## 3D <br> acquisition

# Computer Vision 

## 3D acquisition taxonomy



# Computer Vision 

## 3D acquisition taxonomy



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Stereo

The underlying principle is "triangulation" :


## (Passive) stereo

## Simple configuration :



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A simple stereo setup

$\square$ identical cameras
coplanar image planes
$\square$ aligned x-axes

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## A simple stereo setup

Reminder :

the camera projection can be formulated as
$\rho p=K R^{t}(P-C)$
for some non-zero $\rho \in \mathbb{R}$
Here $R$ is the identity...

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A simple stereo setup


$$
\rho\left(\begin{array}{l}
x \\
y \\
1
\end{array}\right)=K\left(\begin{array}{l}
X \\
Y \\
Z
\end{array}\right)\left|\rho^{\prime}\left(\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right)=K\left(\begin{array}{l}
X-b \\
Y \\
Z
\end{array}\right)\right| K=\left(\begin{array}{ccc}
f k_{x} 0 & 0 \\
0 & f k_{y} & 0 \\
0 & 0 & 1
\end{array}\right)
$$

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A simple stereo setup


$$
\left\{\begin{array} { l } 
{ x = \frac { f k _ { x } X } { Z } , } \\
{ y = \frac { f k _ { y } Y } { Z } , }
\end{array} \text { and } \left\{\begin{array}{l}
x^{\prime}=\frac{f k_{x}(X-b)}{Z}, \\
y^{\prime}=\frac{f k_{y} Y}{Z},
\end{array}\right.\right.
$$

Note that $y=y^{\prime}$

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## A simple stereo setup

The 3D coordinates of the point are

$$
\begin{aligned}
X & =b \frac{x}{\left(x-x^{\prime}\right)} \\
Y & =b \frac{k_{x}}{k_{y}} \frac{y}{\left(x-x^{\prime}\right)} \\
Z & =b k_{x} \frac{f}{\left(x-x^{\prime}\right)}
\end{aligned}
$$

$\left(x-x^{\prime}\right)$ is the so-called disparity
Stereo is imprecise for far away objects, but increasing $b$ and/or $f$ can increase depth resolution

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## A simple stereo setup

Notice: for this simple setup, same disparity means same depth


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same disparity means same depth


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## A simple stereo setup

Increasing $b$ increases depth resolution


one has to strike a balance with visibility...

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## A simple stereo setup

Increasing $f$ increases depth resolution

one has to strike a balance with visibility...

## Computer Vision <br> Remarks

$\square$ 1. increasing $b$ and/or $f$ increases depth resolution but reduces simultaneous visibility
$\square$ 2. iso-disparity loci are depth planes, not so for other configurations
$\square$ 3. human stereo vision only works up to $\pm 10 \mathrm{~m}$
$\square 4$. the real problem is finding correspondences

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## A simple stereo setup



The HARD problem is finding the correspondences

Notice : no reconstruction for the untextured back wall...

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The HARD problem is finding the correspondences

Notice : no reconstruction for the untextured back wall...

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## Stereo, the general setup

we start by the relation between the two projections of a point
in the second image the point must be along the projection of the viewing ray for the first camera :


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## Stereo, the general setup

We cast this constraint in mathematical expressions :
$p$ and $p$ ' are the two images of $P$

$$
\begin{aligned}
& \mu p=K R^{t}(P-C) \\
& \rho^{\prime} p^{\prime}=K^{\prime} R^{\prime t}\left(P-C^{\prime}\right)
\end{aligned}
$$

w.r.t. world frame $P$ is on the ray with equation

$$
P=C+\mu R K^{-1} p \quad \text { for some } \mu \in \mathbb{R}
$$

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## Stereo, the general setup

so, the ray is given by

$$
P=C+\mu R K^{-1} p \text { for some } \mu \in \mathbb{R}
$$

now we project it onto the second image in general, points project as follows :

$$
\rho_{p}^{\prime} p^{\prime}=K^{\prime} R^{\prime \prime}\left(P-C^{\prime}\right)
$$

and thus, filling in the ray's equation
$\rho^{\prime} p^{\prime}=\mu K^{\prime} R^{\prime t} R K^{-1} p+K^{\prime} R^{\prime t}\left(C-C^{\prime}\right)$

