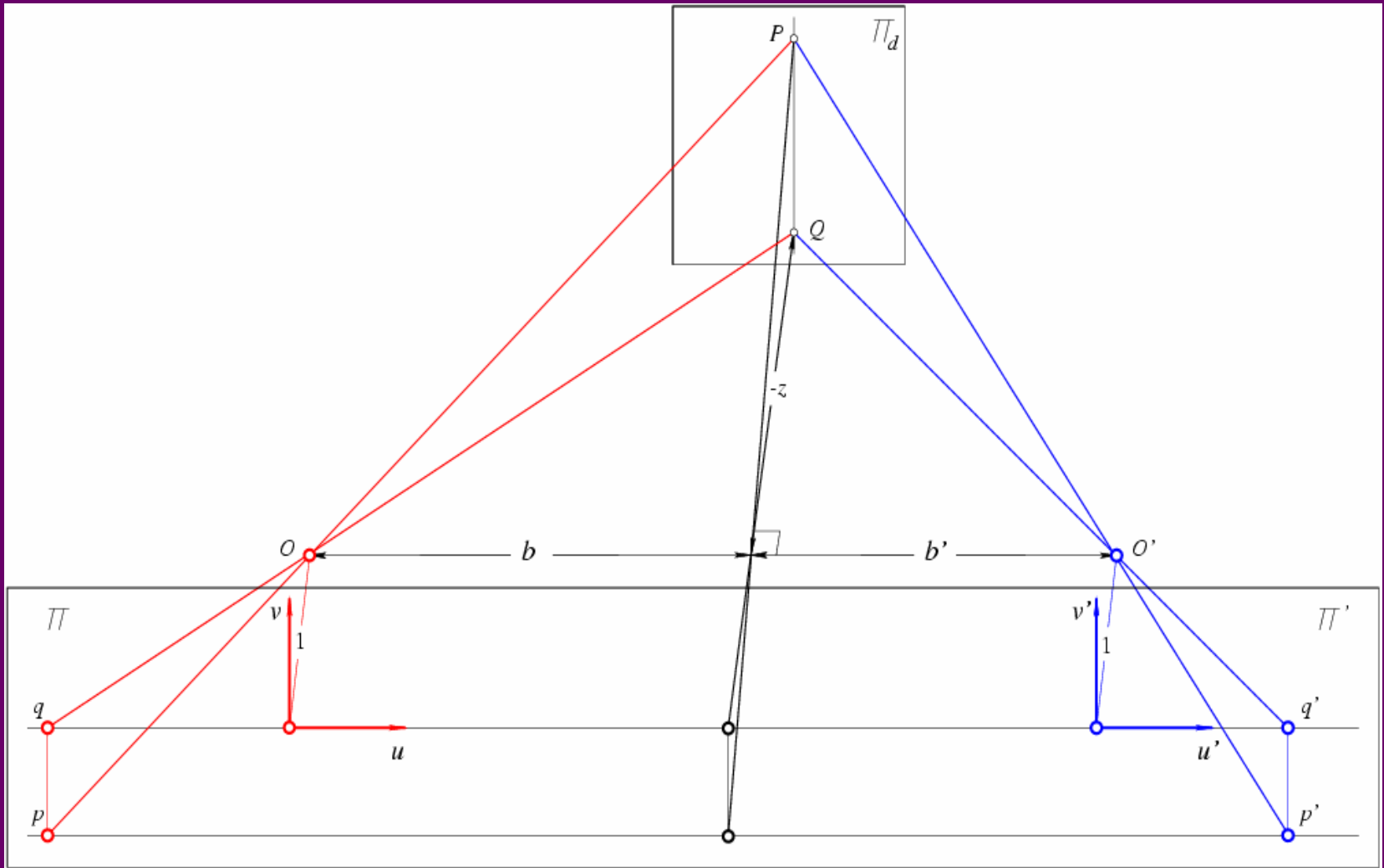
$$d^2(p, q) + d^2(p', q')$$

Rectification



All epipolar lines are parallel in the rectified image plane.

Reconstruction from Rectified Images



Disparity: $d = u' - u$.



Depth: $z = -B/d$.

Stereopsis

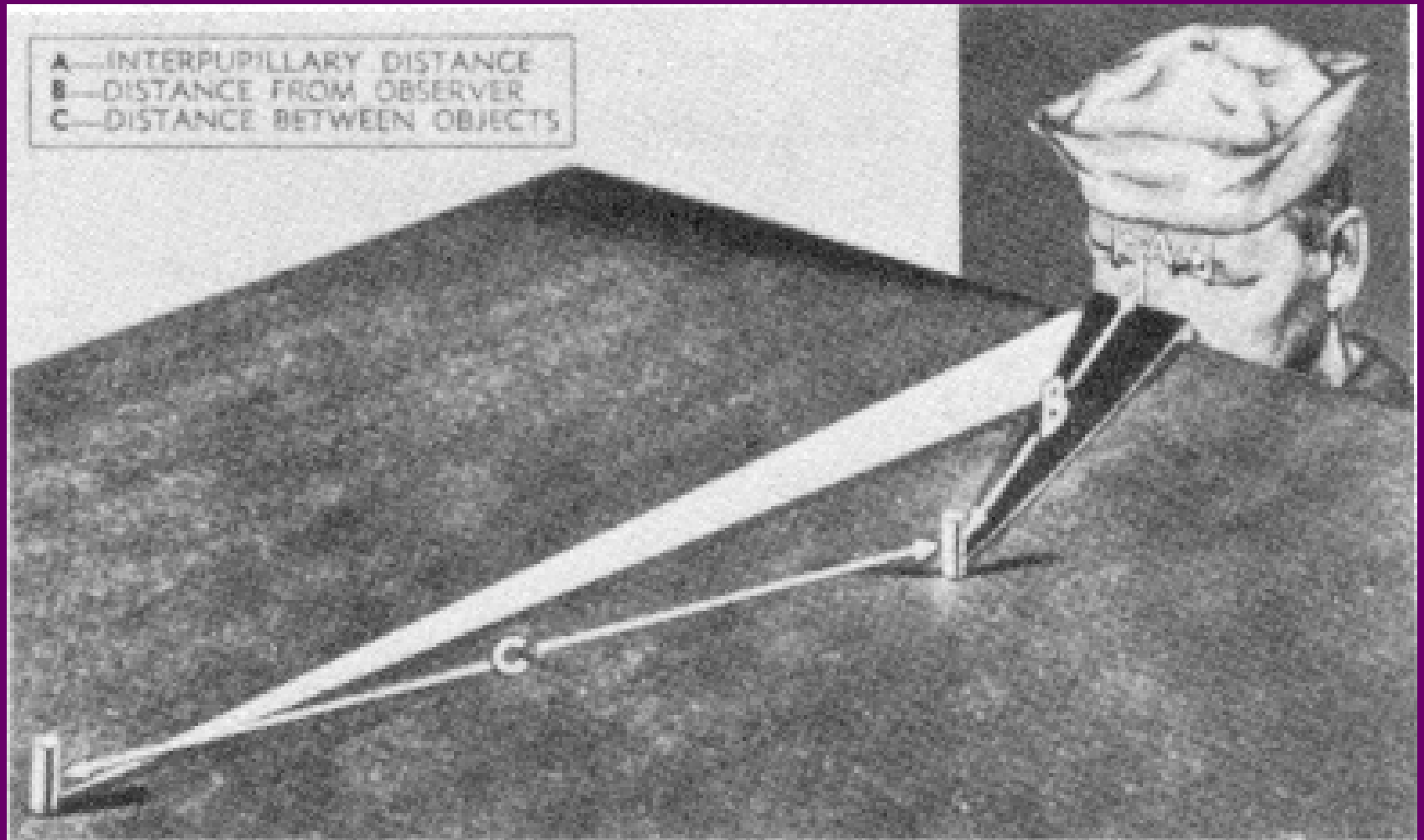
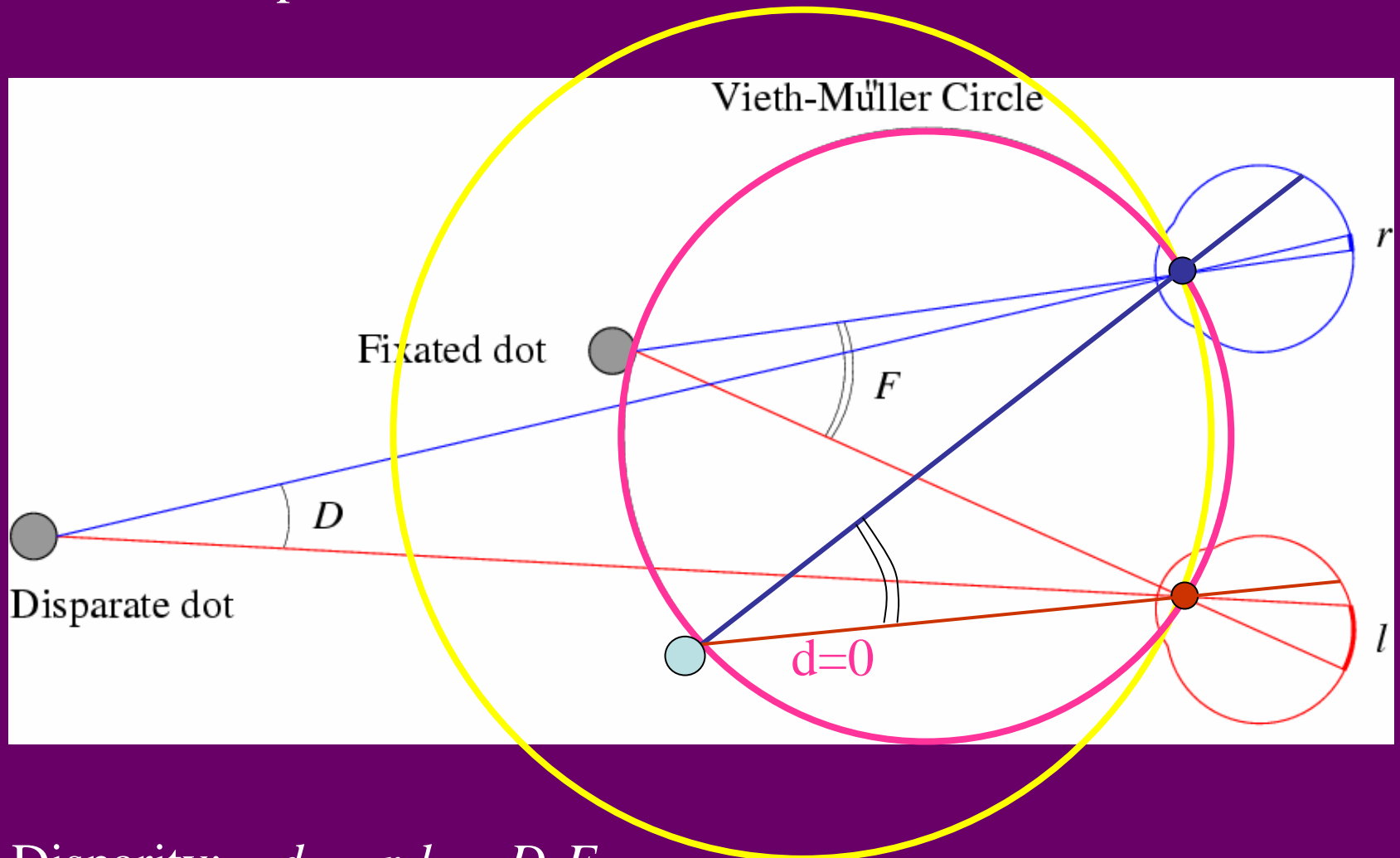


