

Visual Servoing

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4632B/8803
Mobile Manipulation
Lecture 8



From: <http://www.hsi.gatech.edu/visitors/maps/>



4th floor
4100Q
M Building
167

First office
on HSI side

Find the KUKA cup.



Manipulate the KUKA cup.



Overview

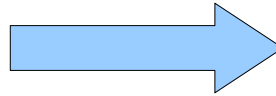
- Manipulation
- Global approach
- Local approach
- Visual Servoing

Manipulating a rigid body

- Grasping it
- Transporting it
- Placing it

Rigid body transformations

Initial pose

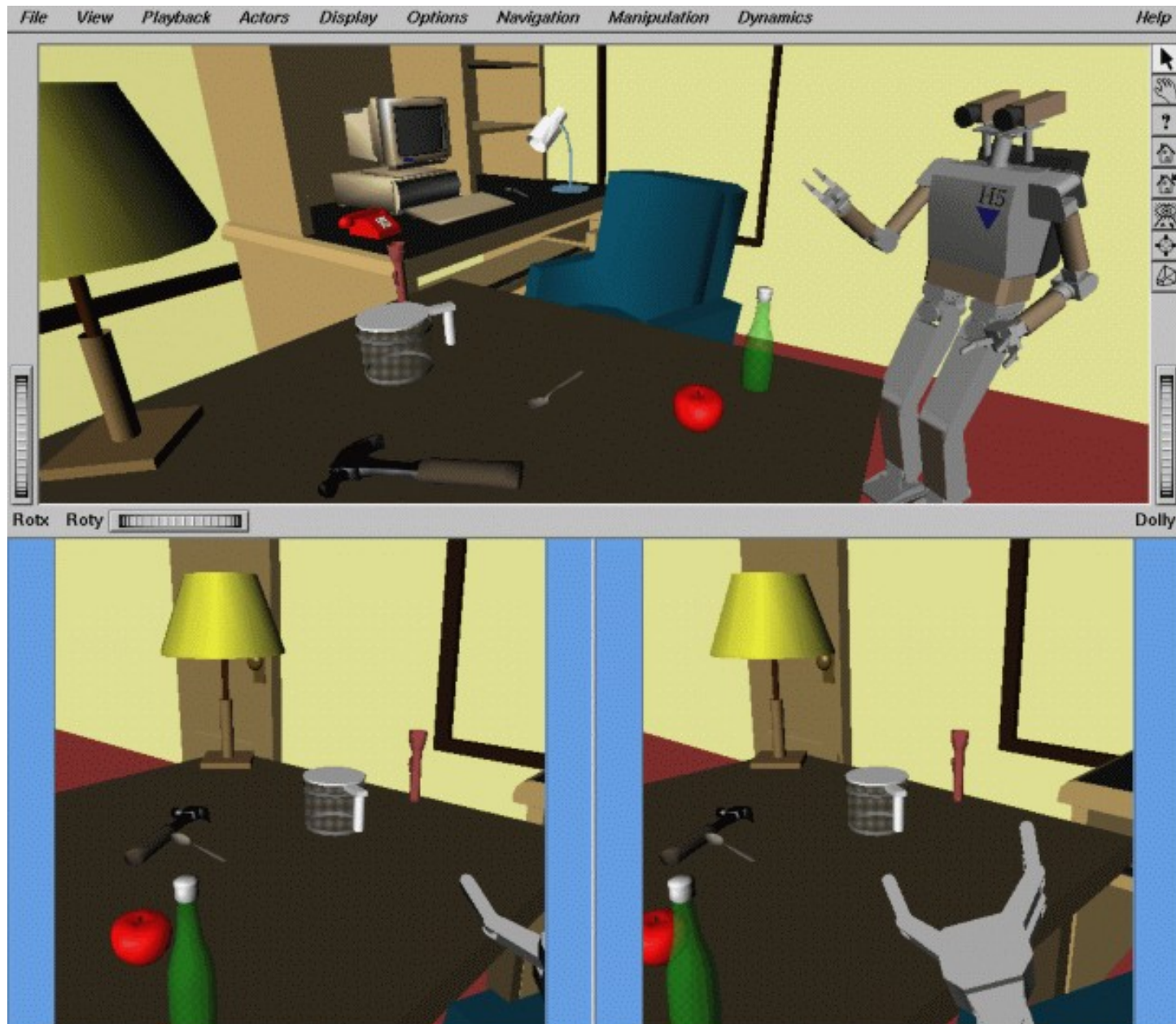


Final pose

We have a generative model!

Motion planning

(search for an entire trajectory)

