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# Stereo Vision – Correspondence

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# Problem Definition

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- Correspondence problem
  - What parts of left and right image are projections of the same point in the 3D scene
- Simple stereo configuration
  - Corresponding points are on same horizontal line
- Assumptions
  - Most scene points are visible from both regions
  - Corresponding image regions are similar
- Search problem
  - Given scene element on left image search for
  - What parts of left and right images are parts of same object?
- Two decisions
  - Which element to match
  - Which similarity measure to adopt

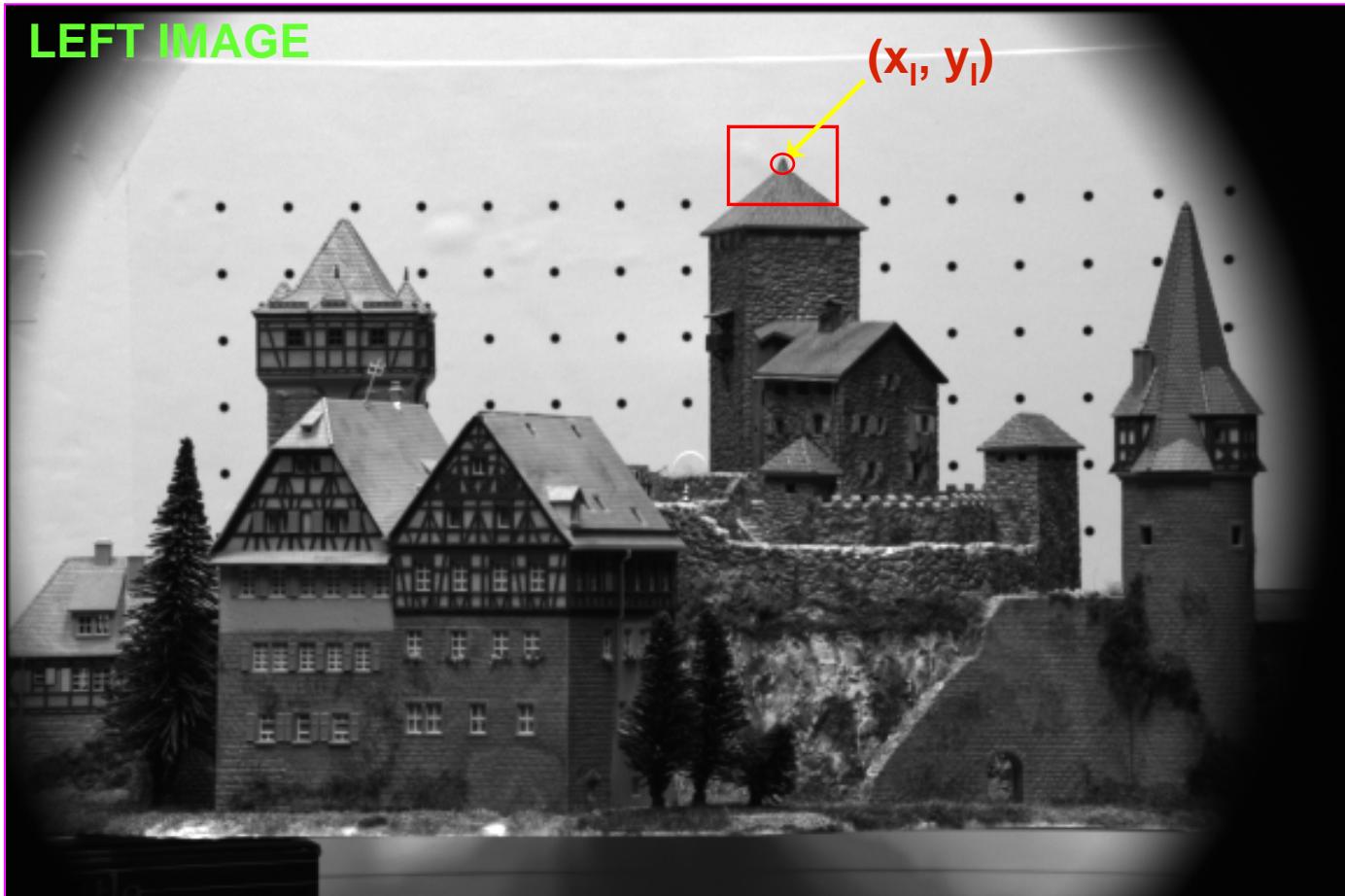
# Correspondence and Feature Methods

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- Two basic approaches
- Correlation methods
  - Apply to all image points
  - Elements are image windows of fixed size
  - Similarity measure is correlation between two windows in the left and right images
  - Corresponding element is window that maximizes similarity criterion within a search window
- Feature methods
  - Apply only to a sparse set of feature points
  - Narrows down feasible matches by using constraints
  - Geometric constraints
  - Analytic constraints – uniqueness and continuity