Figure from US Navy Manual of Basic Optics and Optical Instruments, prepared by Bureau of Naval Personnel. Reprinted by Dover Publications, Inc., 1969.

Human Stereopsis: Reconstruction



Disparity: $d = r-l = D-F.$

$d < 0$

In 3D, the horopter.

Human Stereopsis: Reconstruction

What if F is not known?

Helmoltz (1909):

- There is evidence showing the vergence angles cannot be measured precisely.
- Humans get fooled by bas-relief sculptures.





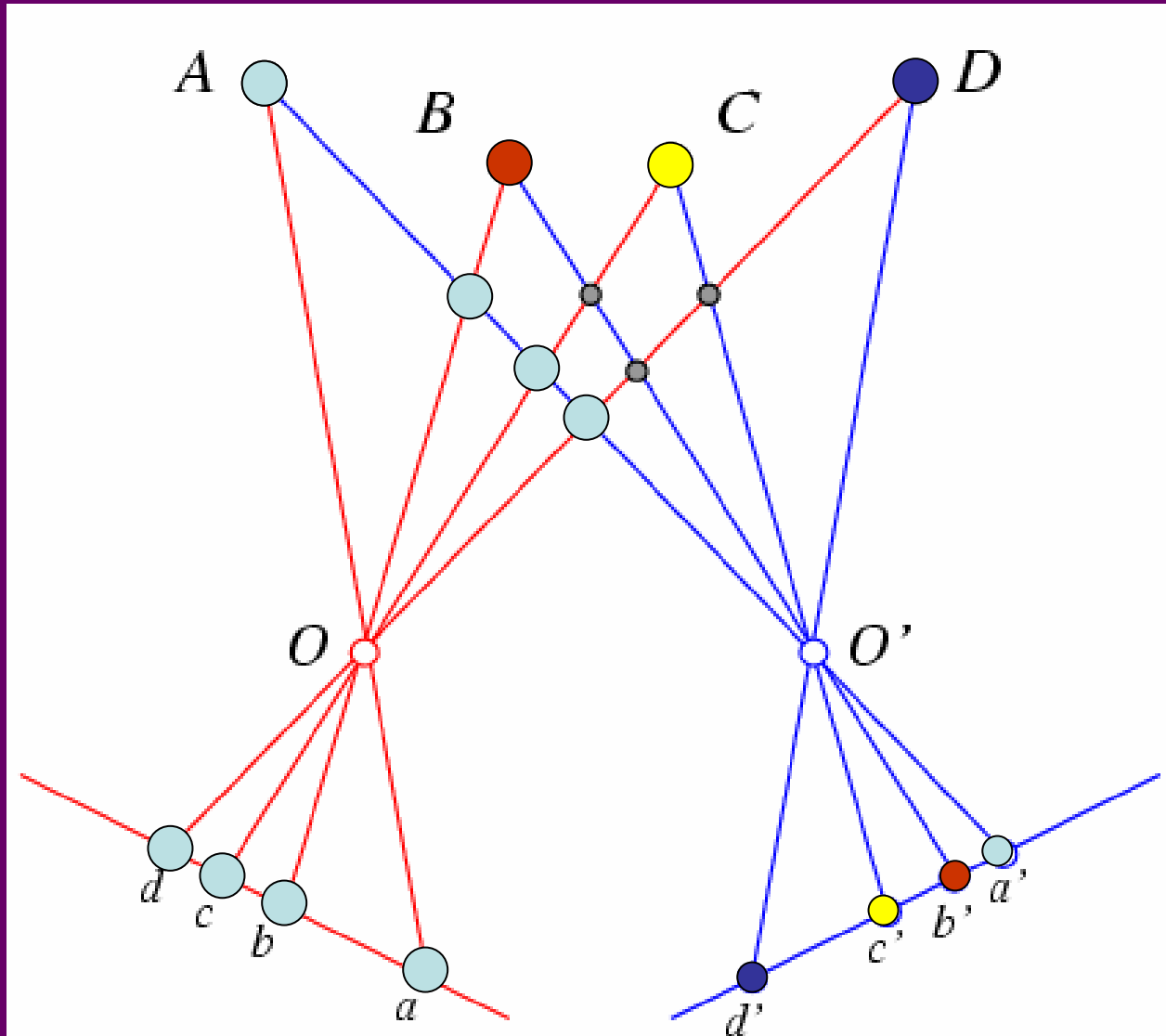
Human Stereopsis: Reconstruction

What if F is not known?