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Stereo, the general setup
the projected ray was found to be
$\rho^{\prime} p^{\prime}=\mu K^{\prime} R^{t} R K^{-1} p+K^{\prime} R^{\prime t}\left(C-C^{\prime}\right)$
the second term is the projection of the 1st camera's center, the so-called epipole

$$
\rho_{e}^{\prime} e^{\prime}=K^{\prime} R^{\prime t}\left(C-C^{\prime}\right)
$$

the first term is the projection of the ray's point at infinity, the so-called vanishing point
finally, adopting the simplifying notation

$$
\begin{aligned}
& A=\frac{1}{\rho_{e}^{\prime}} K^{\prime} R^{\prime t} R K^{-1} \\
& \rho^{\prime} p^{\prime}=\rho_{e}^{\prime}\left(\mu A p+e^{\prime}\right)
\end{aligned}
$$

$A$ is the infinity homography

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## Stereo, the general setup

 note that the epipole lies on all the epipolar lines

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## Stereo, the general setup

$$
\rho^{\prime} p^{\prime}=\rho_{e}^{\prime}\left(\mu A p+e^{\prime}\right)
$$

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## Stereo, the general setup

$$
\rho^{\prime} p^{\prime}=\rho_{e}^{\prime}\left(\mu A p+e^{\prime}\right)
$$

expresses that $p$ 'lies on the line $l$ 'through the epipole $e^{\prime}$ and the vanishing point $A p$ of the ray of sight of $p$ (in the $2^{n d}$ image)

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## Stereo, the general setup

$$
\rho^{\prime} p^{\prime}=\rho_{e}^{\prime}\left(\mu A p+e^{\prime}\right)
$$

the epipolar constraint (epipolar line)
we can rewrite this constraint as

$$
\left|p^{\prime} e^{\prime} A p\right|=p^{\prime \prime}\left(e^{\prime} \times A p\right)=0
$$

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Stereo, the general setup

$$
\left|p^{\prime} e^{\prime} A p\right|=p^{\prime \prime}\left(e^{\prime} \times A p\right)=0
$$

can be written, given

$$
\begin{aligned}
& {\left[e^{\prime}\right]_{\times}=\left(\begin{array}{rrr}
0 & -e_{3}^{\prime} & e_{2}^{\prime} \\
e_{3}^{\prime} & 0 & -e_{1}^{\prime} \\
-e_{2}^{\prime} & e_{1}^{\prime} & 0
\end{array}\right)} \\
& \text { as } \\
& \left|p^{\prime} e^{\prime} A p\right|=p^{\prime t}\left[e^{\prime}\right]_{\times} A p \\
& F=\left[e^{\prime}\right]_{\times} A \text { is the fundamental matrix }
\end{aligned}
$$

$F$ is a $3 \times 3$ matrix, but has rank 2

## Stereo, the general setup

$$
p^{\prime t}\left[e^{\prime}\right]_{\times} A p=0 \rightarrow p^{\prime t} F p=0
$$

The 3-vector $p^{\prime} t F$ contains the line coordinates of the epipolar line of $p^{\prime}$ (i.e. a line in the 1 st image)

The 3-vector $F p$ contains the line coordinates of the epipolar line of $p$ (i.e. a line in the 2 nd image)

Hence, the epipolar matrix works in both directions

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Stereo, the general setup


Andrea Fusiello, CVonline

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Epipolar geometry cont'd


## Computer <br> Epipolar geometry cont'd

 Vision- Epipolar lines are in mutual correspondence

- allows to separate matching problem: matching pts on an epipolar line to pts on the corresponding epipolar line

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## Exploiting epipolar geometry

Separate 2D correspondence search problem to 1D search problem by using two view geometry


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Epipolar geometry cont'd



## Stereo, the general setup

$\square$ one point yields one equation $p^{\prime t} F p=0$ that is linear in the entries of the fundamental matrix $F$
so, we can actually obtain F without any prior knowledge about camera settings if we have sufficient pairs of corresponding points !!
b F can be computed linearly from 8 pairs of corresponding points, i.e. already from 8 'correspondences’ (not 9 , as this is a homogeneous system and one coefficient can be fixed to value 1 to fix the scale !)

