

## AIBO Camera Stabilization

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### AIBO vision is *bumpy*

- Legged locomotion induces vibration.
- Head (camera mount) is a great big cantilevered mass.



## Camera problems: the lineup

- Obvious problems due to exposure time/cheap optics:



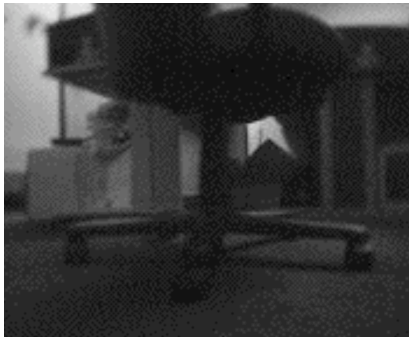
*Smear*



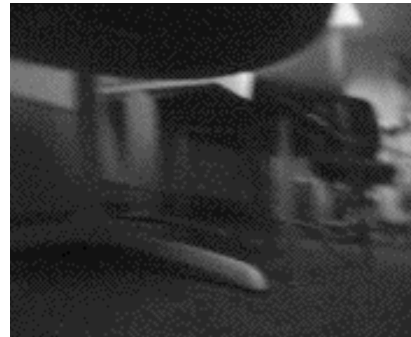
*Lens distortion*

## Camera problems: the lineup

- Subtle problems due to sample rate:



*Stretching*



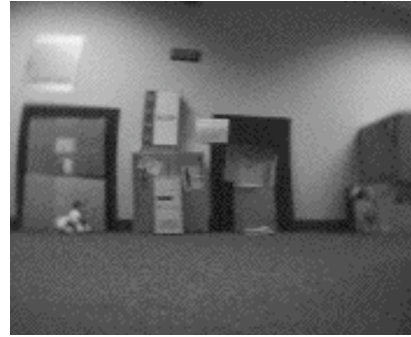
*Skew/bending*

## Camera problems: the lineup

- Subtle problems due to sample rate:



*Stretching*



*Skew/bending*

**Goal: Take AIBO from this...**



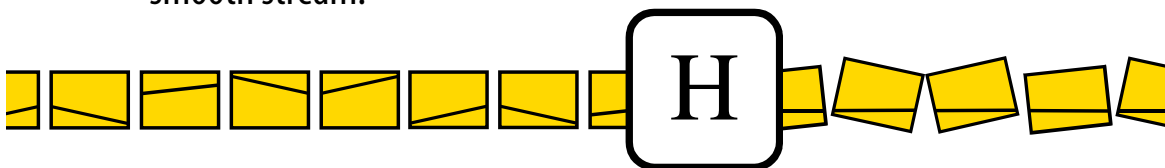
...to this:

*Smoooooooooth...*



## Stabilization Basics

- Compute homographies between successive images in your sequence.
- Transform sequence images one by one to make a continuous, smooth stream.



- Problems: error accumulates, especially with more degrees of freedom (e.g. affine transformations).
  - Many papers are about dealing with this.

## **Mosaicing Video Sequences**

### **Netzer, Gotsman**

- Suggests using a sliding window of multiple images to compute more accurate registration of each frame
  - We ignore this, too computationally intensive, introduces lag in images
- Treat each new image as one of translation, rigid, similarity, affine, or projective transformation. Try each, pick the one with the lowest error, with some bonus to “simpler” transformations.
  - Instead of trying all 5 on each frame at run time, we did some trials and found rigid transformations satisfactory.

## **Camera Stabilization Based on 2.5D Motion Estimation and Inertial Motion Filtering**

### **Zhu, Xu, Yang, Jin**

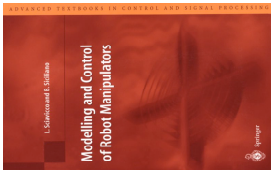
- Typically, camera movements fall into a few classes of motion. (e.g. panning, dolly, tracking, ...) We can pass through movement on the dominant dimension and stabilize on the non-dominant dimension.
  - Since our motions are typically constrained to the horizontal plane, we can compensate for vertical bouncing and rotation, but leave horizontal motion unchecked.

## Our approach

- AIBO vibration is very regular.
- Rotation oscillates around  $0^\circ$ .
- Vertical bouncing oscillates around a fixed value.
- Horizontal bouncing oscillates around a value determined by AIBO's turning and sidestepping velocities (which we know!).