Helmoltz (1909):

- There is evidence showing the vergence angles cannot be measured precisely.
- Humans get fooled by bas-relief sculptures.
- Relative depth can be judged accurately.

(Binocular) Fusion

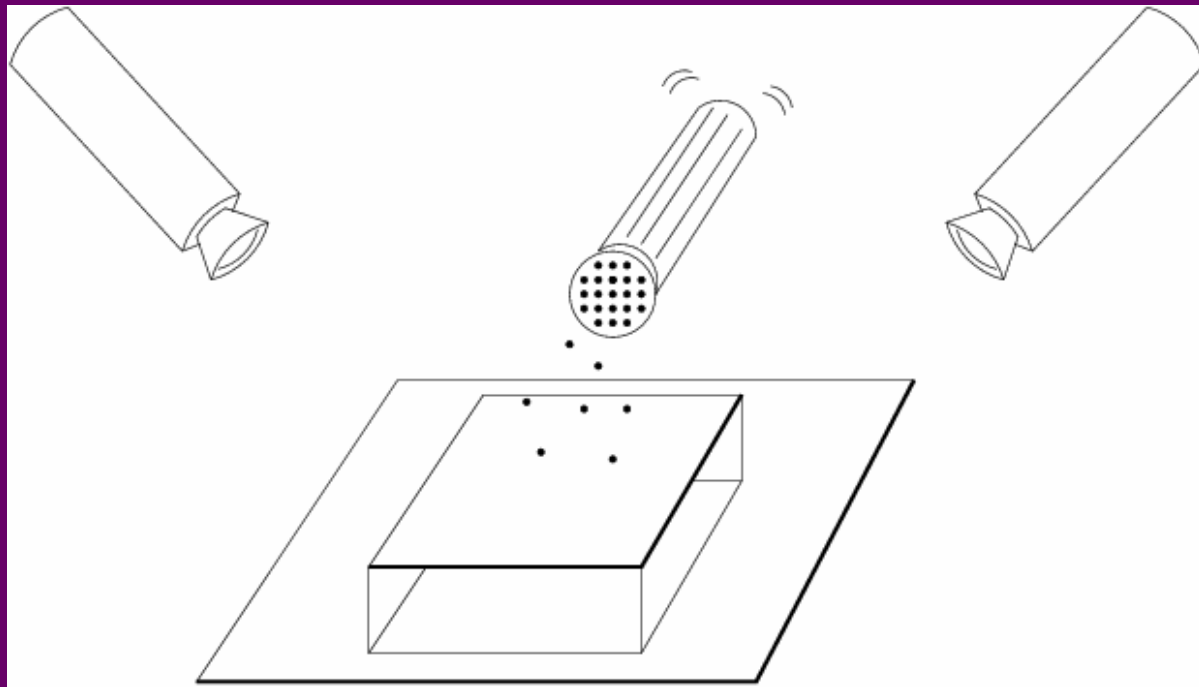


Human Stereopsis: Binocular Fusion

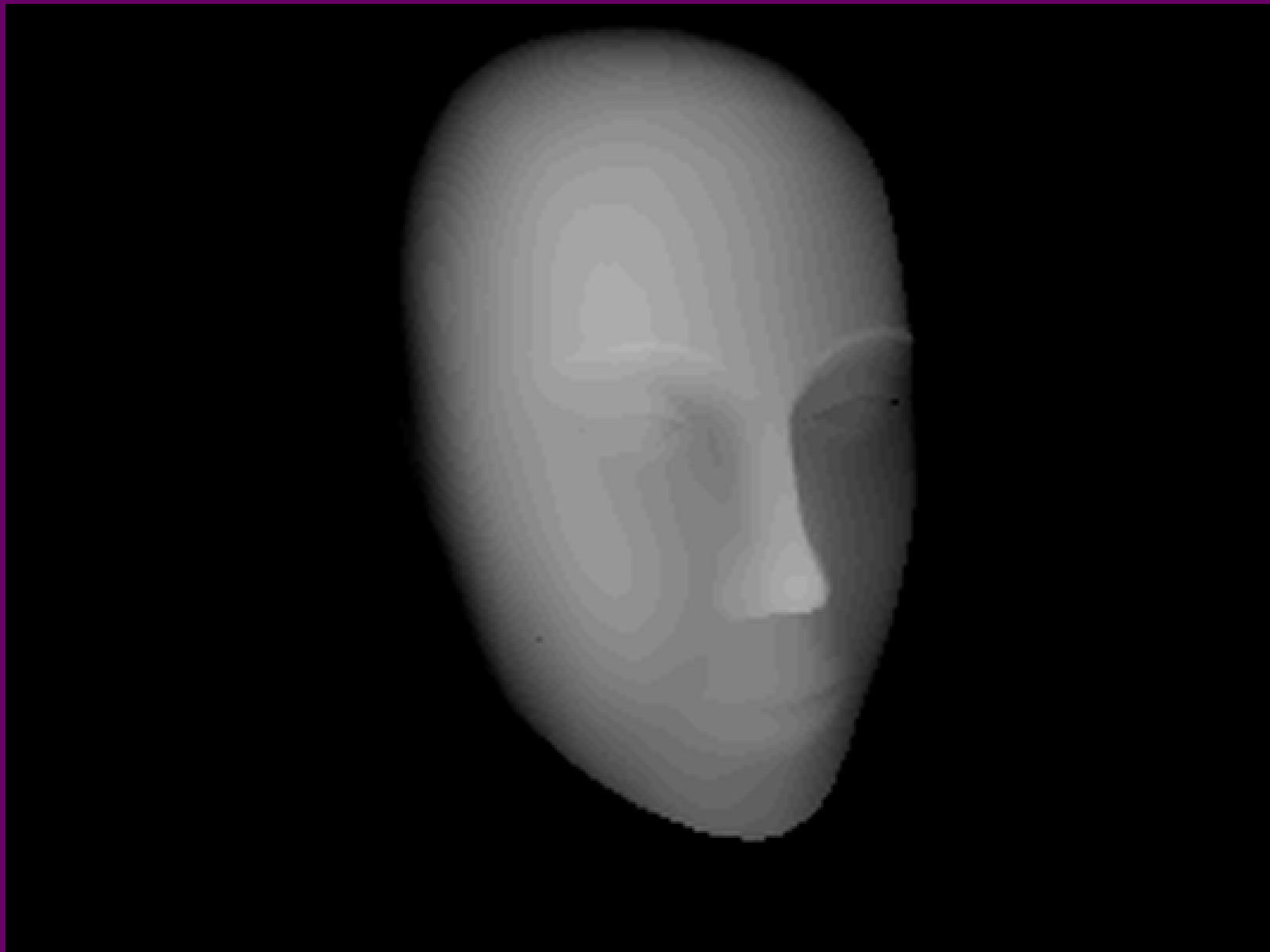
How are the correspondences established?

Julesz (1971): Is the mechanism for binocular fusion a monocular process or a binocular one??

- There is anecdotal evidence for the latter (camouflage).