- F being rank 2 yields an additional, but non-linear constraint. Thus, 7 correspondences suffice to non-linearly solve for $F$


## Stereo, the general setup

## Remarks :

- Of course, in practice one wants to use as many Correspondences as available, e.g. for obtaining a least-squares solution, based on the linear system, followed by a step to impose rank 2.
- Often, $F$ is found through a procedure called RANSAC (RANdom Sample Consensus). It starts from a randomly drawn subset of correspondences of minimal size (e.g. 8), and then keeps on drawing until a subset is found that yields an F so that many correspondences are seen to obey the epipolar constraint. RANSAC is good to fend off against correspondences that are wrong ('outliers')


## Relations between 3 views

one could use more than 2 images, e.g. 3 suppose $P$ projects to $p, p$, and $p$ "
$p$ " is found at the intersection of epipolar lines :
image 1

fails when the epipolar lines coincide

$$
\Rightarrow \quad \text { trifocal constraints }
$$

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## Correspondence problem : constraints

Reducing the search space :

- 1. Points on the epipolar line
- 2. Min. and max. depth $\Rightarrow$ line segment
- 3. Preservation of order
- 4. Smoothness of the disparity field


## Correspondence problem : methods

1. correlation
deformations...
small window $\Rightarrow$ noise!
$\square$ large window $\Rightarrow$ bad localisation
2. feature-based
$\square$ mainly edges and corners
$\square$ sparse depth image
3. regularisation methods

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## Stereo, the general setup

3D reconstruction

$$
\begin{aligned}
& P=C+\mu R K^{-1} p \\
& P=C^{\prime}+\mu^{\prime} R^{\prime} K^{\prime-1} p^{\prime}
\end{aligned}
$$

Yields 6 equations in 5 unknowns $\mathrm{X}, \mathrm{Y}, \mathrm{Z}$ and $\mu, \mu^{\prime}$

However, due to noise and errors, the rays may not intersect!
$\Rightarrow$ e.g. use the middle where the rays come closest

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## 3D city models - ground level

Mobile mapping example - for measuring


## Computer <br> 3D city models - ground level

 VisionCan also be turned into 3D for visualisation, but one needs to stay close to the camera viewpoints.

The example shown is of Quebec

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## 3D city models - ground level

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## Uncalibrated reconstruction

From 2 views...


If the camera translates...
An affine reconstruction can be made A projective reconstruction is always possible (if no pure rot.)

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## Uncalibrated reconstruction

From 3 general views taken with the same camera parameters...


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## Uncalibrated reconstruction



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## Uncalibrated reconstruction



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## Uncalibrated reconstruction - example



Univ. of Leuven

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## Shape-from-stills

## Input Images

 shots taken with Canon EOS D60(Resolution: 6,3 Megapixel )

## Computer Shape-from-stills

## www.arc3d.be

Webservice,
free for non-commercial use

# Computer Vision 

## 3D acquisition taxonomy



## Active triangulation

INTERSECTION LASER RAY AND VIEWING RAY


CAMERA'S CENTER OF PROJECTION

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## Active triangulation



CAMERA'S CENTER OF PROJECTION

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## Active triangulation

LASER


Two lines do normally not intersect... Noise disrupts triangulation

LASER SPOT SEEN BY THE
CAMERA

NO INTERSECTION YF SOME ERRORS IMTHE LINE EQS!