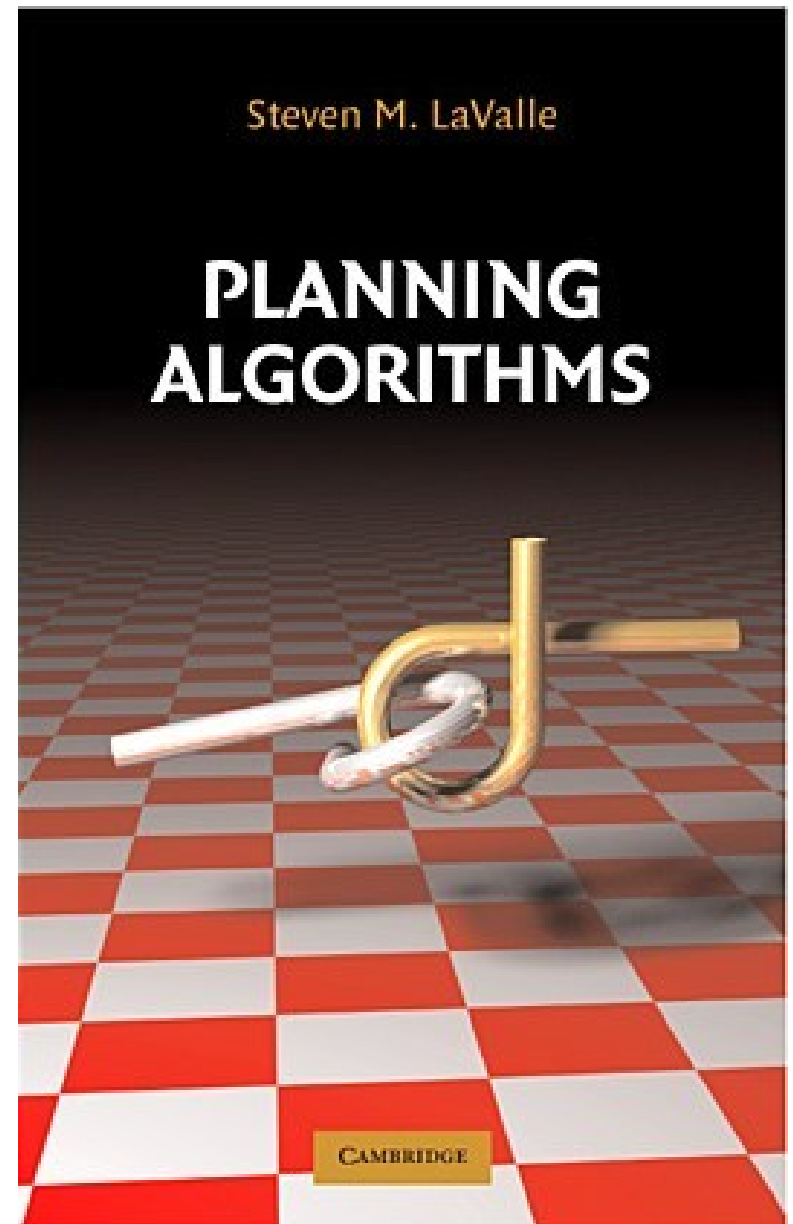


From: James Kuffner's home page (CMU)

<http://www.kuffner.org/james/humanoid/pics/mainWindowSnap.gif>

Motion planning

- Global solutions
- Drawbacks
 - Difficult to move from simulation to the real world
 - Sensing rarely included
 - Usually assume known state
 - Usually assume well-modeled transitions between states
 - May not meet real-time constraints



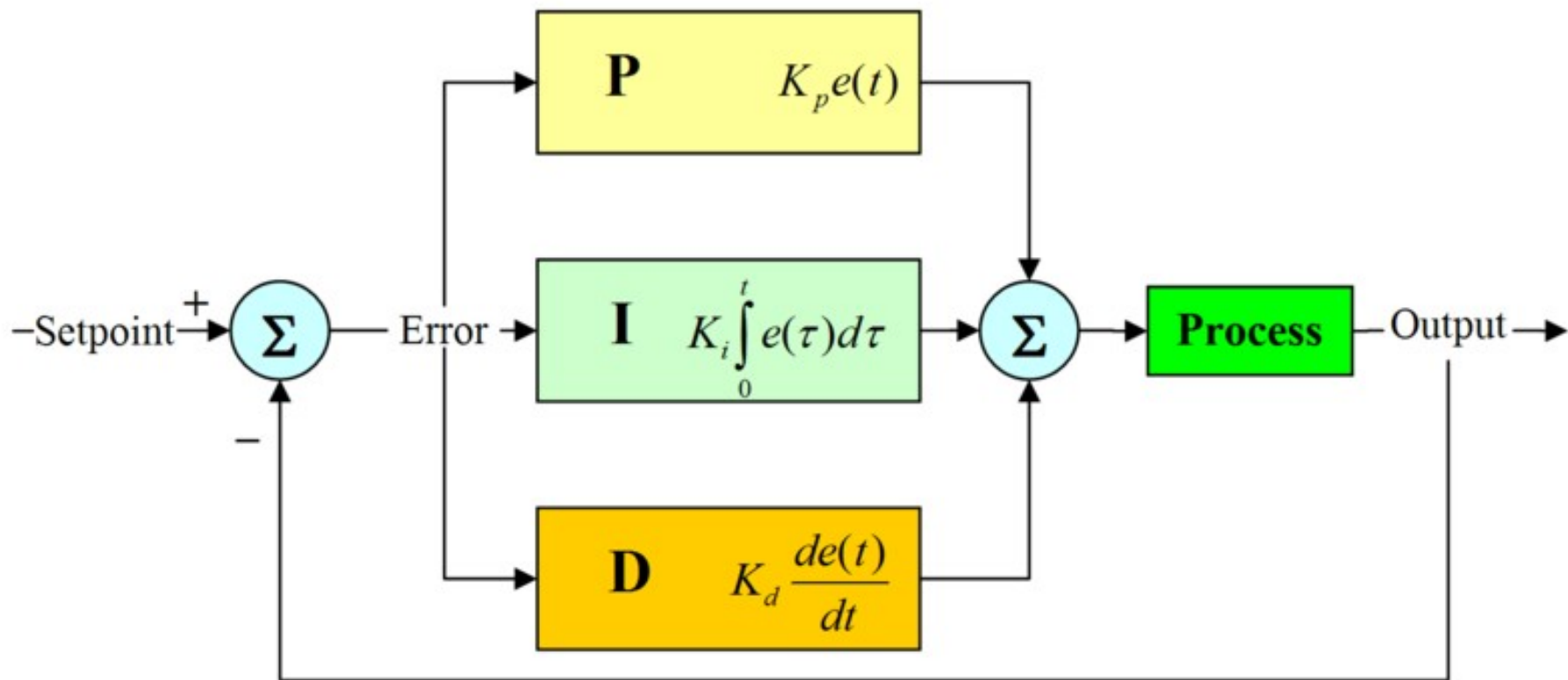
Can we find an efficient and robust
local solution?