**So...**

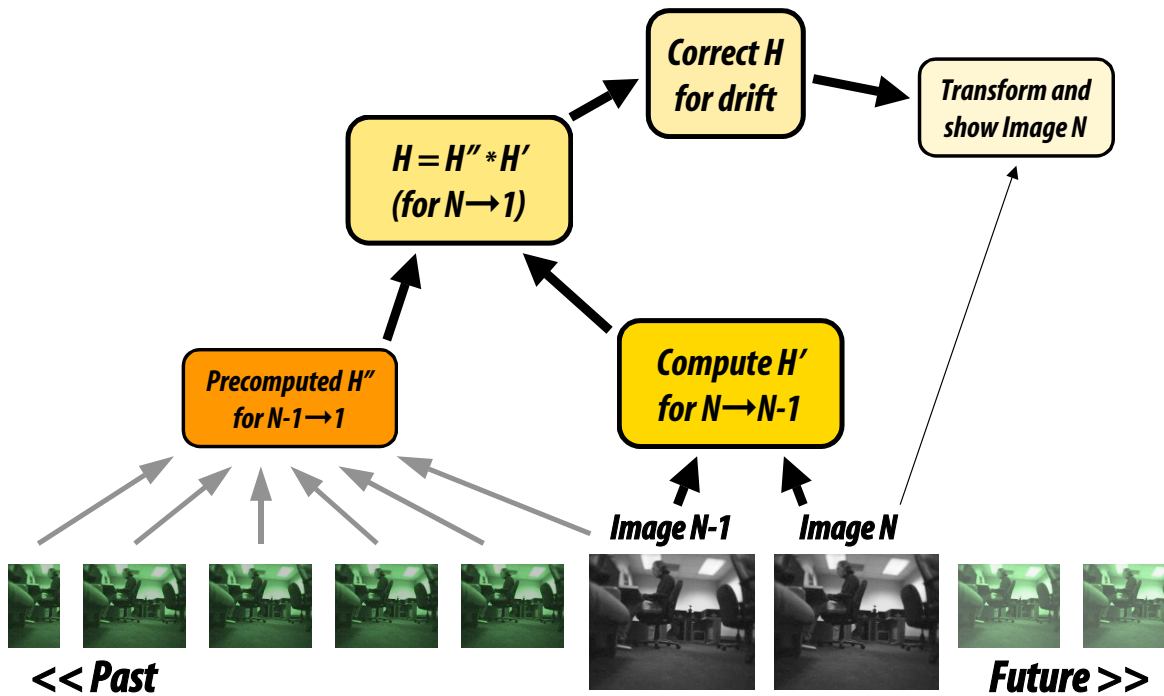


### It's a control problem now!



- Over many frames, image motion should tend toward fixed (or predictable) values.
- Use an “image placement controller” that allows high-frequency changes in placement, but enforces this constraint.
- Specifically, we extract and adjust the  $x, y$  coordinates and  $\theta$  rotation used for image registration.

## The obligatory flowchart



## How to find corresponding points

- **Q:** How do you find corresponding points in Image N and Image N-1?
- **A:** Andrew and Ranjith's RANSAC from Assignment 5.
- **Q:** Oh.
- **A:** It's pretty robust, even to blurry, smeared images.

## Why find H indirectly?

- We could simply find the H from Image N to the **corrected, transformed** Image N-1, right?
- **Wrong-o!** The corrected image is jagged, noisy. RANSAC would **freak**.
- Instead, first find H' from Image N to the **normal** Image N-1.
- Then premultiply it with the accumulated H (H'') from the normal Image N-1 to Image 1.

## How do we get x, y, and $\theta$ ?

- It's easy if our "homography" is just a rigid transform.

$$\begin{array}{c} \text{arccotan} \rightarrow \theta \\ \left[ \begin{array}{ccc} \cos \theta & -\sin \theta & t_x \\ \sin \theta & \cos \theta & t_y \\ 0 & 0 & 1 \end{array} \right] \end{array}$$

- It's easy to adjust them, too.
- A cop out? Perhaps... it doesn't fix all the aberrations in the AIBO image. It's a start, though.



## Finding rigid transformations (1 of 2)

- Step 1: Constrained least squares.

**P: Image N point**  
**u,v: N+1 point**

$$\begin{bmatrix} \cos \theta & -\sin \theta & t_x \\ \sin \theta & \cos \theta & t_y \\ 0 & 0 & 1 \end{bmatrix} \implies \begin{bmatrix} a & -b & c \\ b & a & d \\ 0 & 0 & e \end{bmatrix}$$

$$\begin{aligned} -m_1^\top P_i + u_i m_3^\top P_i &= 0 & \implies & -aP_i(1) + bP_i(2) - cP_i(3) + 0 + u_i e P_i(3) = 0 \\ -m_2^\top P_i + v_i m_3^\top P_i &= 0 & \implies & -bP_i(1) - aP_i(2) + 0 - cP_i(3) + v_i e P_i(3) = 0 \end{aligned}$$

$$\begin{bmatrix} \ddots & & & & & \\ -P_i(1) & P_i(2) & -P_i(3) & 0 & u_i P_i(3) & \\ -P_i(2) & -P_i(1) & 0 & -P_i(3) & v_i P_i(3) & \\ & & \ddots & & & \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \\ e \end{bmatrix} = \mathbf{0}$$

**The U matrix**

## Finding rigid transformations (2 of 2)

- Step 2: Throw away image scaling.

$$\begin{bmatrix} a & -b & c \\ b & a & d \\ 0 & 0 & e \end{bmatrix}$$

**[a/e b/e] is not constrained by constrained least squares to be of length 1.**

- Divide  $a, b, c, d,$  and  $e$  by  $e$ , then by the length of  $[a \ b]$ .
- Otherwise the image will shrink as you walk forward.

## Fighting drift, step by step

- Isolate  $\theta$  from H
- Apply really simple correction: premultiply H by a rotation matrix that rotates by  $-\theta/\text{constant}$  (forcing it back toward 0).
- Isolate  $t_x$  and  $t_y$  from H
- Apply similar correction. We should force  $t_y$  toward a predicted value based on turning and sidestepping. We don't right now, forcing it to 0 instead.

## Demo time!

- Hallway scene
  - Normal, stabilized, and side-by-side
- Lab scene
  - Normal, stabilized, and side-by-side



**Any questions?**