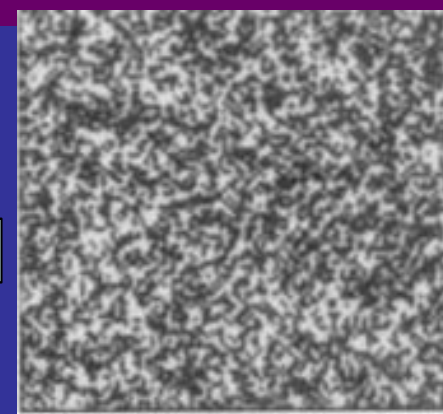
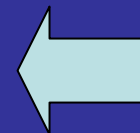
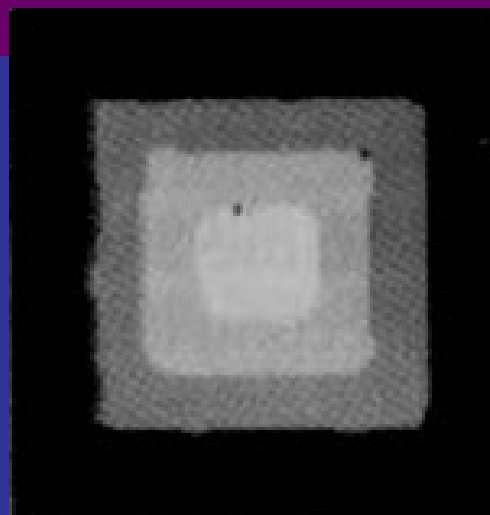
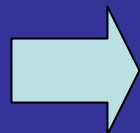
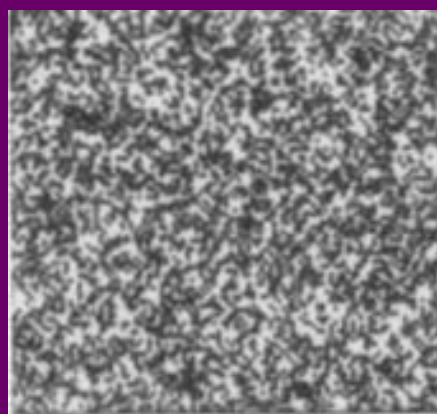
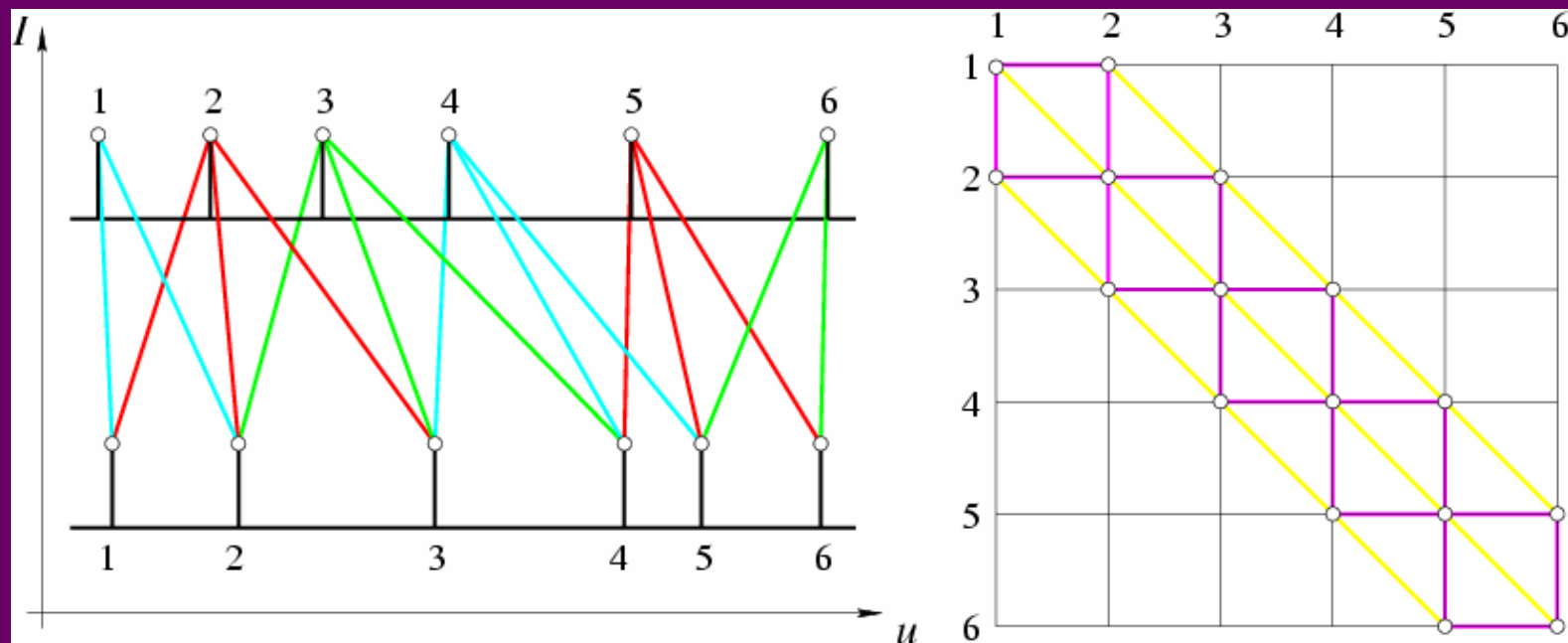
- Random dot stereograms provide an objective answer



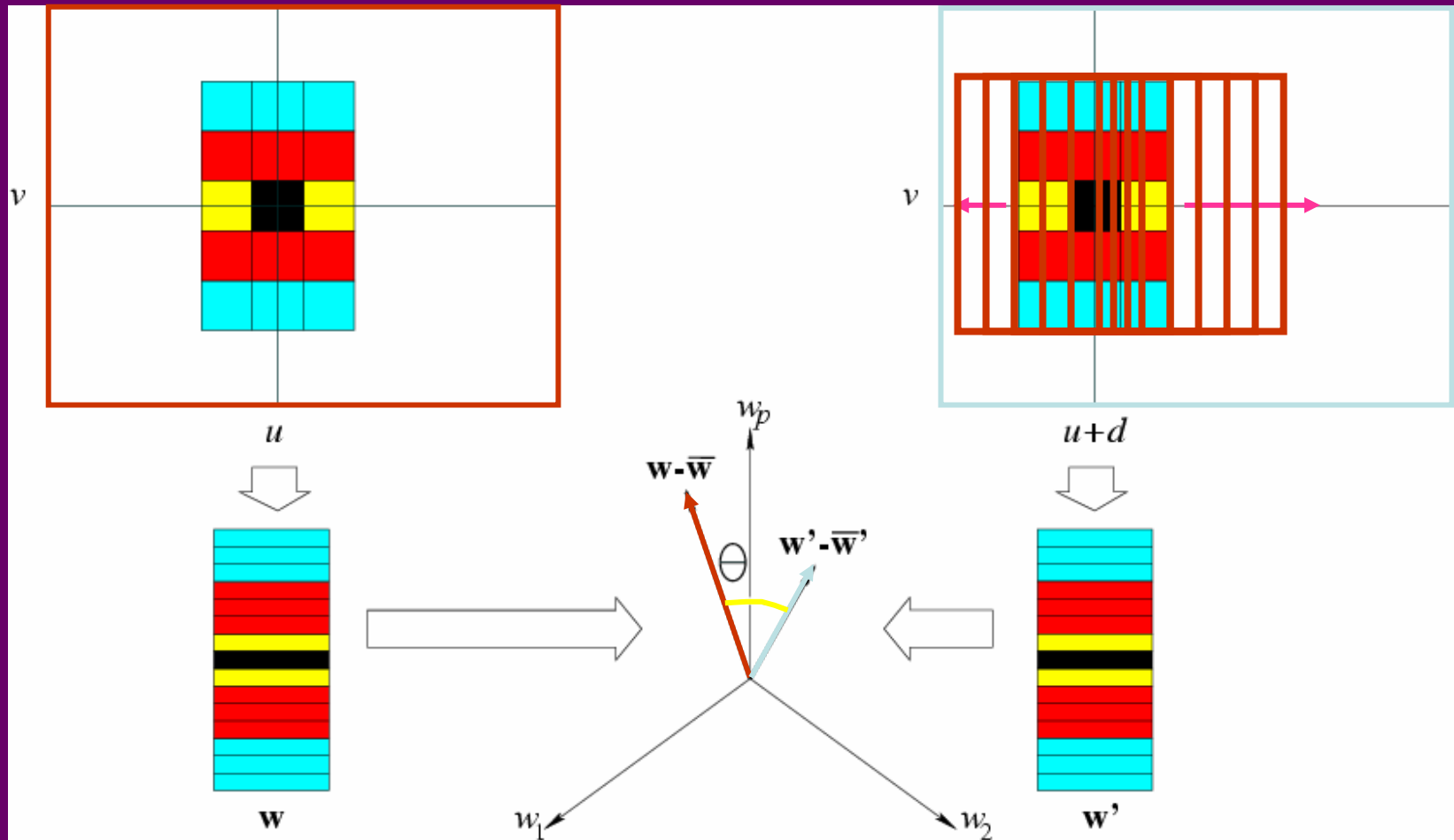
Some fun for yourself

- Google “random dot stereograms” for some fun examples
- For example:
 - <http://www.tanos.co.uk/gallery/randomdot/>