Rigid body transformations

Initial pose  Final pose

- Define an error function
- Locally minimize this error function
- Simple feedback control (PID)
- Not sensor specific
 - e.g., Rod Grupen's group at UMass Amherst

proportional-integral-derivative controller (PID controller)



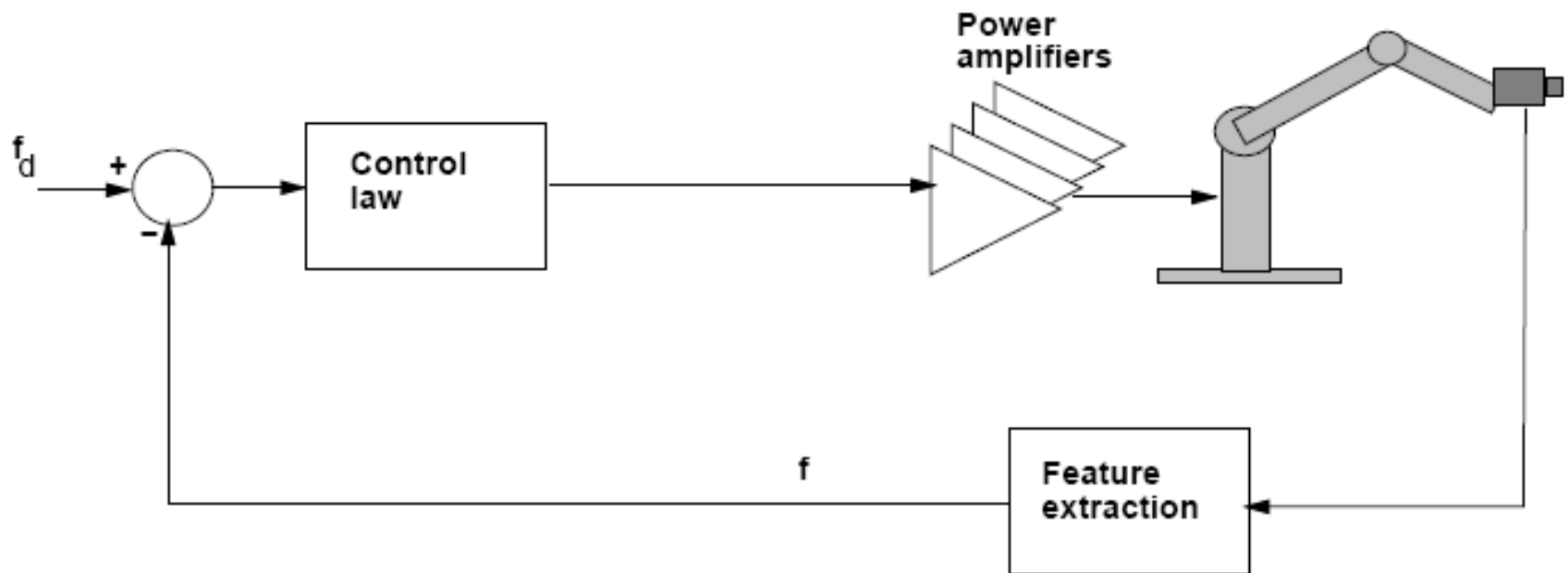
From: wikipedia article on PID control

<http://en.wikipedia.org/wiki/Image:Pid-feedback-nct-int-correct.png>

What poses? Relative to what?

- Grasping it
- Transporting it
- Releasing it

Visual Servoing



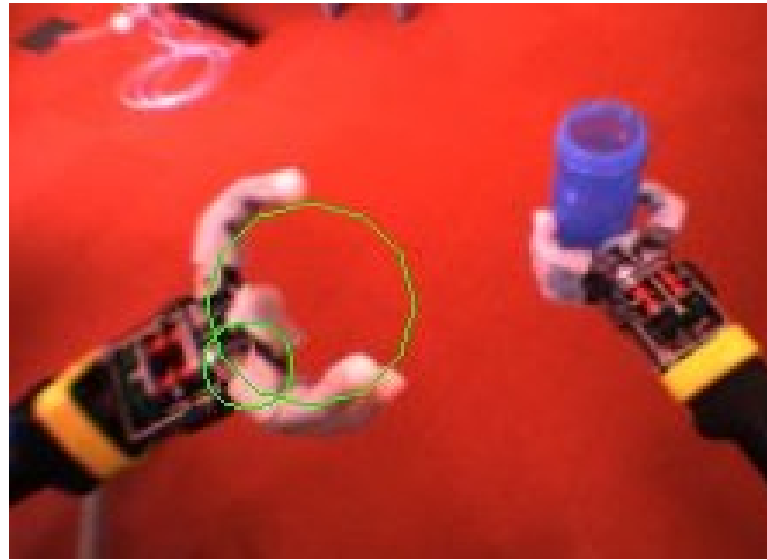
Visual servoing

S. Hutchinson, G. Hager, and P. Corke,
A tutorial on visual servo control,
IEEE Transactions on Robotics and Automation,
vol. 12, pp. 651–670, October 1996.