# Correlation Approach

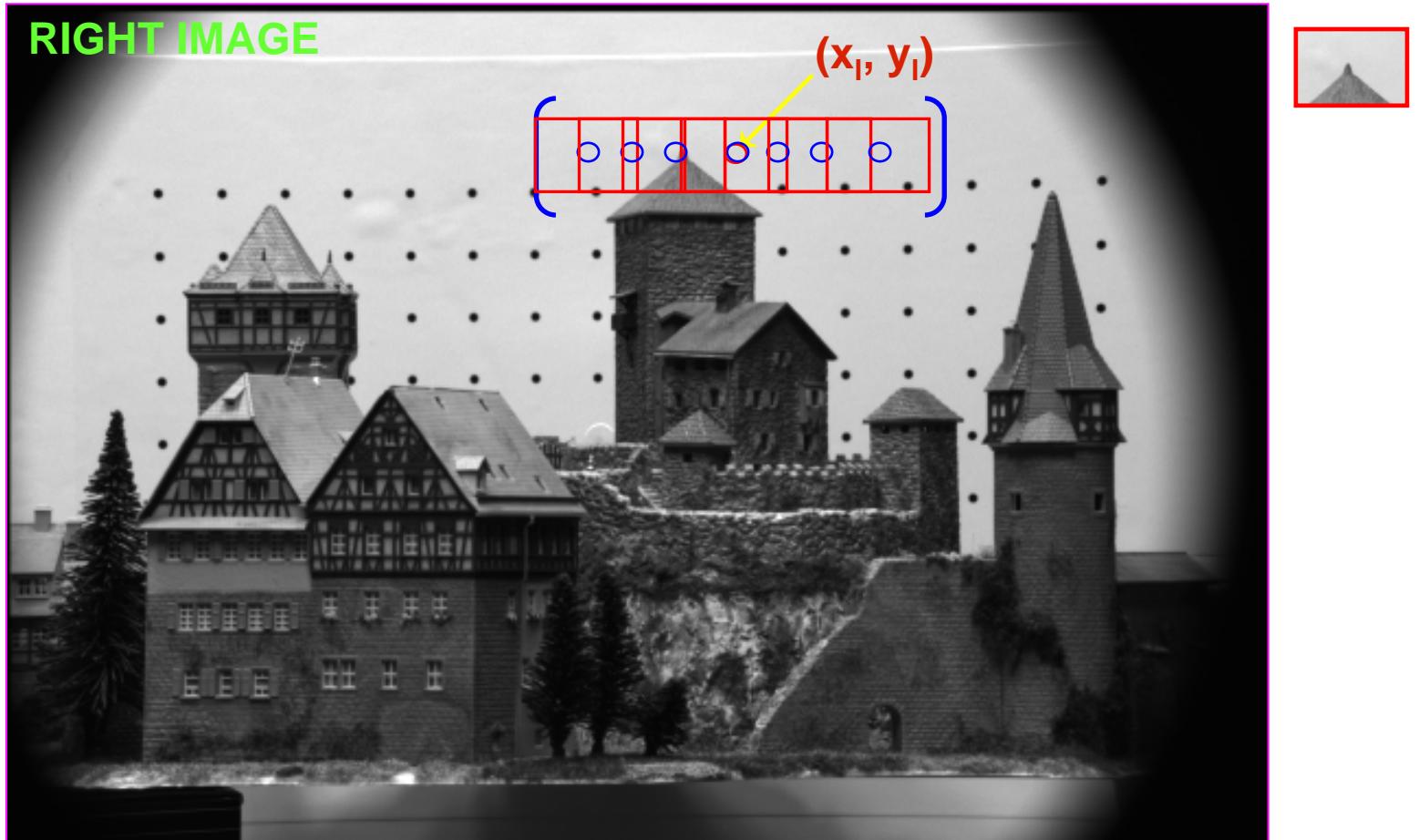
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For Each point  $(x_l, y_l)$  in the left image, define a window centered at the point

