

# Towards Visual SLAM in Dynamic Environments

MIT 16.412J Spring 2004

Vikash K. Mansinghka

## Key Points

- Visual metric SLAM in environments with motion is an open problem
- SIFT SLAM error handling makes odd choices:
  - Cheats on position estimation (by trusting vision) - see report
  - Sets arbitrary matching thresholds for vision
- These choices impact SLAM performance:
  - Imposes a limit on motion error (odometry vs. actual)
  - Slow-moving objects can arbitrarily foul localization
  - Matching thresholds may not be best (though “good enough”)
- Partial solution: Minimize uncertainty and use optical flow

## SIFT Landmark Tracking

- Predict where landmarks should appear (reliability, speed)
- Note: Robot moves in  $xz$  plane
- Given  $[p, q, \delta]$  and old relative position  $[X, Y, Z]$ , find expected position  $[X', Y', Z']$  by:

$$X' = (X - p)\cos(\delta) - (Z - q)\sin(\delta)$$

$$Y' = Y$$

$$Z' = (X - p)\sin(\delta) - (Z - q)\cos(\delta)$$

- By pinhole camera model ( $(u_0, v_0)$  image center coords,  $I$  interocular distance,  $f$  focal length):

$$r' = v_0 - f \frac{Y'}{Z'}$$

$$c' = u_0 + f \frac{X'}{Z'}$$

$$d' = f \frac{I}{Z'}$$

$$\sigma' = \sigma \frac{Z}{Z'}$$

## SIFT Landmark Tracking

- $V$  is camera field of view angle (60 degrees)

- A landmark is expected to be in view if:

$$Z' > 0$$

$$\tan^{-1}\left(\frac{|X'|}{Z'}\right) < \frac{V}{2}$$

$$\tan^{-1}\left(\frac{|Y'|}{Z'}\right) < \frac{V}{2}$$

- An expected landmark matches an observed landmark if:
  - Obs. center within a 10x10 region around expected
  - Obs. scale within 20% of expected
  - Obs. orientation within 20 degrees of expected
  - Obs. disparity within 20% of expected

## Impact of Loose Matching Tolerance



- Mario does not know landmarks on Wario come from same source
- Robot (Mario) confused as long as Wario moves slowly

### Possible Solution

- Add landmarks only when sensor error very low (robot stationary)
- Do not accept landmarks from image regions with motion
- Approximate motion either via SIFT matching or other optical flow scheme
- Alternately: Only trust largest set of consistent errors (how?)