Examples



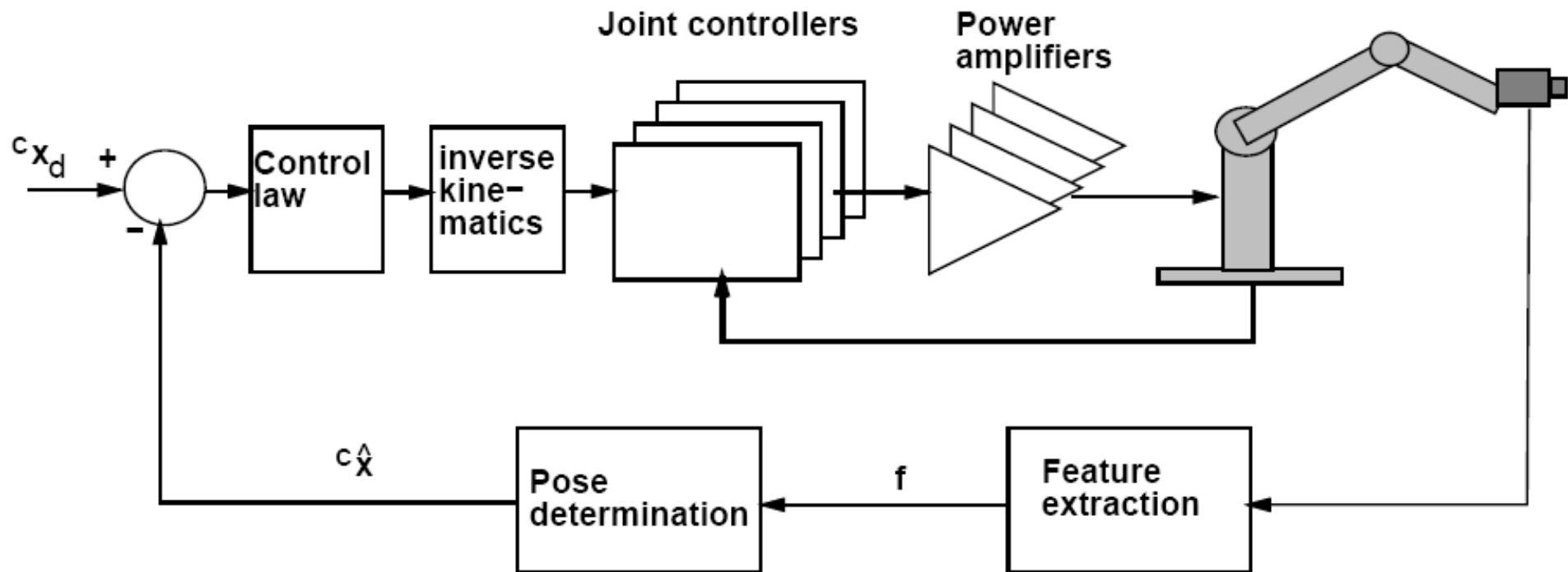
Examples



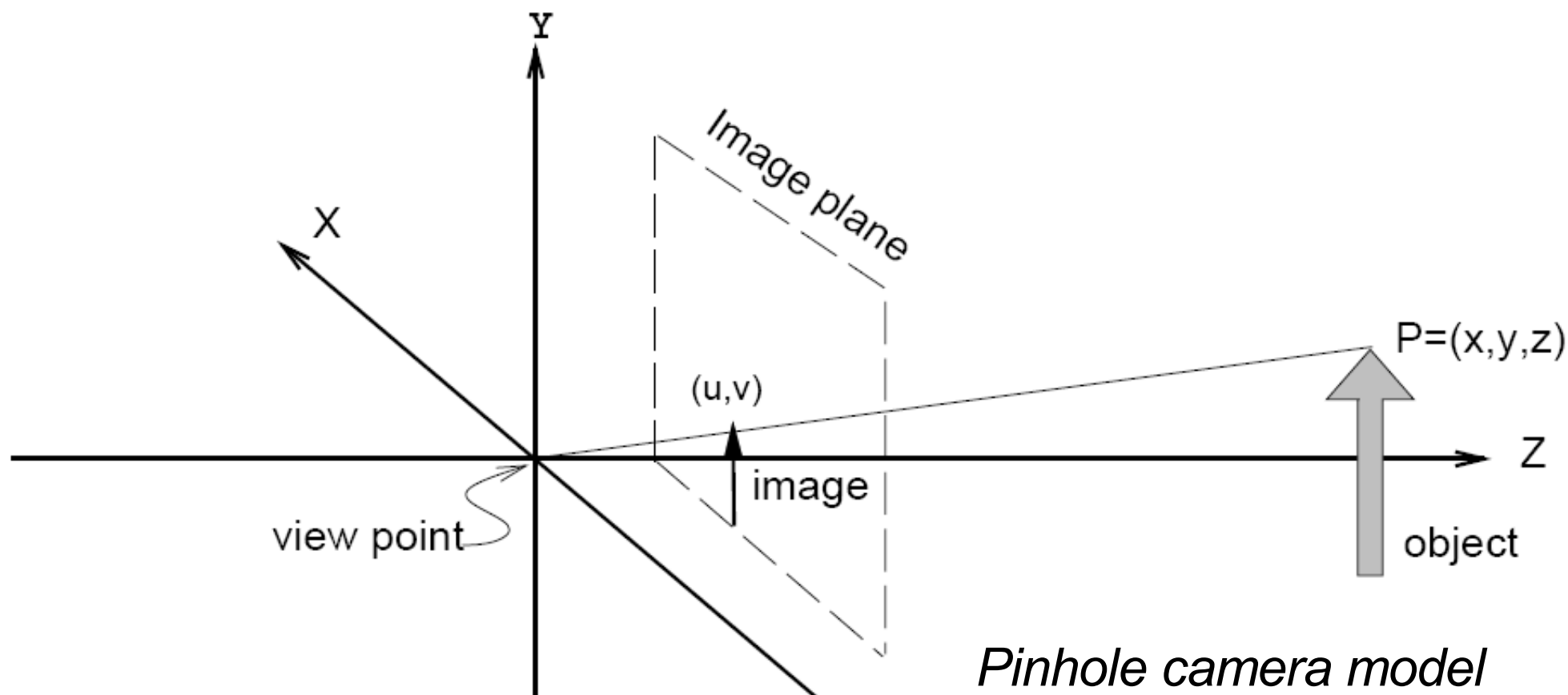
Major options

- Camera placement
 - End-effector mounted
 - Eye-in-hand
 - EOL (endpoint open-loop)
 - Fixed in the workspace
 - Might be PTZ
 - ECL (endpoint closed-loop)
- Error function
 - 3D pose
 - Image features
- Control hierarchy

Error function in 3D



Generative Model (Basic Image Formation)



$$\pi(x, y, z) = \begin{bmatrix} u \\ v \end{bmatrix} = \frac{\lambda}{z} \begin{bmatrix} x \\ y \end{bmatrix}$$

Perspective projection

Camera Calibration

- http://www.vision.caltech.edu/bouguetj/calib_doc/
- Estimates parameters
 - focal length, principal point, skew coefficient, distortions (radial and tangential)
- Rectify the images



Example

fixed camera, point to point

$$\mathbf{E}_{pp}(\mathbf{x}_e; \mathbf{S}, {}^e\mathbf{P}) = \mathbf{x}_e \circ {}^e\mathbf{P} - \mathbf{S}$$

$$\mathbf{u}_3 = -k E_{pp}(\hat{\mathbf{x}}_e; \hat{\mathbf{x}}_c \circ {}^c\hat{\mathbf{S}}, {}^e\mathbf{P}) = -k \left(\mathbf{x}_e \circ {}^e\mathbf{P} - \hat{\mathbf{x}}_c \circ {}^c\hat{\mathbf{S}} \right)$$