# Correlation Approach

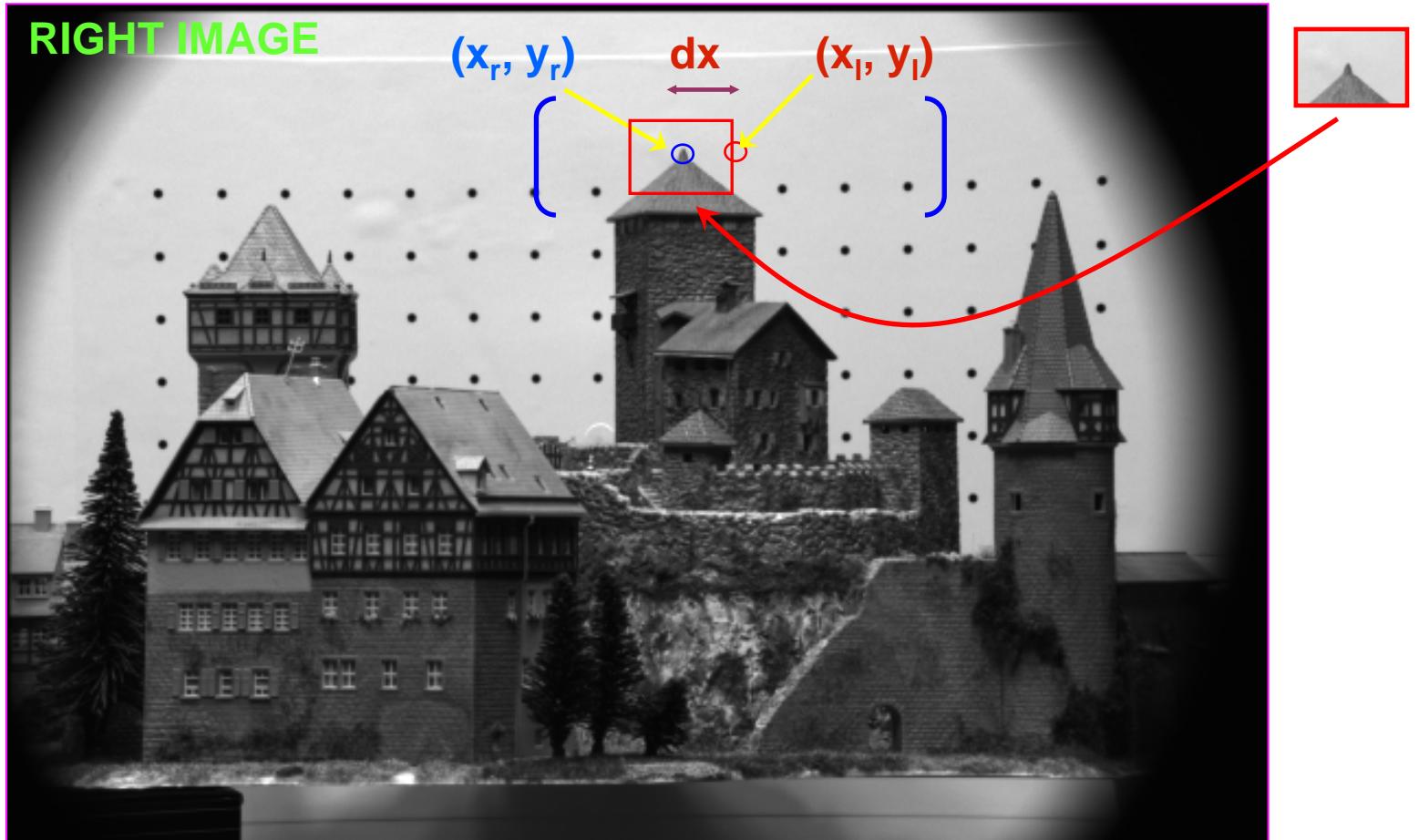
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... search its corresponding point within a search region  
in the right image

# Correlation Approach

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... the disparity ( $dx, dy$ ) is the displacement when the correlation is maximum

# Correlation Approach

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## Elements to be matched

- Image window of fixed size centered at each pixel in the left image

## Similarity criterion

- A measure of similarity between windows in the two images
- The corresponding element is given by window that maximizes the similarity criterion within a search region

## Search regions

- Theoretically, search region can be reduced to a 1-D segment, along the horizontal line (in future we will use term epipolar line), and within the disparity range.
- In practice, search a slightly larger region due to errors in calibration

# Correlation Approach

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Equations = w is the window size

$$c(dx, dy) = \sum_{k=-W}^{W} \sum_{l=-W}^{W} \psi(I_l(x_l + k, y_l + l), I_r(x_l + dx + k, y_l + dy + l))$$

disparity

$$\bar{\mathbf{d}} = (\bar{dx}, \bar{dy}) = \arg \max_{\mathbf{d} \in R} \{c(dx, dy)\}$$

Similarity criterion

- Cross-Correlation

$$\Psi(u, v) = uv$$

- Sum of Square Difference (SSD)

