




STEREO CORRESPONDENCE

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Vision Guided Control
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Computer Stereo: Applications



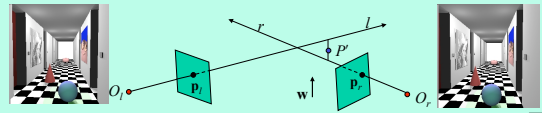
Medical diagnostics, architecture, autonomous navigation, cartography, biometrics, material science, etc...

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Triangulation from Projections

- Point P : projecting into the pair of corresponding points p_l and p_r
- Ideally, P at the **intersection** of the two optical rays: respectively, from O_l through p_l and from O_r through p_r
 - Due to approximate cameras' parameters and imprecise image locations: the actual two rays may not actually intersect in space
 - **Estimated intersection** : the point on minimum distance from both rays



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Three Basic Cases

Depending on the amount of a priori cameras' knowledge:

1. **Both intrinsic and extrinsic parameters**: the unique reconstruction of a 3-D scene by **triangulation**
2. **Only the intrinsic parameters**: a 3-D scene is still reconstructed and also the extrinsic parameters are estimated, but **up to an unknown scaling factor**
3. **Only pixel correspondences**: a 3-D scene is still reconstructed, but **up to an unknown, global projective transformation**

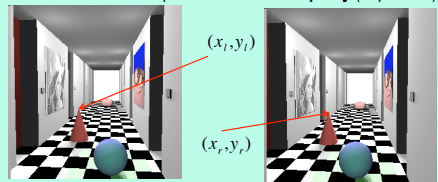
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Matching Fundamentals

Computational stereo: 3-D coordinates of visible points from 2-D coordinates of the corresponding image pixels

- Difference of these pixel coordinates: **disparity** (or parallax)



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Matching Fundamentals

- Generally, horizontal (x -) and vertical (y -) disparities
 - x -disparity: difference $d_{x,y} = x_l - x_r$ between the corresponding x -coordinates
 - y -disparity: difference $\delta_{x,y} = y_l - y_r$ between the corresponding y -coordinates
 - Usually, a horizontal stereo baseline and small y -disparities
 - *Canonical stereo geometry (an epipolar pair)*: no y -disparities

Stereo image matching: to find disparities for all visible 3D points in a stereo pair

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Stereo Correspondence

Disparity Map

Continuous vector-valued function $\mathbf{d}(x,y) = [d_{x,y}, \delta_{x,y}]$

- Mapping the coordinates of binocularly visible points in one image to the corresponding coordinates in the other image:
 $(x,y) \leftrightarrow (x-d_{x,y}, y-\delta_{x,y}) \Leftrightarrow g_{L;x,y} \Leftrightarrow g_{R;x-d_{x,y}, y-\delta_{x,y}}$
- Mapping is undefined for partially occluded points having no stereo correspondence

Search region: for a position (x,y) in the left image \Leftrightarrow a set of candidates to be explored in the right image:
 $\{(x',y') : x-d_{\max} \leq x' \leq x-d_{\min}; y-\delta_{\max} \leq y' \leq y-\delta_{\min}\}$

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“Corridor”: Disparity Map

Left image of a stereo pair

Right image of a stereo pair

Grey-coded disparity map

Cyclopean image of the 3-D scene

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“Artificial Rock”: Disparity Map

Left image of a stereo pair

Right image of a stereo pair

Grey-coded disparity map

Cyclopean image of the 3-D scene

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“Alex”: Disparity Map

Upper image of a vertical stereo pair

Bottom image of a vertical stereo pair

Grey-coded disparity map

Cyclopean image of the 3-D face

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Ill-posed Stereo Problem

- Inverse optical problem - no unique solution!