$\hat{\mathbf{x}}_e, \hat{\mathbf{x}}_c$ or ${}^c\hat{\mathbf{S}}$

robot kinematics, camera calibration and
visual reconstruction

Example eye-in-hand, EOL

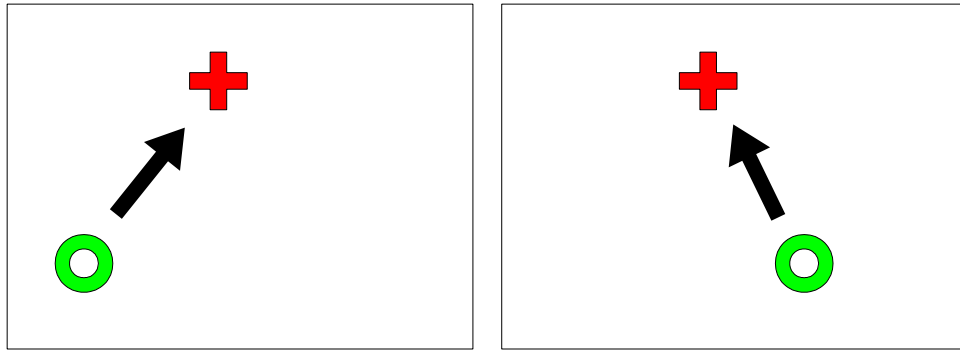
$${}^e\mathbf{E}_{pp}(\mathbf{x}_e; \mathbf{S}, {}^e\mathbf{P}) = {}^e\mathbf{P} - {}^e\mathbf{x}_0 \circ \mathbf{S}$$

$$\hat{\mathbf{S}} = \hat{\mathbf{x}}_e \circ {}^e\hat{\mathbf{x}}_c \circ {}^c\hat{\mathbf{S}}$$

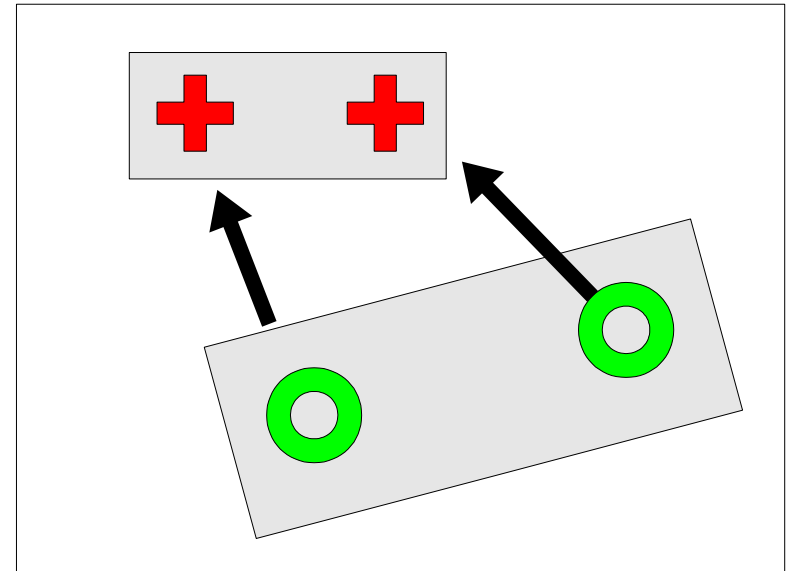
$$\begin{aligned} {}^e\mathbf{u}_3 &= -k {}^e\mathbf{E}_{pp}(\hat{\mathbf{x}}_e; \hat{\mathbf{x}}_e \circ {}^e\hat{\mathbf{x}}_c \circ {}^c\hat{\mathbf{S}}, {}^e\mathbf{P}) \\ &= -k({}^e\mathbf{P} - {}^e\mathbf{x}_0 \circ {}^0\hat{\mathbf{x}}_e \circ {}^e\hat{\mathbf{x}}_c \circ {}^c\hat{\mathbf{S}}) = -k({}^e\mathbf{P} - {}^e\hat{\mathbf{x}}_c \circ {}^c\hat{\mathbf{S}}) \end{aligned}$$

servo without reconstruction

(error as a function of image features)