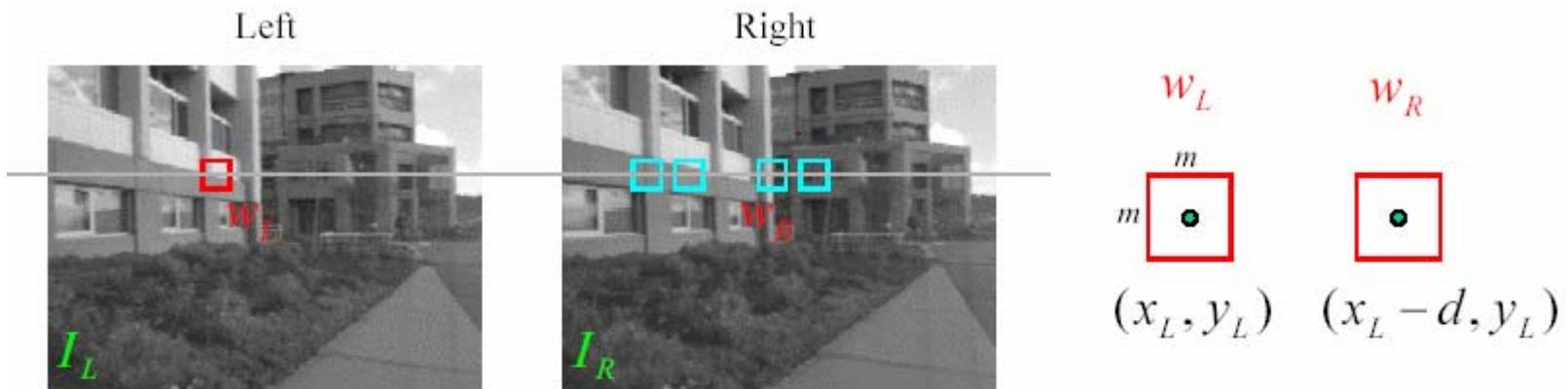
$$\Psi(u, v) = -(u - v)^2$$

- Sum of Absolute Difference(SAD)

$$\Psi(u, v) = -|u - v|$$

# Sum of Squared Differences (SSD)

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$w_L$  and  $w_R$  are corresponding  $m$  by  $m$  windows of pixels.

We define the window function :

$$W_m(x, y) = \{u, v \mid x - \frac{m}{2} \leq u \leq x + \frac{m}{2}, y - \frac{m}{2} \leq v \leq y + \frac{m}{2}\}$$

The SSD cost measures the intensity difference as a function of disparity :