## Active triangulation

INTERSECTION LASER PLANE \& OBJECT SURFACE
LASER WITH CYLINDRICAL LENSE IN FRONT

POINT ON THE LASER LINE SEEN BY THE CAMERA

## Active triangulation <br> INTERSECTION

 LASER PLANE \& OBJECT SURFACELASER WITH CYLINDRICAL LENSE IN FRONT

A plane and a line do normally intersect...
Noise has little Influence on the triangulation

POINT ON THE LASER LINE SEEN BY THE CAMERA

CAMERA'S CENTER OF PROJECTION

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## Active triangulation



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## Active triangulation

Triangulation $\rightarrow$ 3D measurements


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## Active triangulation

Camera image


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## Active triangulation



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## Active triangulation

## Example 1 Cyberware laser scanners



Desktop model for small objects

Head scanner


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## Active triangulation

## Example 2 Minolta



Portable desktop model

## 3D acquisition taxonomy



## Structured light

patterns of a special shape are projected onto the scene
deformations of the patterns yield information on the shape

Focus is on combining a good resolution with a minimum number of pattern projections

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## Serial binary patterns

A sequence of patterns with increasingly fine subdivisions

Yields $2^{n}$ identifiable lines for only $n$ patterns


## Reducing the nmb of projections: colour

Binary patterns
Yields $2^{n}$ identifiable lines for only $n$ patterns
Using colours, e.g. 3,
Yields $3^{\text {n }}$ identifiable lines for only $n$ patterns


Interference from object colours...

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## One-shot implementation

## 3D from a single frame - KULeuven '96:



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One-shot implementation
KULeuven '81: checkerboard pattern with column code example :


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## 3D reconstruction for the example



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## An application in agriculture



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One-shot 3D acquisition

## Leuven ShapeCam



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## Shape + texture often needed

Higher resolution
Texture is also extracted


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## James Bond

Die another day

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## Active triangulation

## Recent, commercial example



## KINECT

for Brecoryea
Kinect 3D camera, affordable and compact solution by Microsoft.

Projects a 2D point pattern in the NIR, to make it invisible to the human eye

Computer
Kinect: $9 x 9$ patches with locally unique code Vision


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## Kinect as one-shot, low-cost scanner

Excerpt from the dense NIR dot pattern:

http://research.microsoft.com/apps/video/default.aspx?i 15

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## Face animation - input



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## Face animation - replay + effects



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## Facial motion capture

motion capture for League of Extraordinary Gentlemen


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## Facial motion capture

(2) COMPUTERCAFE


LC015 Eyetronics 1/291

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## Phase shift

## color wheel



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## Phase shift

$$
\begin{aligned}
& I_{r}=A+R \cos (\phi-\theta) \\
& I_{g}=A+R \cos (\phi) \\
& I_{b}=A+R \cos (\phi+\theta)
\end{aligned}
$$

1. detect phase from 3 subsequently projected cosine patterns, shifted over 120 degrees
2. unwrap the phases / additional stereo
3. texture is obtained by summing the 3 images / color camera w. slower integration

## Computer

## Phase shift

 Vision$A=\frac{I_{r}+I_{g}+I_{b}}{3}$
$\phi=\arctan \left(\tan \left(\frac{\theta}{2}\right) \frac{I_{r}-I_{b}}{2 I_{g}-I_{r}-I_{b}}\right)$


Vision

## 4D acquisition

Motion retargetting, from 3D phase shift scans

## Face/Off: Live Facial Puppetry <br> PaperID 102

# Computer Vision 

## 3D acquisition taxonomy



## Time-of-flight

measurement of the time a modulated light signal needs to travel before returning to the sensor
this time is proportional to the distance
waves:

1. radar
2. sonar
3. optical radar
low freq. electromagnetic acoustic waves
optical waves
working principles :
4. pulsed
5. phase shifts

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Time-of-flight
Example 1: Cyrax


Example 2: Riegl


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## Time-of-flight: example

Cyrax ""
3D Laser Mapping

## System

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Cyrax

Accurate, detailed, fast measuring


Integrated modeling

## Cyrax



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## Pulsed laser (time-of-flight)