Multi-camera



Single camera

Error measurements imply 3D

feature based servoing

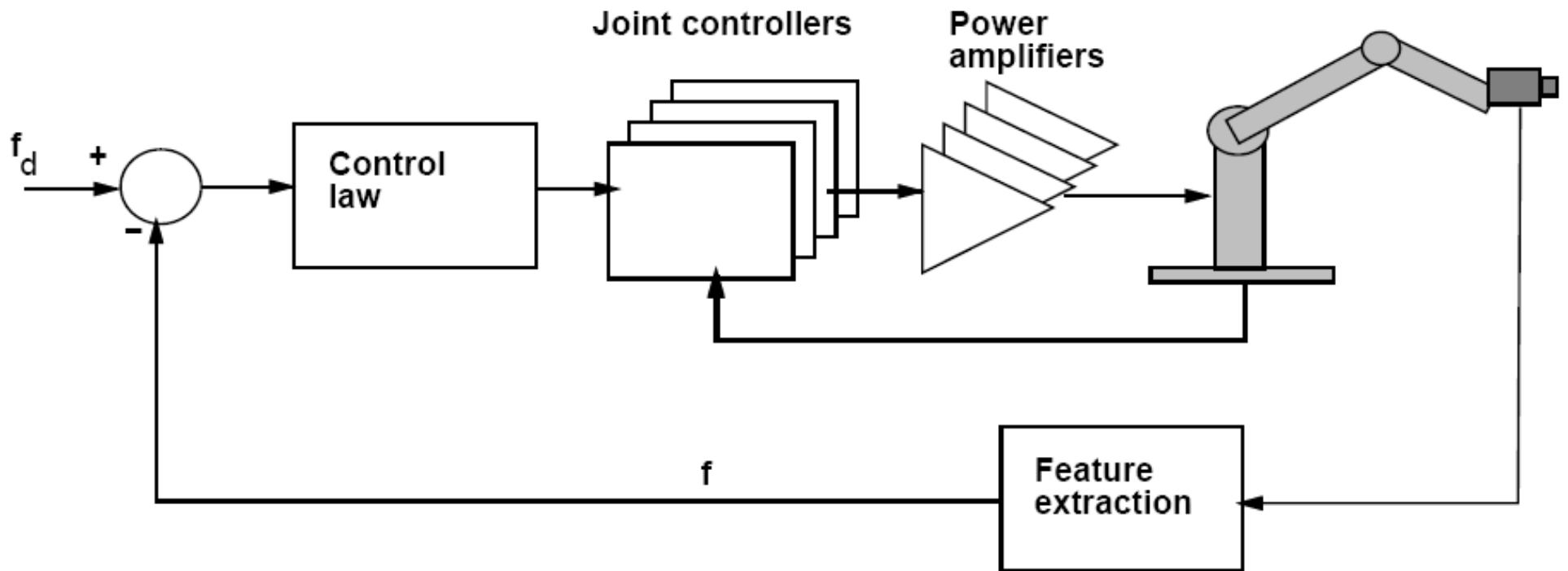


Image Jacobian

$$\dot{\mathbf{f}} = \mathbf{J}_v \dot{\mathbf{r}}$$

$$J_v(\mathbf{r}) = \left[\frac{\partial}{\partial \mathbf{r}} \right] = \begin{bmatrix} \frac{\partial v_1(\mathbf{r})}{\partial r_1} & \cdots & \frac{\partial v_1(\mathbf{r})}{\partial r_m} \\ \vdots & & \vdots \\ \frac{\partial v_k(\mathbf{r})}{\partial r_1} & \cdots & \frac{\partial v_k(\mathbf{r})}{\partial r_m} \end{bmatrix}$$

Example Image Jacobian

$$\begin{bmatrix} \dot{u} \\ \dot{v} \end{bmatrix} = \begin{bmatrix} \frac{\lambda}{z} & 0 & \frac{-u}{z} & \frac{-uv}{\lambda} & \frac{\lambda^2 + u^2}{\lambda} & -v \\ 0 & \frac{\lambda}{z} & \frac{-v}{z} & \frac{-\lambda^2 - v^2}{\lambda} & \frac{uv}{\lambda} & u \end{bmatrix} \begin{bmatrix} T_x \\ T_y \\ T_z \\ \omega_x \\ \omega_y \\ \omega_z \end{bmatrix}$$

Inverting the Image Jacobian

$$\dot{\mathbf{r}} = \mathbf{J}_v^{-1} \dot{\mathbf{f}}$$

$$\dot{\mathbf{r}} = \mathbf{J}_v^+ \dot{\mathbf{f}} + (\mathbf{I} - \mathbf{J}_v^+ \mathbf{J}_v) \mathbf{b}$$

Overconstrained

$$\mathbf{J}_v^+ = (\mathbf{J}_v^T \mathbf{J}_v)^{-1} \mathbf{J}_v^T$$

$$\dot{\mathbf{r}} = \mathbf{J}_v^+ \dot{\mathbf{f}}$$

Underconstrained

$$\mathbf{J}_v^+ = \mathbf{J}_v^T (\mathbf{J}_v \mathbf{J}_v^T)^{-1}$$

$$(\mathbf{I} - \mathbf{J}_v^+ \mathbf{J}_v) \mathbf{b}$$

lie in the null space of \mathbf{J}_v

Resolved rate motion control

$$\mathbf{e}(\mathbf{f}) = \mathbf{f}_d - \mathbf{f}$$

$$\mathbf{u} = \mathbf{J}_v^{-1}(\mathbf{r})\dot{\mathbf{f}}$$

$$\mathbf{u} = \mathbf{K}\mathbf{J}_v^{-1}(\mathbf{r})\mathbf{e}(\mathbf{f})$$

servo without calibration

- Estimate the jacobian online

Estimation of the full Jacobian was solved mathematically by Broyden in the 60's [20], and later rediscovered in robotics by [11, 10, 15]. A first order updating formula which converges to the Jacobian after n linearly independent moves is:

$$\hat{J}_{k+1} = \hat{J}_k + \frac{(\Delta \mathbf{y}_{measured} - \hat{J}_k \mathbf{x}) \mathbf{x}^T}{\mathbf{x}^T \mathbf{x}}$$

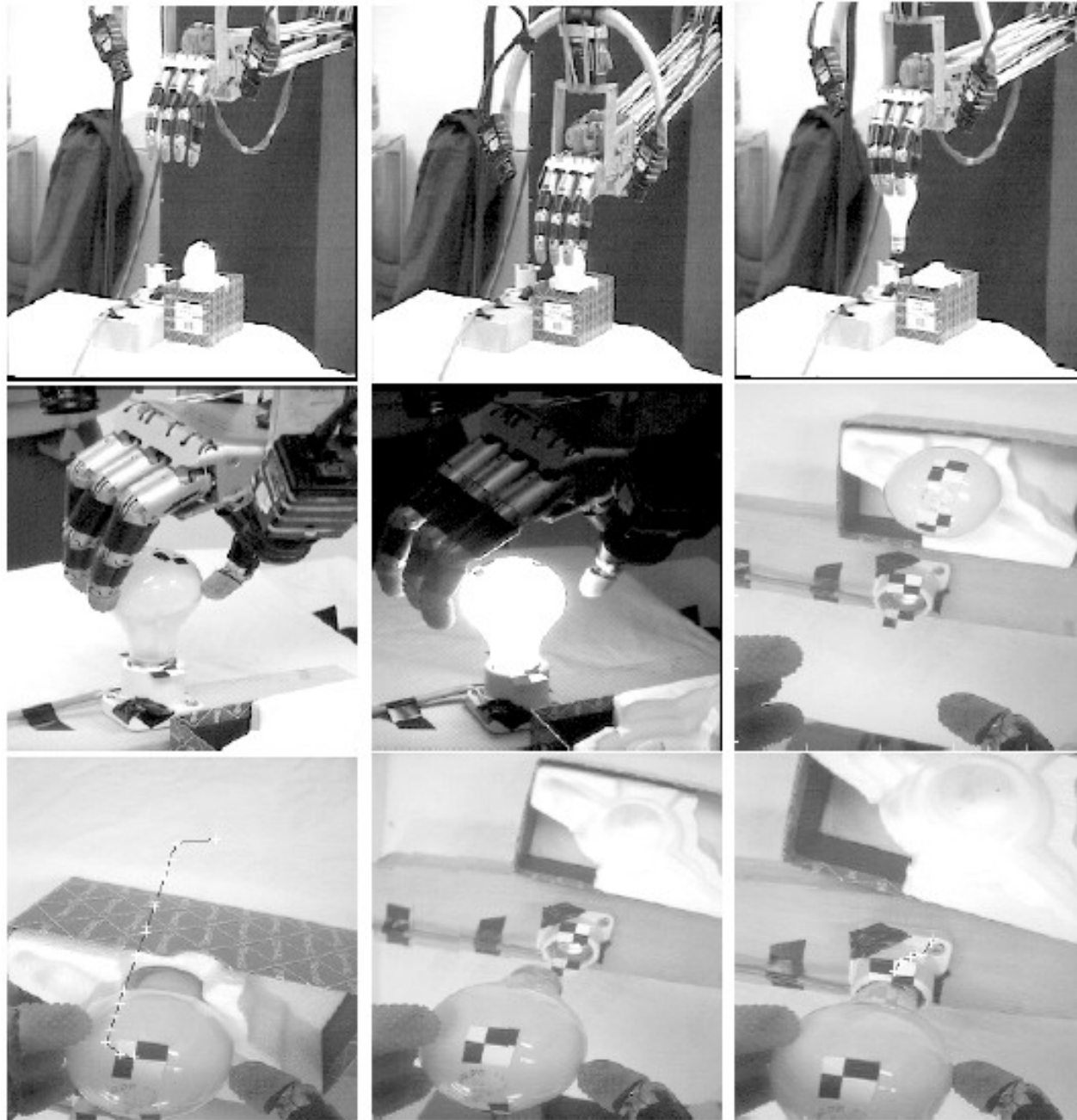
Note that this estimation accepts movements along arbitrary directions \mathbf{x} and thus needs no additional data other than what is available as a part of the manipulation task we want to solve.

