## Stereo

## Triangulation

- Suppose we have two cameras
- Calibrated: parameters known
- And a pair of corresponding pixels
- Find 3D location of point!

$P^{(2)}$



## Binocular stereo

- Given two calibrated cameras
- Find pairs of corresponding pixels
- Use corresponding image locations to set up equations on world coordinates
- Solve!


## Binocular stereo

- General case: cameras can be arbitrary locations and orientations



## Binocular stereo

- Special case: cameras are parallel to each other and translated along $X$ axis



## Stereo with rectified cameras

- Special case: cameras are parallel to each other and translated along $X$ axis



## Stereo head



Kinect / depth cameras


## Stereo with "rectified cameras"



## Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of $1^{\text {st }}$ camera

$$
\begin{aligned}
& \overrightarrow{\mathbf{x}}_{i m g}^{(1)} \equiv\left[\begin{array}{ll}
I & \mathbf{0}
\end{array}\right] \overrightarrow{\mathbf{x}}_{w} \\
& \overrightarrow{\mathbf{x}}_{i m g}^{(2)} \equiv\left[\begin{array}{ll}
I & \mathbf{t}
\end{array}\right] \overrightarrow{\mathbf{x}}_{w} \\
& \mathbf{t}=\left[\begin{array}{c}
t_{x} \\
0 \\
0
\end{array}\right]
\end{aligned}
$$

## Perspective projection in rectified cameras

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I & \mathbf{t}
\end{array}\right] \overrightarrow{\mathbf{x}}_{w} \\
& \mathbf{t}=\left[\begin{array}{c}
t_{x} \\
0 \\
0
\end{array}\right] \overrightarrow{\mathbf{x}}_{w}=\left[\begin{array}{c}
X \\
Y \\
Z \\
1
\end{array}\right]=\left[\begin{array}{c}
\mathbf{x}_{w} \\
1
\end{array}\right]
\end{aligned}
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\end{array}\right]\left[\begin{array}{c}
\mathbf{x}_{w} \\
1
\end{array}\right]=\mathbf{x}_{w}=\left[\begin{array}{l}
X \\
Y \\
Z
\end{array}\right] \\
& \overrightarrow{\mathbf{x}}_{i m g}^{(2)} \equiv\left[\begin{array}{ll}
I & \mathbf{t}
\end{array}\right]\left[\begin{array}{c}
\mathbf{x}_{w} \\
1
\end{array}\right]=\mathbf{x}_{w}+\mathbf{t}=\left[\begin{array}{c}
X+t_{x} \\
Y \\
Z
\end{array}\right]
\end{aligned}
$$

## Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of $1^{\text {st }}$ camera

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\begin{aligned}
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X \\
Y \\
Z
\end{array}\right] \\
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X+t_{x} \\
Y \\
Z
\end{array}\right]
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x_{1} \\
y_{1} \\
1
\end{array}\right] \equiv\left[\begin{array}{l}
X \\
Y \\
Z
\end{array}\right]} \\
& {\left[\begin{array}{c}
x_{2} \\
y_{2} \\
1
\end{array}\right] \equiv\left[\begin{array}{c}
X+t_{x} \\
Y \\
Z
\end{array}\right]}
\end{aligned}
$$

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\begin{aligned}
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x_{1} \\
y_{1} \\
1
\end{array}\right] \equiv\left[\begin{array}{l}
X \\
Y \\
Z
\end{array}\right]} \\
& {\left[\begin{array}{c}
x_{2} \\
y_{2} \\
1
\end{array}\right] \equiv\left[\begin{array}{c}
X+t_{x} \\
Y \\
Z
\end{array}\right]}
\end{aligned}
$$

## Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of $1^{\text {st }}$ camera

$$
\begin{aligned}
& {\left[\begin{array}{c}
\lambda x_{1} \\
\lambda y_{1} \\
\lambda
\end{array}\right]=\left[\begin{array}{l}
X \\
Y \\
Z
\end{array}\right]} \\
& {\left[\begin{array}{c}
\lambda x_{2} \\
\lambda y_{2} \\
\lambda
\end{array}\right]=\left[\begin{array}{c}
X+t_{x} \\
Y \\
Z
\end{array}\right]}
\end{aligned}
$$

## Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of $1^{\text {st }}$ camera
X coordinate differs by $\mathrm{t}_{\mathrm{x}} / \mathrm{Z}$
$x_{1}=\frac{X}{Z} \quad x_{2}=\frac{X+t_{x}}{Z}$
$y_{1}=\frac{Y}{Z} \quad y_{2}=\frac{Y}{Z}$
$Y$ coordinate is the same!


## Perspective projection in rectified cameras

- X coordinate differs by $\mathrm{t}_{\mathrm{x}} / Z$
- That is, difference in X coordinate is inversely proportional to depth
- Difference in X coordinate is called disparity
- Translation between cameras ( tx ) is called baseline
- disparity = baseline / depth


## The disparity image

- For pixel ( $\mathrm{x}, \mathrm{y}$ ) in one image, only need to know disparity to get correspondence
- Create an image with color at ( $\mathrm{x}, \mathrm{y}$ ) = disparity

right image

disparity
left image


## Perspective projection in rectified cameras



- For rectified cameras, correspondence problem is easier
- Only requires searching along a particular row.

NCC - Normalized Cross
Correlation

- Lighting and color change pixel intensities
- Example: increase brightness / contrast
- $I^{\prime}=\alpha I+\beta$
- Subtract patch mean: invariance to $\beta$
- Divide by norm of vector: invariance to $\alpha$
- $x^{\prime}=x-\langle x\rangle$
- $x^{\prime \prime}=\frac{x \prime}{\|x \prime\|}$
- similarity $=x^{\prime \prime} \cdot y^{\prime \prime}$


## Cross-correlation of neighborhood


regions $A, B$, write as vectors $\mathbf{a}, \mathbf{b}$
translate so that mean is zero
$\mathrm{a} \rightarrow \mathrm{a}-\langle\mathrm{a}\rangle, \mathrm{b} \rightarrow \mathrm{b}-\langle\mathbf{b}\rangle$
cross correlation $=\frac{\mathbf{a} \cdot \mathbf{b}}{|\mathbf{a}||\mathbf{b}|}$

left image band
right image band

## cross correlation



## The NCC cost volume

- Consider M x N image
- Suppose there are D possible disparities.
- For every pixel, D possible scores
- Can be written as an M x N x D array
- To get disparity, take max along $3^{\text {rd }}$ axis


## Computing the NCC volume

1. For every pixel ( $\mathrm{x}, \mathrm{y}$ )
2. For every disparity d
3. Get normalized patch from image 1 at ( $x, y$ )
4. Get normalized patch from image 2 at $(x+d, y)$
5. Compute NCC

## Computing the NCC volume

## 1. For every disparity d

## 1. For every pixel ( $x, y$ )

1. Get normalized patch from image 1 at $(x, y)$
2. Get normalized patch from image 2 at $(x+d, y)$
3. Compute NCC

Assume all pixels lie at same disparity d (i.e., lie on same plane) and compute cost for each

## Plane sweep stereo



## A similar special case

- Fixating camera system
- Eyes fixate on object
- Angle at which they merge proportional to inverse depth


## Stereograms

- Invented by Sir Charles Wheatstone, 1838



The Stereograph as an Educator-Underwood Patent Extension Cabinet in a home Library. (2)



Mark Twain at Pool Table", no date, UCR Museum of Photography