## No reflectors

 needed
## 2mm-6mm accuracy

Distance $=\mathrm{C} \times \Delta \mathrm{T} \div 2$

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Laser sweeps over surface

## 800 pts/sec

## $40^{\circ} \times 40^{\circ}$

Field-of-view (max)

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## Up to <br> 100m range (50m rec) <br> 

Eye-safe Class 2

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Cyrax is also a visualization tool

Cyrax detects the intensity of each reflected laser pulse and colors it


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## Step 1:

## Target the structure



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Step 2:
Scan the structure


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Step 3:

## Color the points



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Step 4:
Model fitting in-the-field


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## Result




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## Project: As-built of

## Chevron hydrocarbon plant



- 400 ’x500' area
- 10 vessels; 5 pumps
- 6,000 objects
- 81 scans from 30 tripod locations
- Cyrax field time $=50 \mathrm{hrs}$

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# Cost Benefits 

## Measuring \& modeling



## Added Value Benefits

- Higher accuracy
- Fewer construction errors
- 6 week schedule savings


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Application Modeling movie sets


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## Lidar data with Riegl LMS-Z390i


courtesy of RWTH Aachen, L. Kobbelt et al.

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## Comparison Lidar - passive

## 3-D Reconstruction based on

## Multi-View Stereo

## LIDAR Measurements

# Computer Vision 

## 3D acquisition taxonomy



## Computer <br> Vision <br> Shape-from-texture

assumes a slanted and tilted surface to have a homogeneous texture
inhomogeneity is regarded as the result of projection
e.g. anisotropy in the statistics of edge orientations
orientations deprojecting to maximally isotropic texture

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# Computer Vision 

## 3D acquisition taxonomy



## Shape-from-contour

makes assumptions about contour shape
E.g. the maximization of area over perimeter squared (compactness)

$$
\text { ellipse } \stackrel{\Downarrow}{\rightarrow} \text { circle }
$$

E.g. assumption of symmetry

Symmetric contours $\stackrel{\downarrow}{\rightarrow}$ surface of revolution

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## Shape-from-contour



## 3D acquisition taxonomy



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## Shape-from-silhouettes



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## Shape from silhouettes - uncalibrated

## tracking of turntable rotation

- volumetric modeling from silhouettes
- triangular textured surface mesh


VRML model

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## Outdoor visual hulls



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## Outdoor visual hulls

 $\Delta$ e
$\Rightarrow$



## 3D acquisition taxonomy



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# REAL-TIME FOCUS RANGE SENSOR 

SHREE X. NAYAR

Masahigo Watamaze
Minori Mocuchi
COLUMAIA UNIVERSITY

## 3D acquisition taxonomy



## Shape-from-shading

Uses directional lighting, often with known direction
local intensity is brought into correspondence with orientation via reflectance maps
orientation of an isolated patch cannot be derived uniquely
extra assumptions on surface smoothness and known normals at the rim

# Computer Vision 

## 3D acquisition taxonomy



## Photometric stereo

constraint propagation eliminated by using light from different directions
simultaneously when the light sources are given different colours

## $\underset{\substack{\text { Computer } \\ \text { Vision }}}{\substack{\text { Mini-dome for photometric stereo } \\ \hline}}$

Instead of working with multi-directional light applied simultaneously with the colour trick, one can also project from many directions in sequence...

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Mini-dome for photometric stereo

## KATHOLIEKE UNIVERSITEIT LEUVEN

Computer
Mini-dome
Vision

## Computer

Mini-dome


## Computer Vision <br> Mini-dome for photometric stereo

## Example for tablet with first world map known,

 an exhibit at the British Museum:http://homes.esat.kuleuven.be/~mproesma/mptmp/cuneiform

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## Mini-dome for photometric stereo


(1) (3)


## Computer Vision <br> 3 D and recognition integrated

 3D City Modeling using Cognitive Loops

Computer
Multi-walker tracker
Vision

$\underset{\substack{\text { Computer } \\ \text { Vision }}}{ }$ Strongest 3D cues for us are 2D...