Martin Jägersand

University of Rochester -> Yale -> University of Alberta

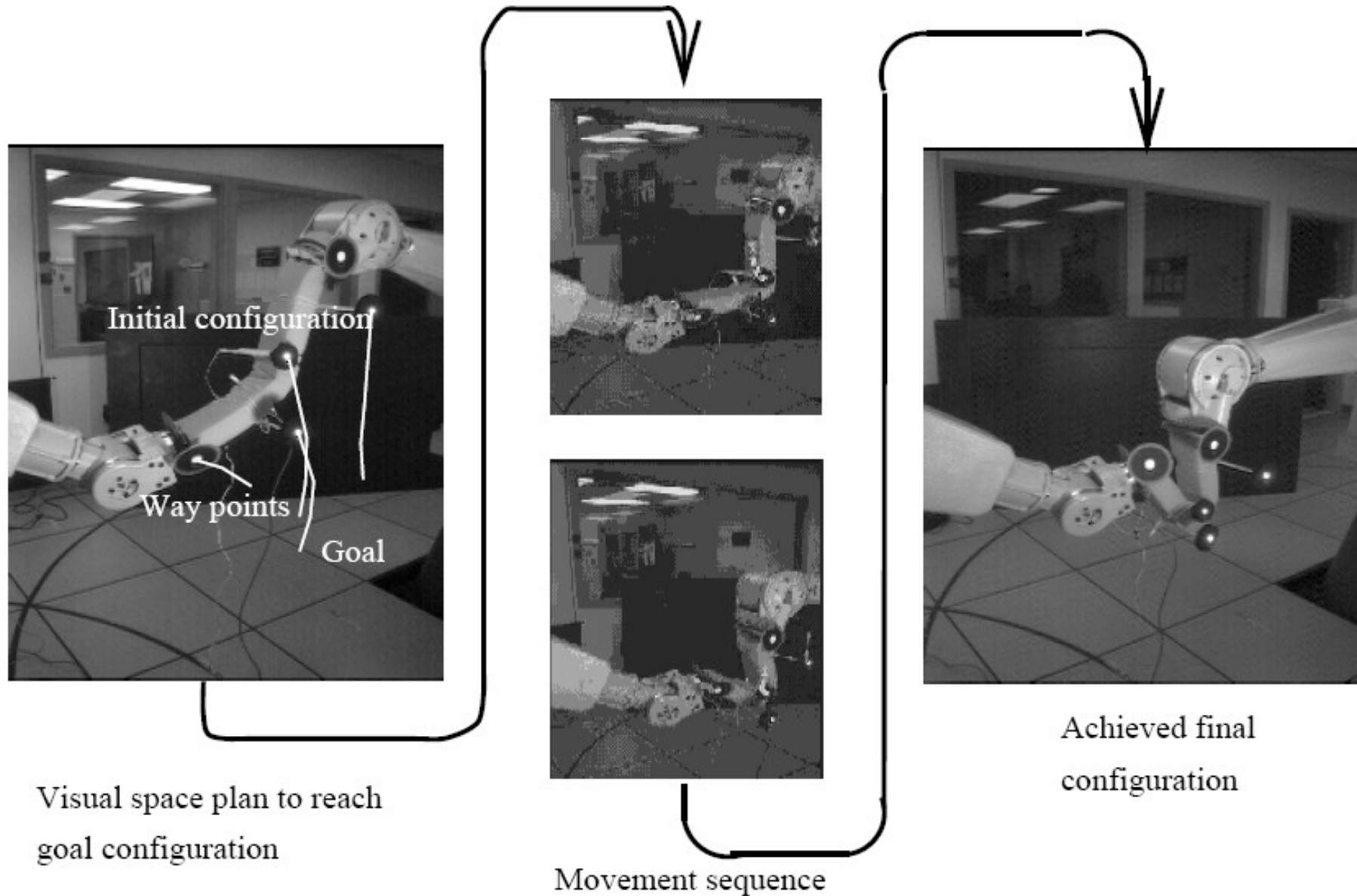
Jägersand M. and Nelson R., **Visual Space Task Specification, Planning and Control**. *In Proc. of IEEE Int. Symp. on Computer Vision 95*, p 521-526, 1995.