Visibility:
 B, C, E - binocular
 A - monocular (L)
 D - monocular (R)

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Symmetric Canonical Geometry

- Disparity map: a set of epipolar profiles
- Points of each profile and corresponding points along the conjugate scan-lines in images have the same y -coordinates
- Only the signals along these scan-lines have to be matched

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Stereo Correspondence

Symmetric Canonical Geometry

- Symmetric coordinates: $[X, y, Z]^T \Leftrightarrow [x_L, y], [x_R, y]$
 - Disparity: $d = x_L - x_R = bf/Z$
- Cyclopean image / disparity map: $(X, y, Z) \rightarrow (x, d)$

Symmetric (x, d) -coordinates:

$$\begin{cases} x = (x_L + x_R)/2 \\ d = x_L - x_R \end{cases} \Leftrightarrow \begin{cases} x_L = x + d/2 \\ x_R = x - d/2 \end{cases}$$

$x_L, x_R = 0, 1, 2, \dots$
 $x = 0, \frac{1}{2}, 1, 1\frac{1}{2}, 2, \dots$
 $d = \dots, -2, -1, 0, 1, 2, \dots$

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Symmetric Canonical Geometry

- B - binocular visibility
- ML - monocular visibility (left image only)
- MR - monocular visibility (right image only)

Integer cyclopean coordinates x - even disparities d
 Half-int cyclopean coordinates x - odd disparities d

Basic restriction: a continuous single surface in cyclopean space

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Symmetric Canonical Geometry

x	0	1	2	2.5	3.5	4	4.5	5.5	6	7
d	0	0	0	1	1	2	1	1	0	0
x_L	0	1	2	3	4	5	5	6	6	7
x_R	0	1	2	2	3	3	4	5	6	7

Due to assumed single continuous surface

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Equivalent Surface Profiles

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Sources of Ill-Posedness

- Multiple surfaces, partial occlusions, uniform texture...

Actual disjoint and equivalent continuous surface profiles

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Sources of Ill-Posedness

Most of 3D scenes do not conform to the single surface assumption and thus to the ordering constraint

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Stereo Correspondence

Visibility of 3-D Points

Multiple disjoint and continuous 3-D profiles: ● binocularly visible points; + : invisible (occluded) points; ○ partially occluded points; ○ transparent points

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Real Vs. Equivalent Profiles

right scan-line piecewise constant signal

left scan-line piecewise constant signal

Continuous profile reconstruction:
 Extreme disjoint variant:
 Real profile: ———

equivalent signal matches

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Distribution of signal differences

true signal differences

Signal-to-signal matching along the same scan-line 173; stereo pair "Isukuba"

Grey-coded signal differences

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3-D Reconstruction

- Because of ill-posedness, it is impossible to reconstruct precisely the original 3D scene from a stereo pair
 - Goal of stereo matching is therefore more limited and more practical: to bring the reconstructed surfaces close enough to those perceived visually or with the photogrammetric tools
- Due to a multiplicity of visually observed scenes - only very general prior knowledge to constrain optical 3-D surfaces under reconstruction
 - E.g. expected smoothness, curvature, discontinuities, etc

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General Matching Constraints

Reflecting intrinsic properties of stereo viewing and a 3-D scene

Epipolar constraint: 1-D search along the conjugate scan-lines

- Rectified stereo pairs reduced to the canonical stereo geometry
- Reduced search region; excluded false matches across the scan-lines

Uniqueness constraint: every pixel in one stereo image has at most one corresponding pixel in the other image

- Every visible 3-D point is observed either binocularly or only monocularly
- Monocular observation: *partial occlusion* (no stereo correspondence)

Disparity range $[d_{min}, d_{max}]$ is typically known for a 3-D scene

- Reduced search region; excluded false matches outside the range

Continuity constraint: smooth surfaces except for object boundary

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General Matching Constraints

Ordering constraint: the same order of corresponding points along the conjugate scan-lines - but only for a *continuous single* profile!

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Simplifying Constraints

- **Equal corresponding signals**
 - Lambertian (direction-independent) reflection of 3-D surfaces
 - Simple matching scores like $|g_{L(x,y)} - g_{R(x,y)}|$ or $(g_{L(x,y)} - g_{R(x,y)})^2$
- **Frontal parallel surfaces**
 - Area-based correlation matching: constant disparity and no occlusions over the matching windows
- **Similarity of features**
 - Feature-based matching: similar and mutually consistent groups of corresponding features in both images