A Cooperative Model (Marr and Poggio, 1976)



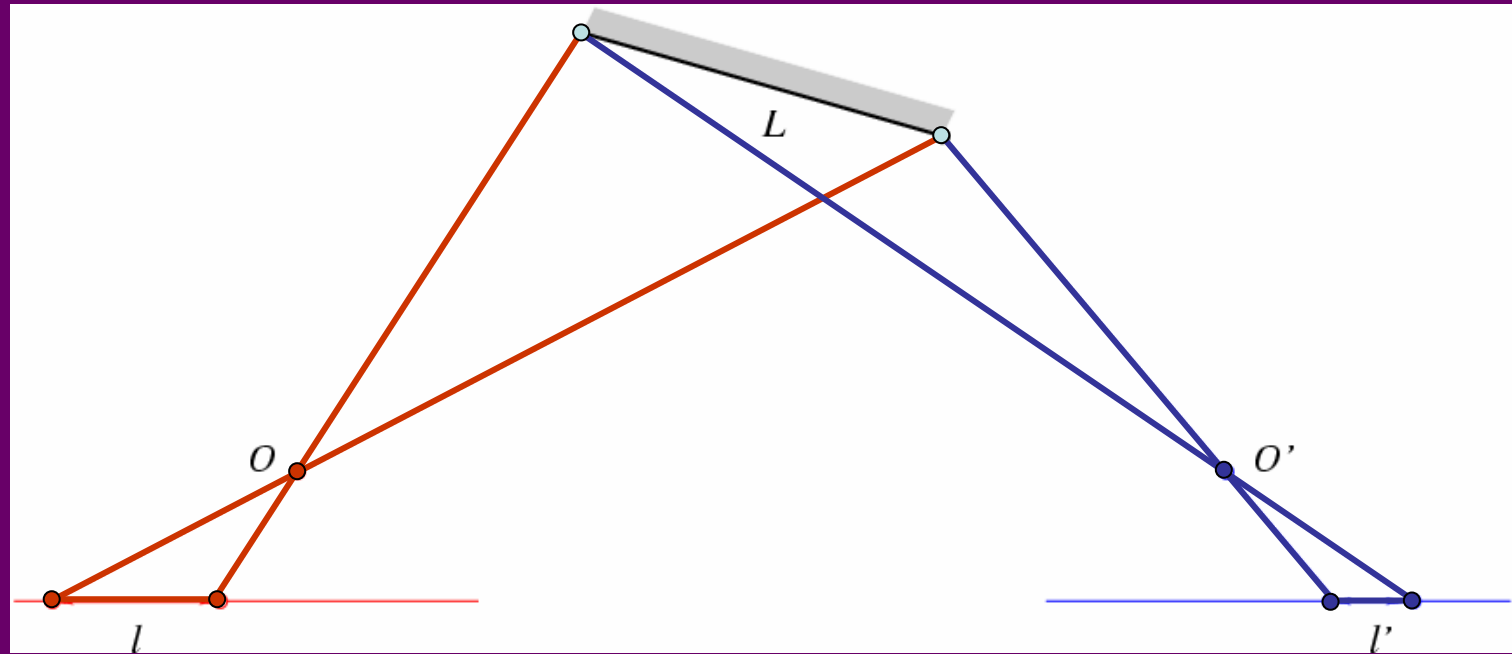
Correlation Methods (1970--)



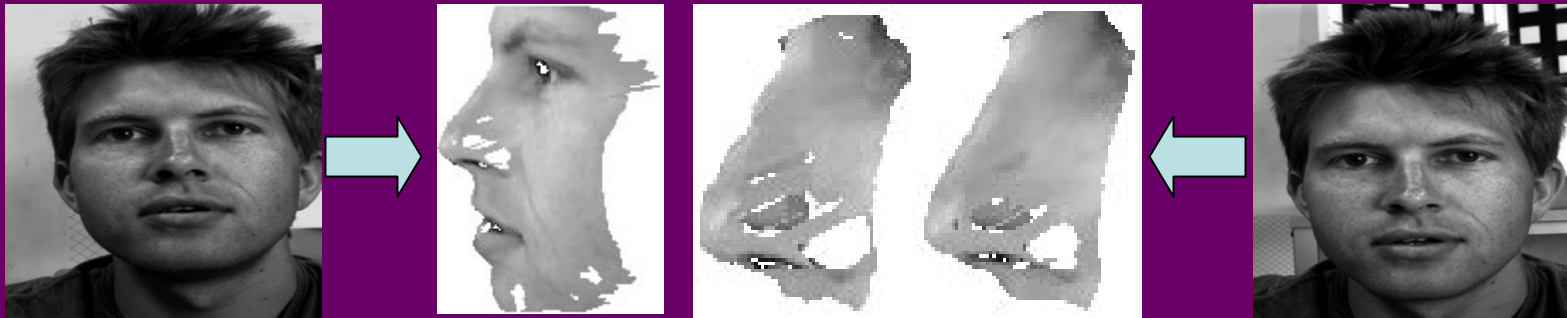
Slide the window along the epipolar line until $w \cdot w'$ is maximized.

Normalized Correlation: minimize θ instead. \longleftrightarrow Minimize $|w - w'|$.²

Correlation Methods: Foreshortening Problems

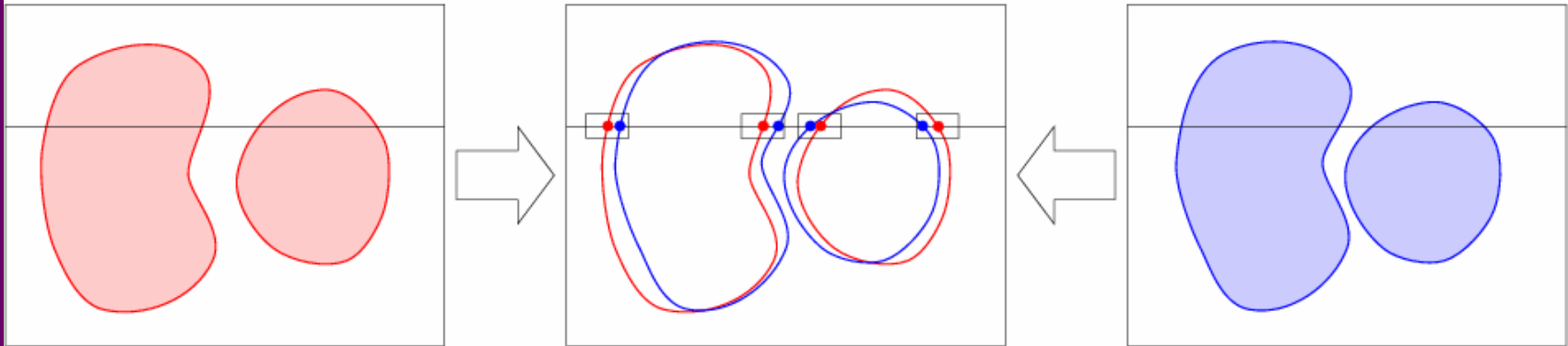


Solution: add a second pass using disparity estimates to warp the correlation windows, e.g. Devernay and Faugeras (1994).