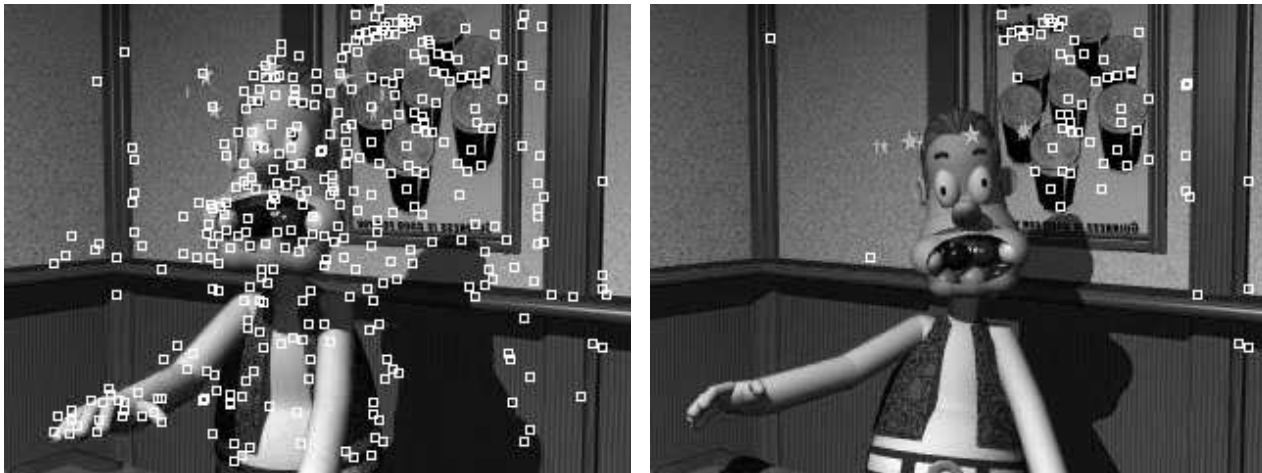
## Horn-Schunck Optical Flow

- Let  $E(x, y, t)$  be image brightness at pixel  $(x, y)$  at time  $t$
- Goal: find  $u = \frac{dx}{dt}$  and  $v = \frac{dy}{dt}$  (horiz, vert flow)
- Assume  $\frac{\partial E}{\partial x} \frac{dx}{dt} + \frac{\partial E}{\partial y} \frac{dy}{dt} + \frac{\partial E}{\partial t} = 0$
- Also minimize measures of the departure from smoothness and above assumption
- One set of choices yields coupled PDEs:  
 $\nabla^2 u = \lambda(E_x u + E_y v + E_t)E_x$  and  $\nabla^2 v = \lambda(E_x u + E_y v + E_t)E_y$

## Proof of Concept for Optical Flow

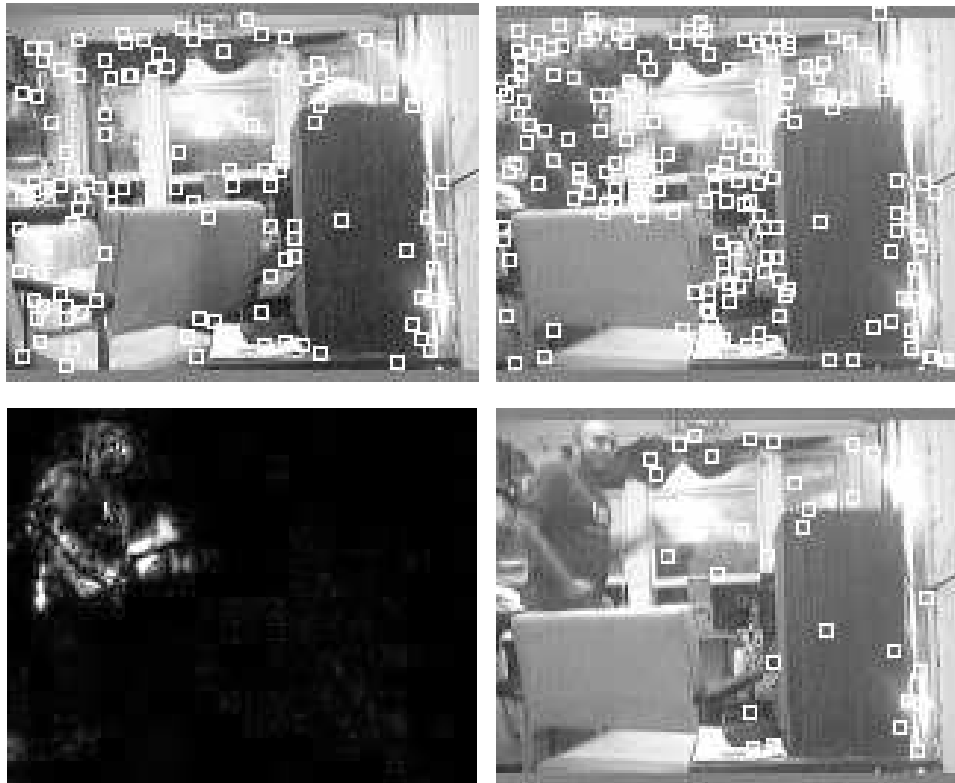
- Uses Horn-Schunck optical flow; could use SIFT matching
- First demo prunes landmarks which move/are missed too much
- Second demo attempts place recognition by ignoring features on moving image regions

## Simple Results



- A restrictive optical flow threshold (0.1) reliably prunes features on moving objects

## Place Recognition Results



- Matching threshold of 20%
- Lowering the optical flow threshold from 30 to 1 raised performance from 43% to 56% in an image with motion
- Low performance due primarily to naive place recognition approach

## Contributions

- Identified limitation in SIFT error handling; probably present in other SLAM systems
- Suggested simple optical flow technique for enabling dynamic visual SLAM
- Suggested other techniques for improving visual SLAM performance (see report)