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Stereo Correspondence

- Similarity between (dissimilarity of) stereo images
 - Under their relative photometric and geometric distortions
 - Different projective views, camera noise, occlusions, ...
- **Photometric distortions**
 - Non-uniform reflection of observed 3-D surface points in different directions
 - Non-uniform and noisy transfer factors over a field-of-view (FOV) of every stereo camera
 - Spatially variant contrast and offset deviations between corresponding signals in stereo images

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Stereo Correspondence

- **Geometric distortions** due to projection of a 3-D scene onto the two image planes
 - Spatially variant disparities of the corresponding points
 - Corresponding regions in stereo images may differ in positions, scales, and orientations
 - Partial occlusions (monocular visibility) of some 3-D points
 - Such regions have no stereo correspondence in principle
 - A *single continuous visible surface*: the images preserve the natural x - and y -order of binocularly visible points (BVP)

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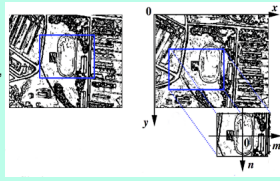
Stereo Correspondence

- Stereo matching techniques differ in:
 - Which image similarities (dissimilarities) are measured
 - Which relative image distortions are taken into account
 - Which constraints / regularising heuristics are involved
 - How a stereo pair is matched as a whole
- Image signals: grey values, colours (RGB, HSV, etc)
- Feature-based vs. intensity-based stereo matching
 - Dense and sparse disparity maps

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Feature-based Matching

- Specific area, linear, and point features individually found in each image of a stereo pair
 - Edges, corners, T- junctions, isolated local shapes, etc
 - Only features are tested for similarity
 - Feature matching usually cannot produce dense disparity maps




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Intensity-based Matching

- Similarity (or dissimilarity) between the images in terms of the initial signals for all binocularly visible 3-D points
 - Signals: grey levels, colours, or multi-band signatures
 - Math models to relate optical signals from the observed 3-D points to the image signals in the corresponding pixels
 - Similarity (dissimilarity $D(d | g_L, g_R)$) between the corresponding image pixels or regions is derived from the model
 - The similarity score has to be invariant to relative geometric and photometric image distortions the model accounts for


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Local and Global Optimisation

- 3-D reconstruction: search for the max similarity (or min dissimilarity) between the corresponding regions / pixels
 - Similarity measure accounts for admissible image distortions and includes regularising constraints:
 - E.g. to deal with partial occlusions or multiple equivalent optima
 - For a single continuous surface: visibility and ordering constraints
- Scenarios of reconstructing a 3-D scene: $d^* = \max_d \{D(d | g_L, g_R)\}$
 - Exhausting variants of visible surfaces by **global optimisation**
 - Independent selection of each 3-D point by **local optimisation**
 - Successive search for each next small surface patch by **local optimisation** in order to add it to the already found surface


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Local Optimisation

- **Pros:** simple computations; easily takes account of both x - and y -disparities of the corresponding pixels
- **Cons** for independent selection of 3-D points: the found surfaces may violate visibility and continuity constraints
- **Cons** for guiding next search by the current surface: due to accumulation of local errors, the search regions after a few steps may become completely wrong
- **Cons** in both cases: it needs intensive on- or off-line editing of the reconstructed 3-D scene to fix errors


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Global Optimisation

- **Pros:** less sensitive to local errors due to constraints on the conjugate scan-lines or the entire stereo images
- **Cons:** generally, it is an **NP-hard** problem (due to 2-D constraints on disparities in the neighbouring points)
 - It is feasible only in particular cases when direct exhaustion of all the variants (with the exponential complexity) is avoided
 - Known approximate solutions are still too complex and thus too slow for processing large-size images of practical interest
 - Profile-by-profile reconstruction by 1-D dynamic programming is fast but takes no account of constraints across the profiles

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Global Optimisation

- Two popular tools for approximate global optimisation:
 - Graph min-cut iterative algorithms
 - Exact solution for a minimum cut / maximum flow problem
 - Approximation of stereo matching with a sequence of graph min-cuts
 - Disparity map: the constrained dissimilarity between two stereo images within a fixed factor from the global minimum
 - Loopy belief-propagation algorithms
 - Computing marginal posterior probabilities of constrained disparities for each surface point
 - Convergence on loop-less graphs (trees)
 - Under specific conditions, convergence on loopy graphs, too

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