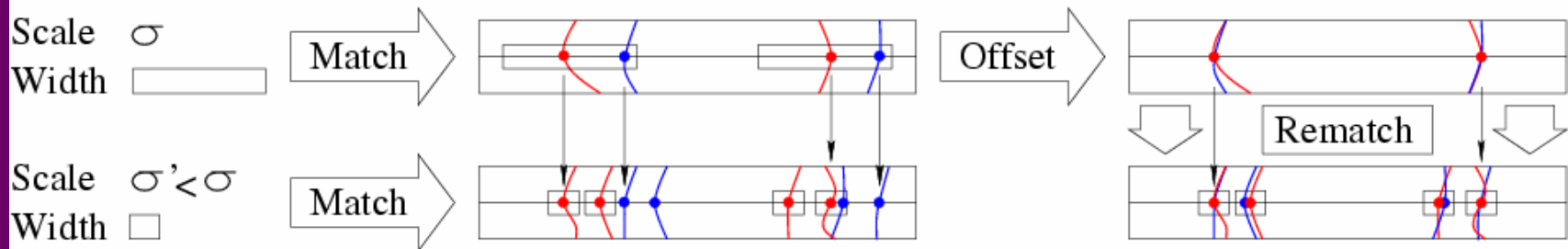


Multi-Scale Edge Matching (Marr, Poggio and Grimson, 1979-81)

Matching zero-crossings at a single scale



Matching zero-crossings at multiple scales



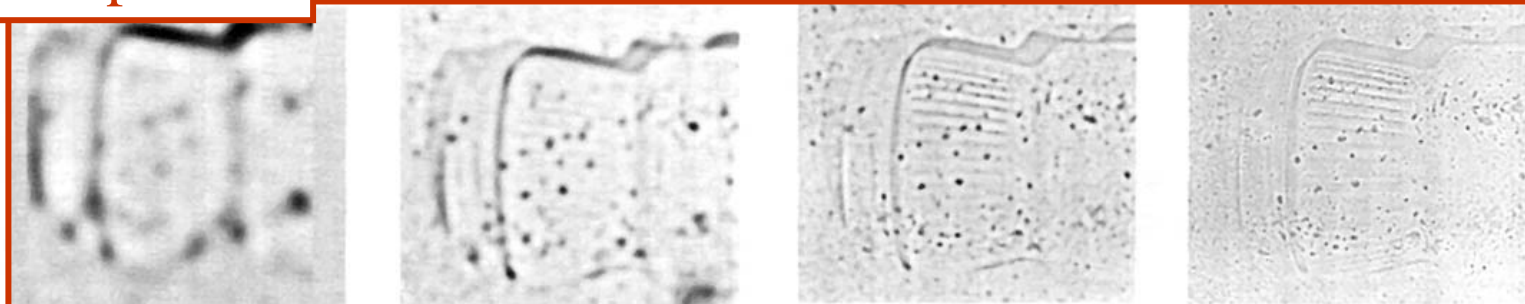
- Edges are found by repeatedly smoothing the image and detecting the zero crossings of the second derivative (Laplacian).
- Matches at coarse scales are used to offset the search for matches at fine scales (equivalent to eye movements).

Multi-Scale Edge Matching (Marr, Poggio and Grimson, 1979-81)