$$C_r(x, y, d) = \sum_{(u, v) \in W_m(x, y)} [I_L(u, v) - I_R(u - d, v)]^2$$

# Correlation Approach

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## PROS

- Easy to implement
- Produces dense disparity map
- Usually is slow

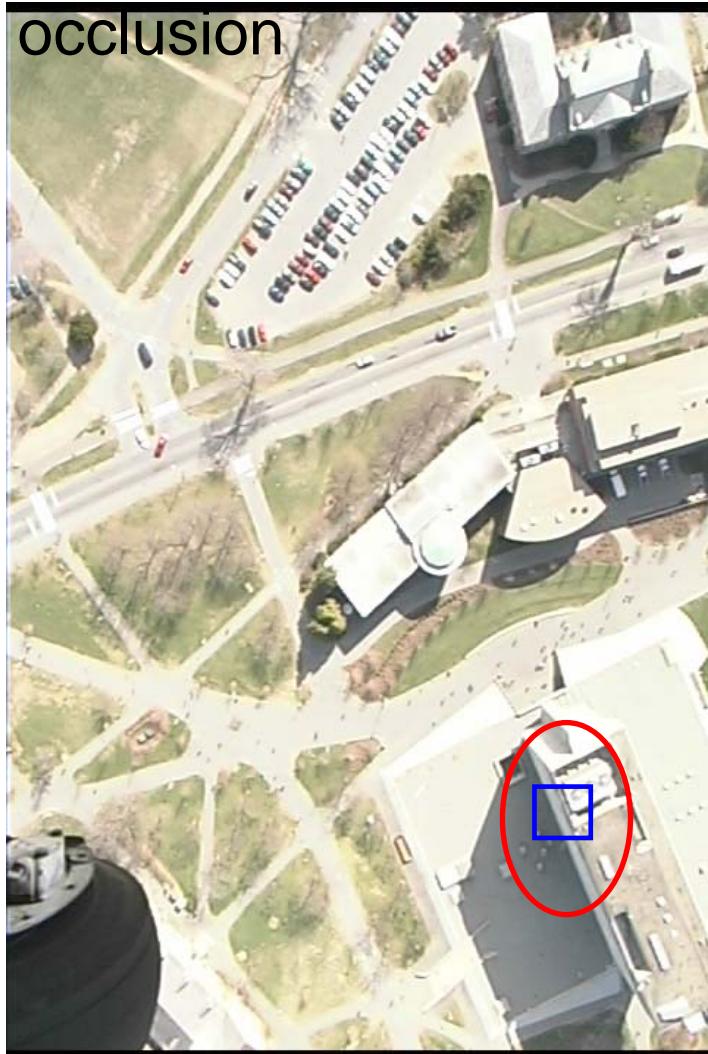
## CONS

- Needs textured images to work well
- Inadequate for matching image pairs from very different viewpoints due to illumination changes
- Window may cover points with quite different disparities
- Inaccurate disparities on the occluding boundaries

# Correlation Approach

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A Stereo Pair of UMass Campus – texture, boundaries and occlusion



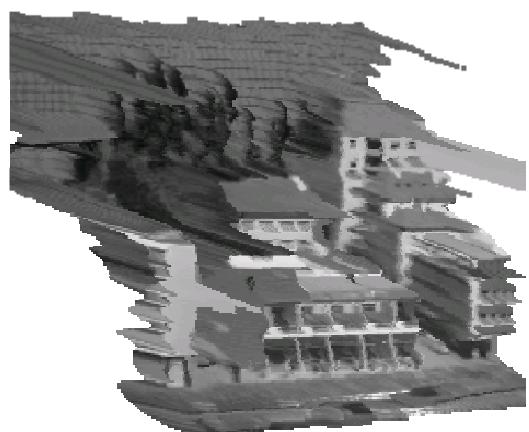
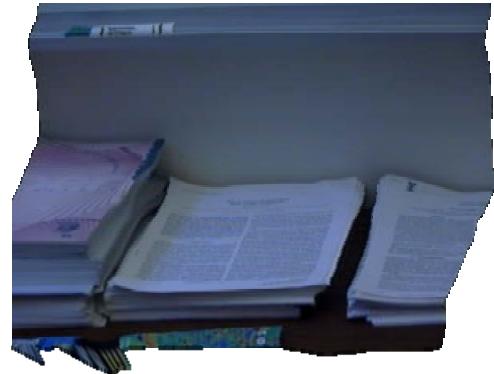
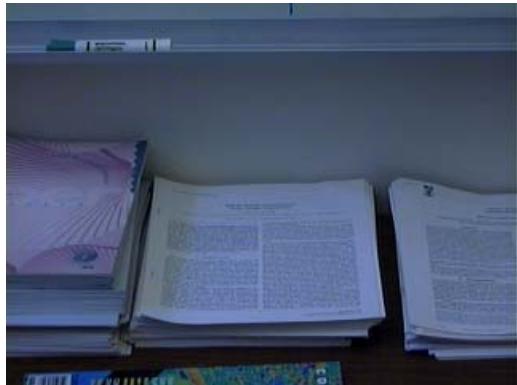
# Disparity Map

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- $D = \|x_1 - x_2\|$  measures the distance between corresponding points in two images
  - Normally disparity is stated as number of pixels
  - Clearly a particular simple stereo configuration has a maximum and minimum possible disparity
- Depth is inversely proportional to disparity
- If we compute the disparity for the entire images then we have a disparity map
- Display it as an image
  - Bright points have highest disparity (closest)
  - Dark points have lowest disparity (farthest)
- Disparity map is a 3D image

# Disparity Map

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# Correspondence Using Correlation

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Left



Disparity Map

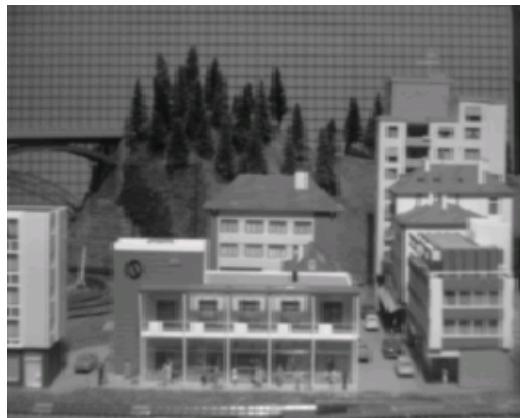


Images courtesy of Point Grey Research

# Dense Stereo Matching

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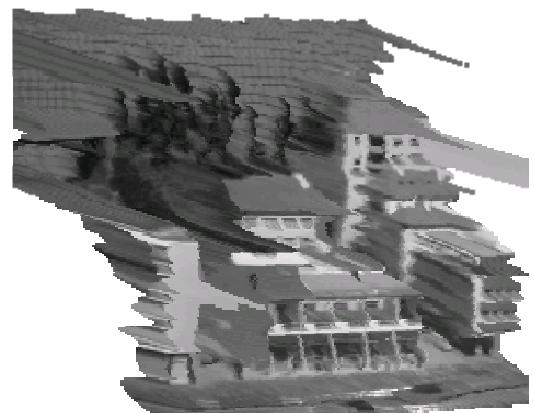
View extrapolation results



