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- Given a pixel in one image, we want to compute its correspondence to a pixel in the other image.
- Various techniques can be used to match pixels based on their local appearance.
- With additional information on positions and calibration data for cameras available.
- How can we exploit this information to reduce the number of potential correspondences and hence speed up the matching and increase its reliability?





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Image Rectification

 The resulting standard rectified geometry is employed in a lot of stereo camera setups and stereo algorithms, and leads to a very simple inverse relationship between 3D depth, Z, and disparities d,

 $d = f^*B/Z$

where f is the focal length (measured in pixels), B is the baseline, and

x' = x + d(x, y), y' = y

 describes the relationship between corresponding pixel coordinates in the left and right images







- Feature-based stereo matching algorithms.
- Extract a set of potentially matchable image locations using either interest operators or edge detectors, then search for corresponding locations using a patch-based metric
- Better approaches focus on first extracting highly reliable features and then use these as seed-points to grow additional matches















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Dense Correspondence

- More stereo matching algorithms today focus on dense correspondence
- Dense correspondence algorithms consist of a set of modules,
 - matching cost computation
 - Cost aggregation
 - Disparity computation and optimization
 - Disparity refinement.
- For example, SSD, normalised cross-correlation, rank transform, etc.
- Hierarchical (coarse-to-fine) algorithms
- More recently, robust measures, including truncated guadratics and *contaminated* Gaussians, have been proposed,
 - to limit the influence of mismatches during aggregation









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Other methods: Shape from X

- In addition to binocular disparity, shading, texture, and focus all play a role in how we perceive depth/shape.
- Shape from shading: as the surface normal changes, the apparent brightness changes as a function of the angle between the local surface orientation and the incident illumination
- Shape from texture: algorithms require a number of processing steps, including the extraction of repeated patterns or the measurement of local frequencies in order to compute local affine deformations, and a subsequent stage to infer local surface orientation.
- Shape from focus: the amount of blur increases as the object's surface moves away from the camera's focusing distance



Stereo Reconstruction Applied to Mars Images

- Introduction to Mars 3D imaging data
- The difference between planetary and general stereo reconstruction methods
- DTM and orthorectified image









The difference between planetary and general stereo reconstruction methods

- Camera modelling
- Map projection
- Pre-processing (least squares bundle adjustment, normalisation, filtering)
- Area based matching
- Initial disparity estimation
- Pyramidal, tiled processing
- DTM and orthorectified image





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Processing example and parameters

	# Pre-Processing / stereo_pprc	
 Interactive session 	# Pre-alignment options	
# Integer Correlation / stereo corr		
***************************************	# Available choices are (nowever not all are supported by all sessions); # NONE (Recommended for anything map projected)	
	# EPIPOLAR (Recommended for Pinhole Sessions)	,
# Select a cost function to use for initialization:	# HOMOGRAPHY (Recommended for ISIS and DG Sessions) alignment-method homography	
# 0 - absolute difference (fast)		ź
# 1 - squared difference (faster but usually bad)	# Intensity Normalization	,
# 2 - normalized cross correlation (recommended)	rorce-use-entire-range # use entire input range	
cost-mode 0	# Select a preprocessing filter:	
# Initialization step: correlation kernel size	# Ø - None	
corr-kernel 25 25	# 1 - Subtracted Mean	
	# 2 - Laplacian of Gaussian (recommended)	
# Initialization step: correlation search range	prefilter-mode Z	
2	# Kernel size (1-signa) for pre-processing	,
# Uncomment the following to use explicit search ranges. Otherwise, a	f	1
# value will be choosen for you.	# Recommend 1.4 px for Laplacian of Gaussian	i -
# corr-search -100 -100 100 100	# Recommend 25 px for Subtracted Mean prefilter-kernel-width 1.4	
# Subpixel Refinement / stereo_rfne		2

# Subpixel step: subpixel modes	Dimension State Constraints of Constants	£
4	# Post Filtering / stereo_fltr	÷.,
# 0 - disable subpixel correlation (fastest)	***************************************	
# 1 - parabola fitting (draft mode - not as accurate)	# Fill in holes on to 100 000 nivels in size with an innointing method	
# 2 - affine adaptive window, bayes EM weighting (slower, but much more accurate) subpixel-mode 2	# disable-fill-holes	2
	# Automatic "erode" low confidence pixels	,
# Subpixel step: correlation kernel size	rm-half-kernel 5 5	í.
subpixel-kernel 25 25	rm-min-matches 60	
Tana and the standard	rm-threshold 3	
	rm-cleanup-passes 1	
	# Triangulation / stereo tri	

	n an an ann a' an an an an ann an a' an ann an	
	# Size max of the universe in meters and altitude off the ground.	
	# Setting both values to zero turns this post-processing step off.	
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	rar-universe-rootus 0.0	









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Introduction to CASP-GO

- General workflow
- Maximum likelihood sub-pixel refinement
- Outlier rejection schemes
- Gotcha (Adaptive Least Square Correlation with region growing) densification
- Co-kriging grid-point interpolation
- ORI co-registration and DTM adjustment







- Sub-pixel disparity maps obtained from the refined initial disparity map is still not perfect
- Two obvious problems are unmatched areas (no disparity available, i.e. errors of omission) and mismatched areas (wrong disparity, i.e. errors of commission).
- Matching search kernel vs. search range?
- Larger kernel + smaller search range + reject mis-matches





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Outlier Rejection Schemes

- (a) disparity value differs than a threshold by a percentage of pixels in a kernel;
- (b) kernel with standard deviation higher than a threshold;
- (c) difference of the mean value of a kernel and neighbouring kernel is higher than a threshold;
- (d) kernel with a neighbouring kernel being rejected by a threshold percentage;
- (e) adjacent disparity values from (a), (b) and (c).





- (a) with given sub-pixel disparity values, retrieve seed tie-points (point correspondences) on the border (within 5-11 pixel width) according to the x and y translation (disparity); Version: 09/06/16
- (b) run ALSC on seed tie-points and store similarity value;
- (c) sort seed tie-points by similarity value;
- (d) a new matching is derived from the adjacent neighbours of the initial tie-point with highest similarity value;
- (e) if the new match is verified by ALSC then it is considered as seed tie-points in next growing;
- (f) this region growing process repeats from (c) to (e) until there are no more acceptable matches;
- (g) retrieve final disparity map after densification. i-Mars.eu









ORI co-registration and DTM adjustment

- HiRISE and CTX datasets are generally not coregistered with the HRSC ORI/DTM (DLR processed v50 products) and MOLA dataset.
- The Mutual Shape Adapted Scale Invariant Feature Transform (MSA-SIFT) algorithm
- Take HRSC ORI as reference image for CTX ORI coregistration and subsequent shift of CTX DTM according to the CTX ORI to HRSC ORI transformation





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