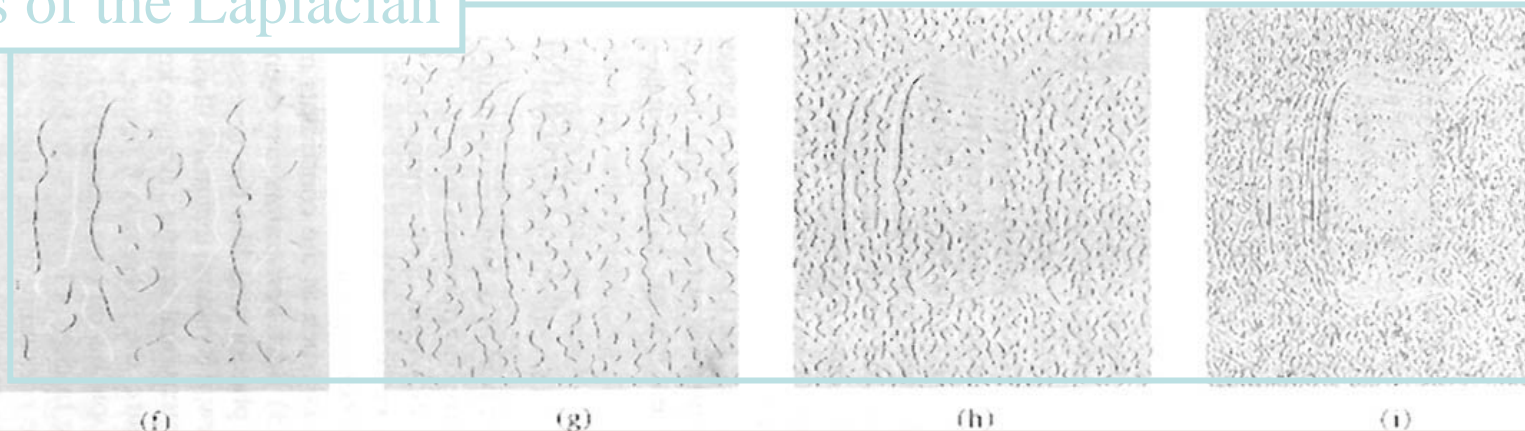


One of the two
input images

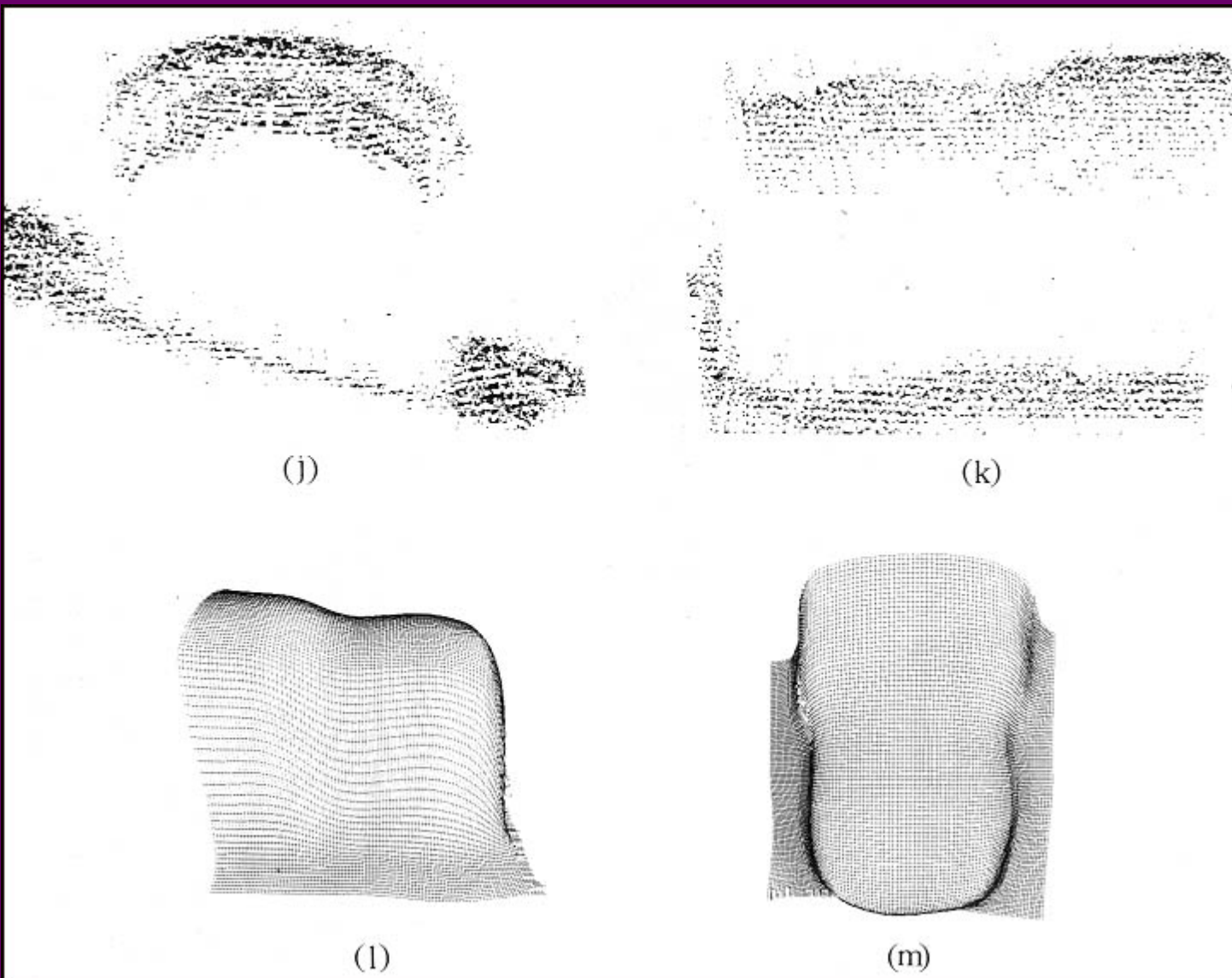
Image Laplacian



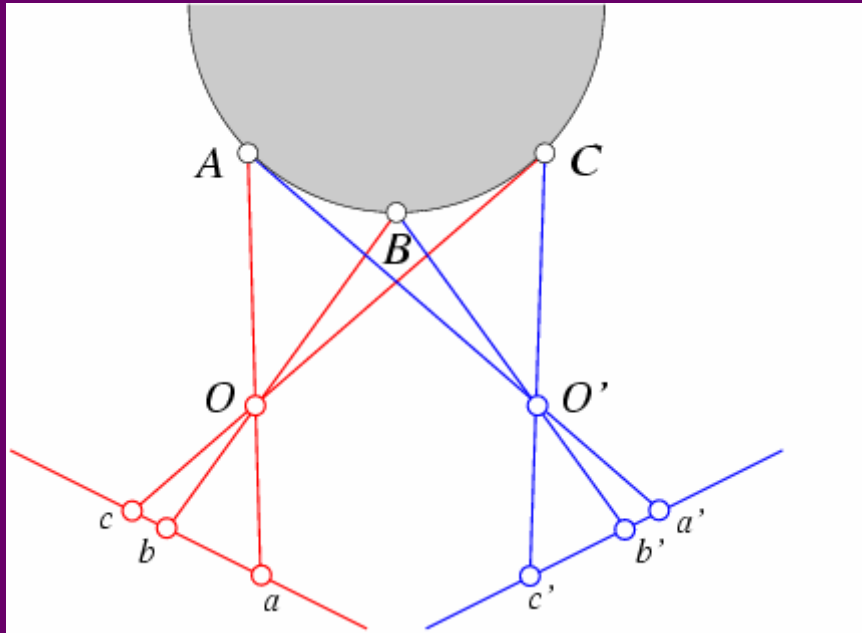
Zeros of the Laplacian



Multi-Scale Edge Matching (Marr, Poggio and Grimson, 1979-81)



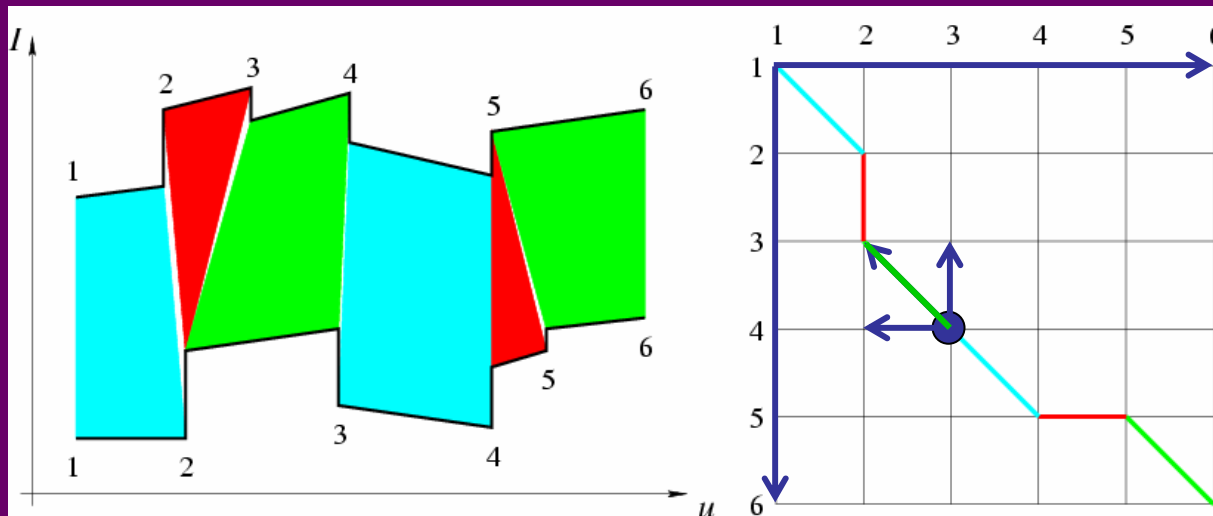
The Ordering Constraint



In general the points are in the same order on both epipolar lines.

But it is not always the case..

Dynamic Programming (Baker and Binford, 1981)