Input



depth image



novel view

[Matthies, Szeliski, Kanade '88]

# Feature-based Approach

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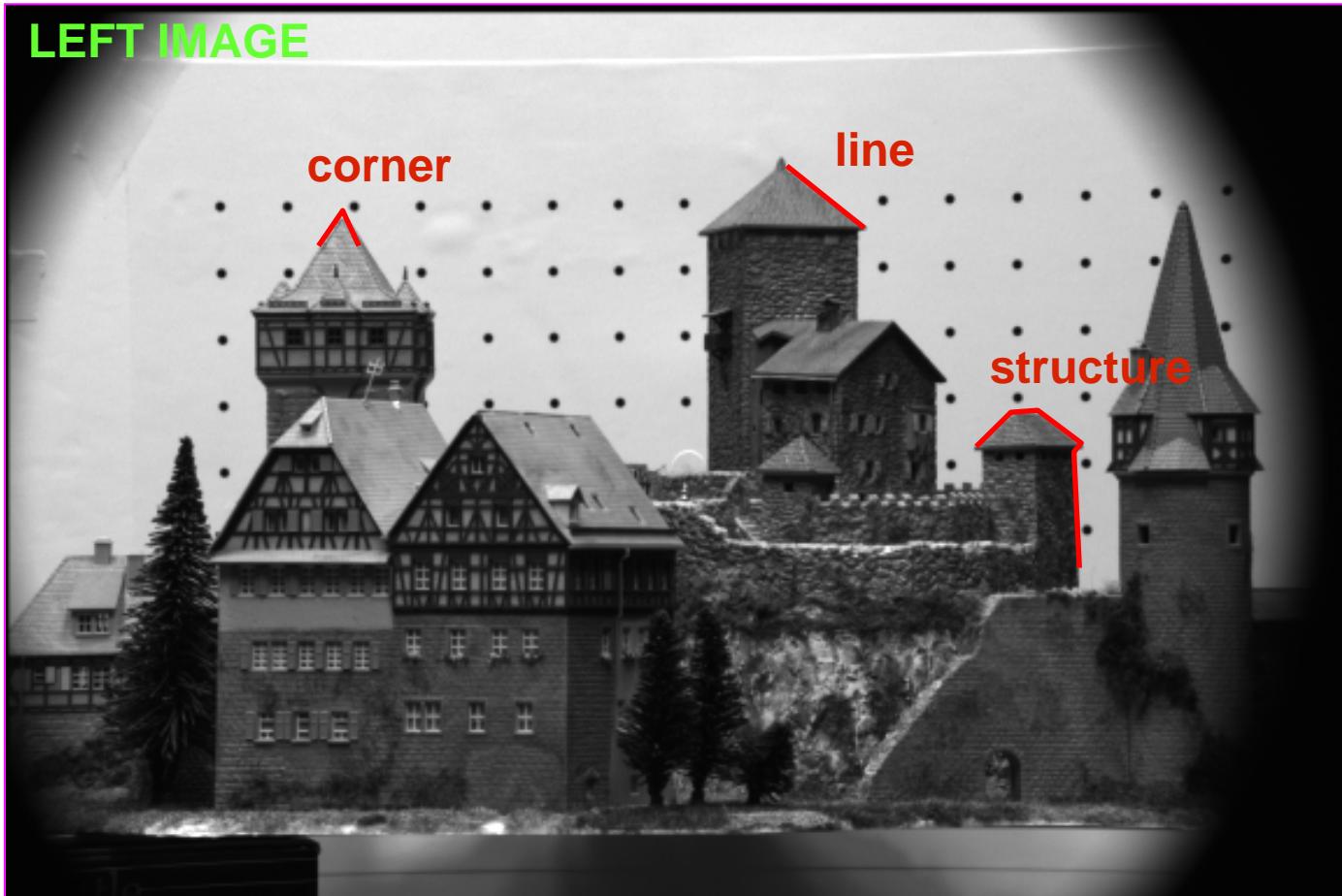
## Features