“A short description of higher level aspects of uncalibrated visual control. Many experiments solving complex manipulation tasks in unstructured environments.” - Martin Jägersand

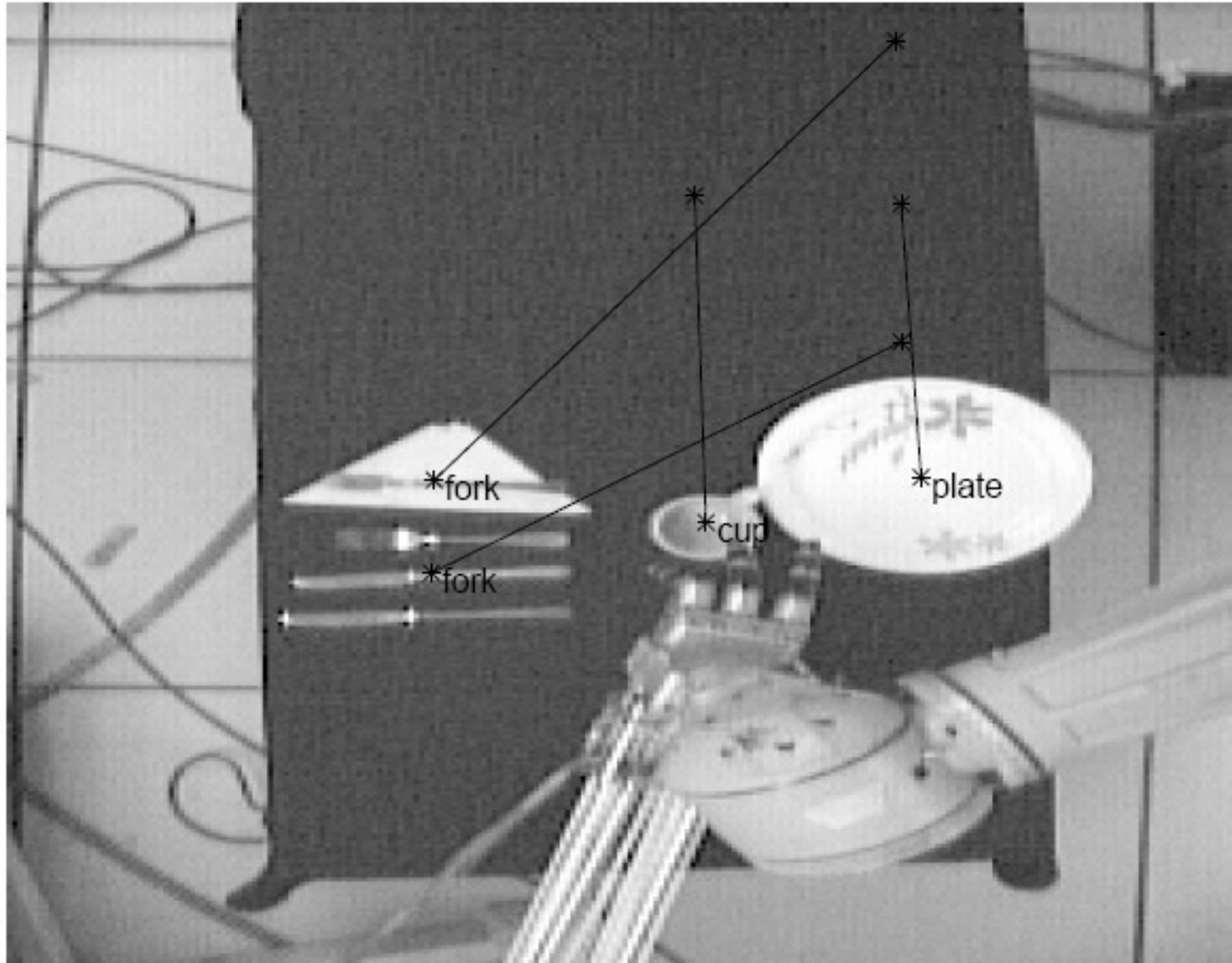


From: Jägersand M. Nelson R. Visual Space Task Specification, Planning and Control
In *Proc. of IEEE Int. Symp. on Computer Vision 95*, p 521-526, 1995.

Non-rigid Manipulation



From: Jägersand M. Nelson R. Visual Space Task Specification, Planning and Control
In *Proc. of IEEE Int. Symp. on Computer Vision 95*, p 521-526, 1995.



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In *Proc. of IEEE Int. Symp. on Computer Vision 95*, p 521-526, 1995.

Discussion questions?

- When is visual servoing a good idea?
- When is visual servoing a bad idea?

What will you do?



Next week

- You're the presenters!

Extra

Velocity of a Rigid Object

$$\mathbf{T}(t) = [T_x(t), T_y(t), T_z(t)]^T$$

$$\boldsymbol{\Omega}(t) = [\omega_x(t), \omega_y(t), \omega_z(t)]^T$$

$$\dot{\mathbf{r}} = \begin{bmatrix} T_x \\ T_y \\ T_z \\ \omega_x \\ \omega_y \\ \omega_z \end{bmatrix}$$

Velocity screw

$$\dot{\mathbf{P}} = A(\mathbf{P})\dot{\mathbf{r}}$$

$$A(\mathbf{P}) = [I_3 \mid -\text{sk}(\mathbf{P})]$$

$$\dot{\mathbf{P}} = \boldsymbol{\Omega} \times \mathbf{P} + \mathbf{T}$$

$$\text{sk}(\mathbf{P}) = \begin{bmatrix} 0 & -z & y \\ z & 0 & -x \\ -y & x & 0 \end{bmatrix}$$

$$\dot{\mathbf{P}} = -\text{sk}(\mathbf{P})\boldsymbol{\Omega} + \mathbf{T}$$

Velocity of a Rigid Object (continued)

$${}^e \dot{\mathbf{r}} = [{}^e \mathbf{T}; {}^e \boldsymbol{\Omega}]$$

$$\dot{\mathbf{r}} = \begin{bmatrix} \boldsymbol{\Omega} \\ \mathbf{T} \end{bmatrix} = \begin{bmatrix} \mathbf{R}_e {}^e \boldsymbol{\Omega} \\ \mathbf{R}_e {}^e \mathbf{T} - {}^e \boldsymbol{\Omega} \times \mathbf{t}_e \end{bmatrix}$$

$$\dot{\mathbf{P}} = A(\mathbf{x}_e \circ {}^e \mathbf{P}) \dot{\mathbf{r}}$$