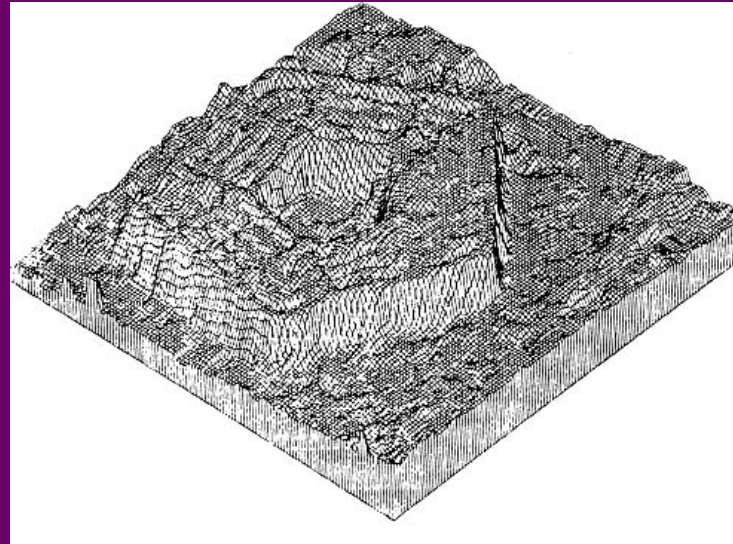


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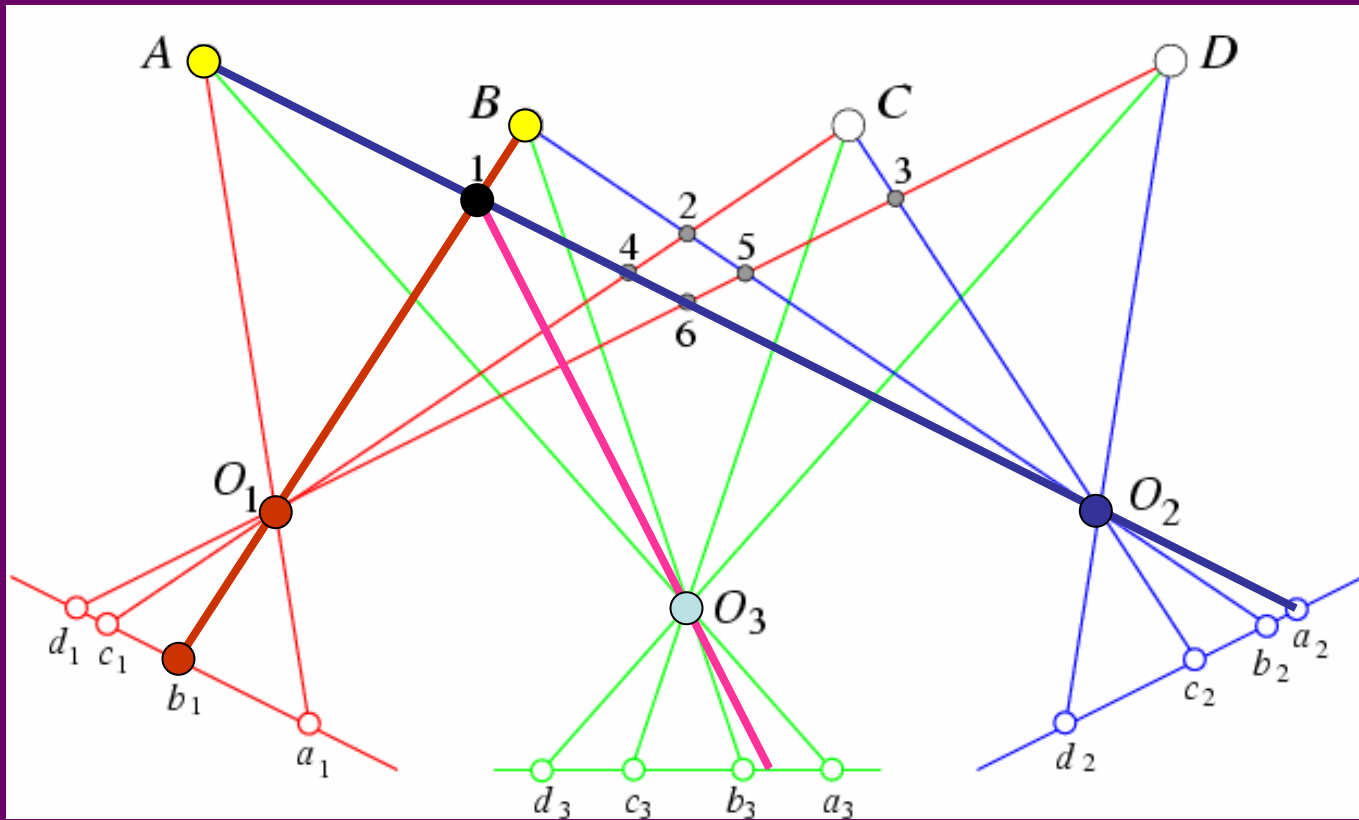
% Loop over all nodes  $(k, l)$  in ascending order.
for  $k = 1$  to  $m$  do
  for  $l = 1$  to  $n$  do
    % Initialize optimal cost  $C(k, l)$  and backward pointer  $B(k, l)$ .
     $C(k, l) \leftarrow +\infty$ ;  $B(k, l) \leftarrow \text{nil}$ ;
    % Loop over all inferior neighbors  $(i, j)$  of  $(k, l)$ .
    for  $(i, j) \in \text{Inferior - Neighbors}(k, l)$  do
      % Compute new path cost and update backward pointer if necessary.
       $d \leftarrow C(i, j) + \text{Arc - Cost}(i, j, k, l)$ ;
      if  $d < C(k, l)$  then  $C(k, l) \leftarrow d$ ;  $B(k, l) \leftarrow (i, j)$  endif;
    endfor;
  endfor;
endfor;

% Construct optimal path by following backward pointers from  $(m, n)$ .
 $P \leftarrow \{(m, n)\}$ ;  $(i, j) \leftarrow (m, n)$ ;
while  $B(i, j) \neq \text{nil}$  do  $(i, j) \leftarrow B(i, j)$ ;  $P \leftarrow \{(i, j)\} \cup P$  endwhile.
    
```

Dynamic Programming (Ohta and Kanade, 1985)



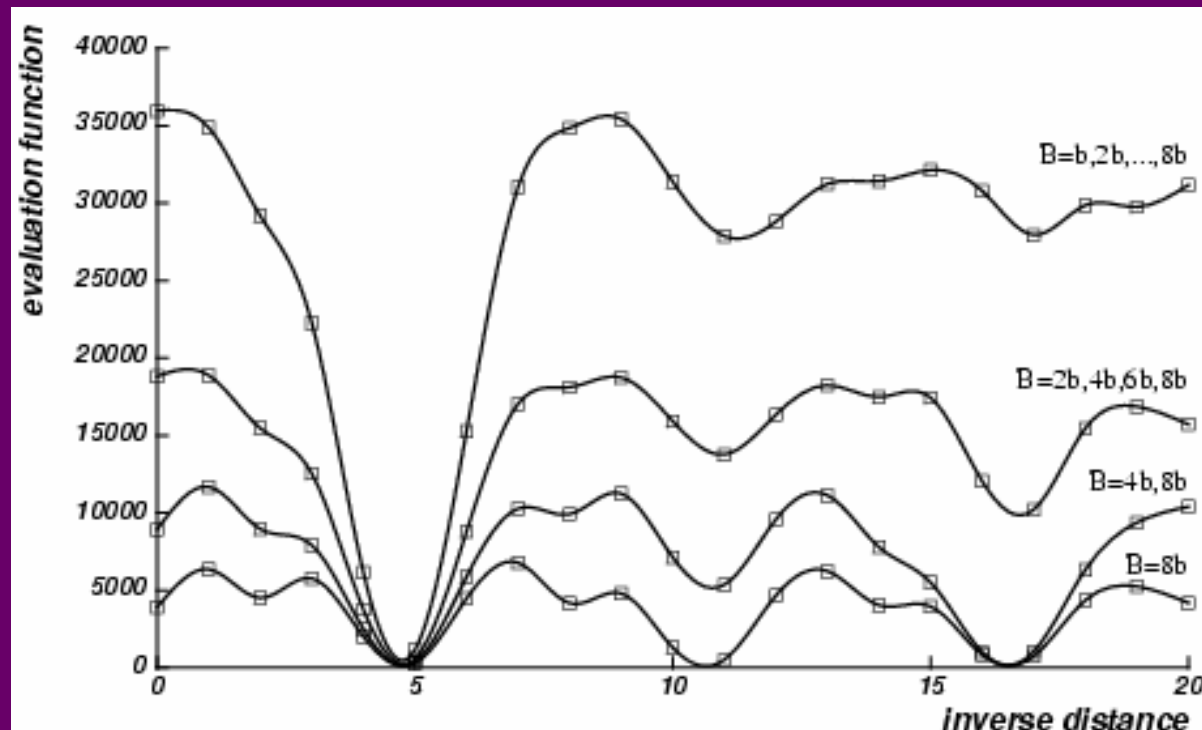
Three Views



The third eye can be used for verification..

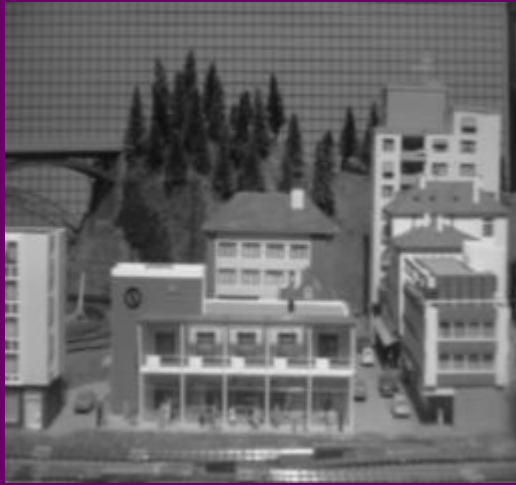
More Views (Okutami and Kanade, 1993)

Pick a reference image, and slide the corresponding window along the corresponding epipolar lines of all other images, using **inverse depth** relative to the first image as the search parameter.



Reprinted from "A Multiple-Baseline Stereo System," by M. Okutami and T. Kanade, IEEE Trans. on Pattern Analysis and Machine Intelligence, 15(4):353-363 (1993). \copyright 1993 IEEE.

Use the sum of correlation scores to rank matches.



I1



I2



I10