- Edge points
- Lines (length, orientation, average contrast)
- Corners

## Matching algorithm

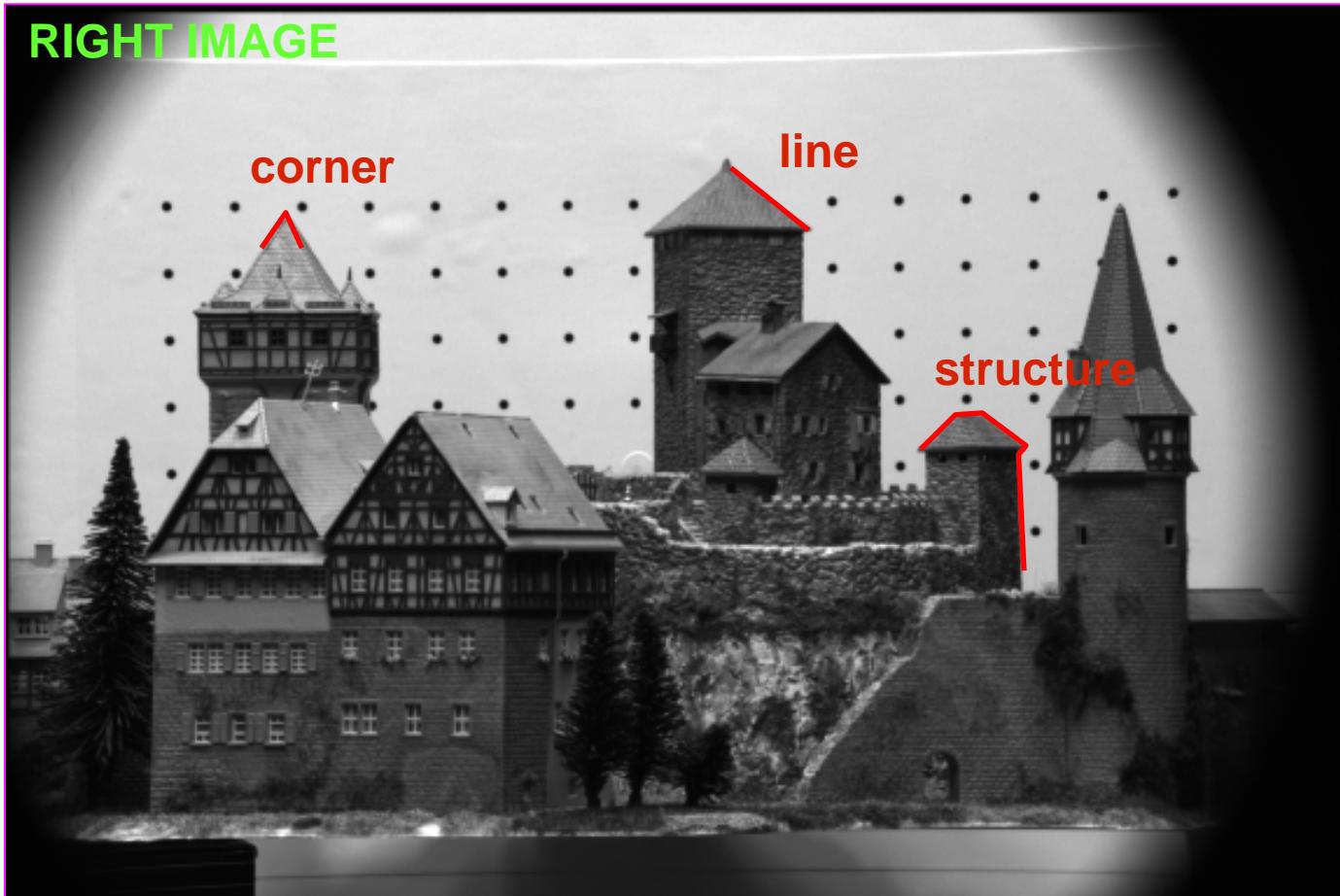
- Extract features in the stereo pair
- Define a suitable similarity measure for these features
- Use constraints to reduce number of matches
  - Geometric constraints
    - Need only match features on same horizontal line
  - Analytic constraints
    - Uniqueness – each feature has at most one match
      - » Often embedded into the left/right constraint
    - Continuity – disparity varies continuously almost everywhere across this image

# Feature-based Approach



For each feature in the left image...

# Feature-based Approach



Search in the right image... the disparity ( $dx$ ,  $dy$ ) is the displacement when the similarity measure is maximum

# Feature-based Approach

## PROS

- Relatively insensitive to illumination changes
- Good for man-made scenes with strong lines but weak texture or textureless surfaces
- Work well on the occluding boundaries (edges)
- Could be faster than the correlation approach

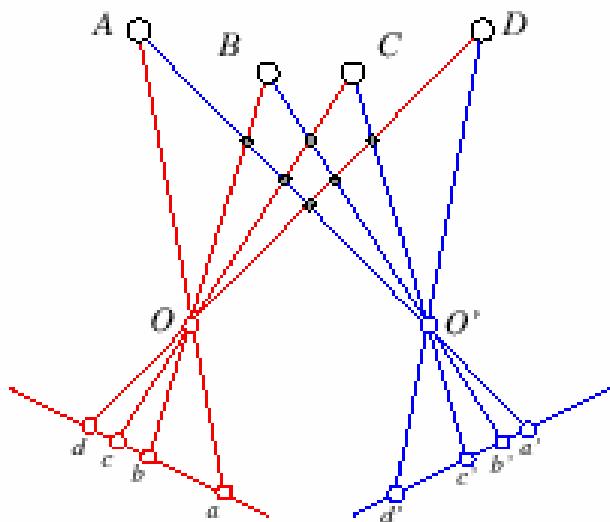
## CONS

- Only sparse depth map
- Feature extraction may be tricky for some features
  - Often uses corners as the features to match
  - Lines (Edges) might be partially extracted in one image
  - How to measure the similarity between two lines?

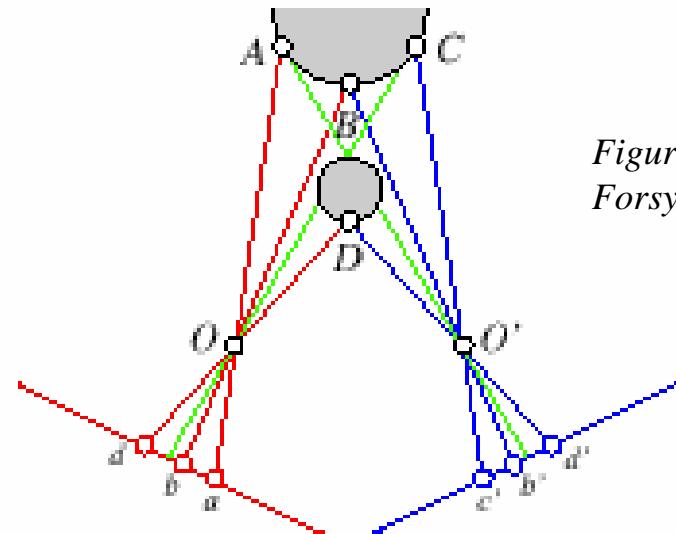
# Correspondence

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It is fundamentally ambiguous, even with stereo constraints



Ordering constraint...



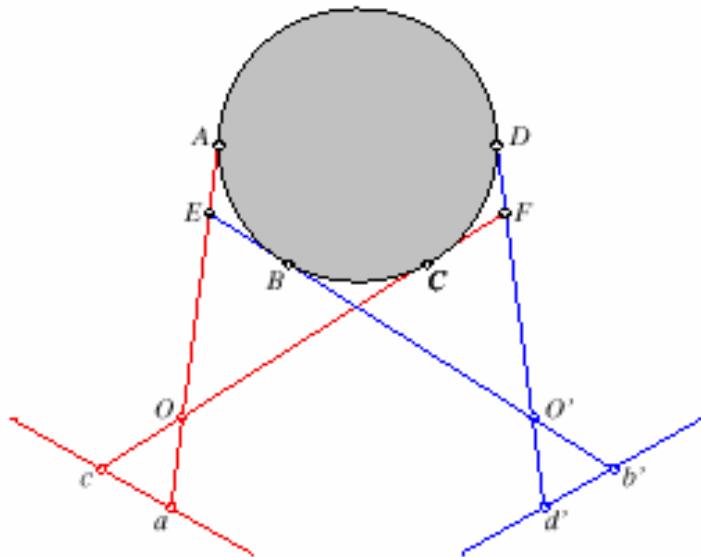
...and its failure

*Figure from  
Forsyth & Ponce*

# A Last Word on Correspondences

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Correspondence fail for smooth surfaces



There is currently no good solution to the correspondence problem

# Problems for Correspondence

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- Occlusions
  - Points with no counterpart in the other image
  - If algorithm produces a match this is an error
  - The wider the stereo baseline the more chance that there are occlusions
- Spurious matches
  - False correspondences produced for whatever reason
  - One reason is because of occlusions as described above
  - Another is that many elements are very similar