

# See, AIBO; Run!

## AIBO Motion and Vision Algorithms



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**AIBO:**



(COURTESY SONY)

Come, Dick.  
Come and see.  
Come, come.  
Come and see.  
Come and see Spot.

**The Dog...  
The Legend.**

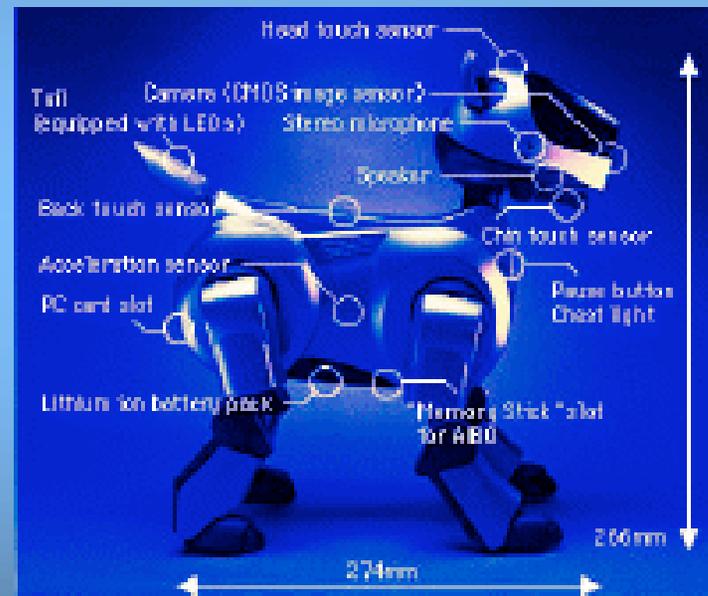
Look, Spot.  
Oh, look.  
Look and see.  
Oh, see.

# History of AIBO

- **Sony's Entertainment Robot**
- **AIBO – Artificially Intelligent roBOt**
- **Or aibo – Japanese for “companion”**
- **Originally intended for purchase by home users**
- **Found to be a relatively cheap, versatile platform that could be used by educators and researchers**

# AIBO: Technical Specifications

- 64 bit Processor
- 20 Degrees of Freedom
- Microphone
- Accelerometer
- Infrared Distance Sensor
- Pressure Sensors
- The Kitchen Sink



# Tekkotsu

- Application framework for AIBO
- Under development at CMU's Robotics Lab
- TekkotsuMon - server-side interface to code running on robot
- To accomplish our goals, we built upon Tekkotsu platform
- Our advisor, Ethan Tira-Thompson, is a chief researcher for the project

# *Run, AIBO, Run!*

- Originally attempted to stabilize image from AIBO's camera
  - Software methods
  - New walking motions
- Modified walk parameters
- Measured performance using accelerometers
- Stability vs. Speed



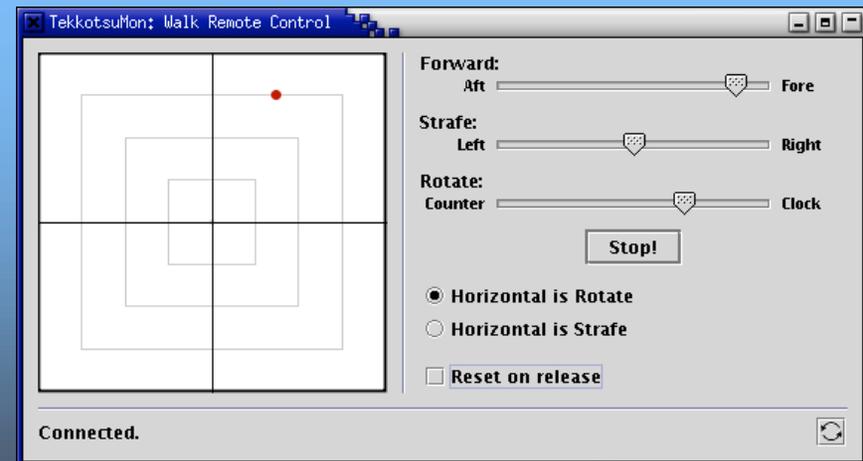
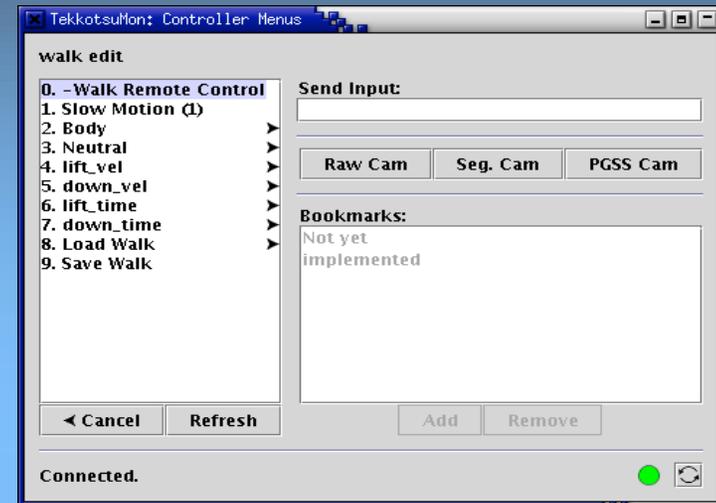
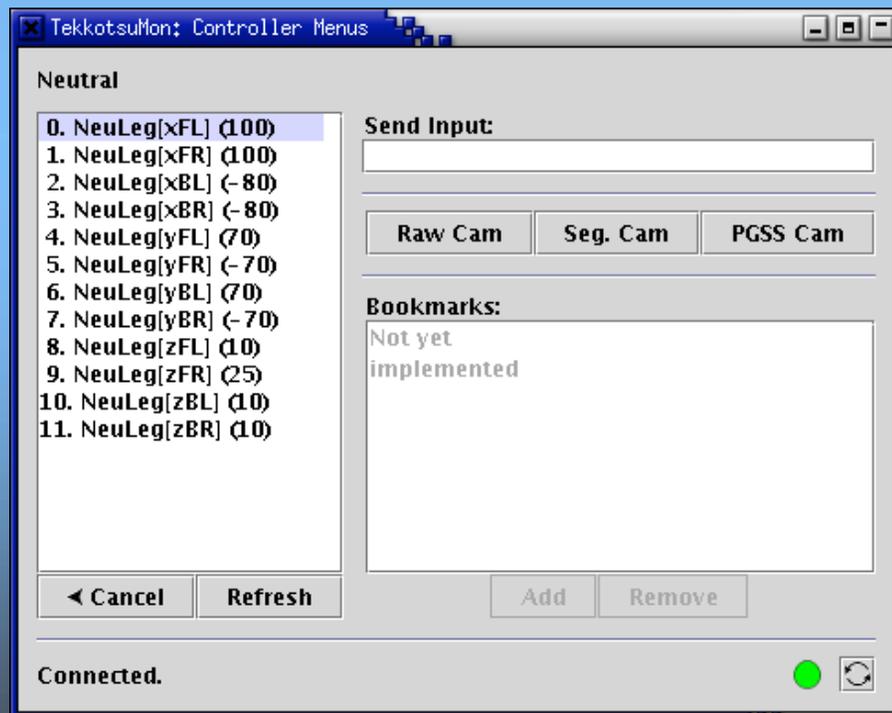
# How Our Walk Was Developed

- Tekkotsu's default walk is taken from CMU's RoboCup soccer team
- Our new motion was created by modifying this walk's parameters



# Get Your Move On!

- Using the TekkotsuMon GUI



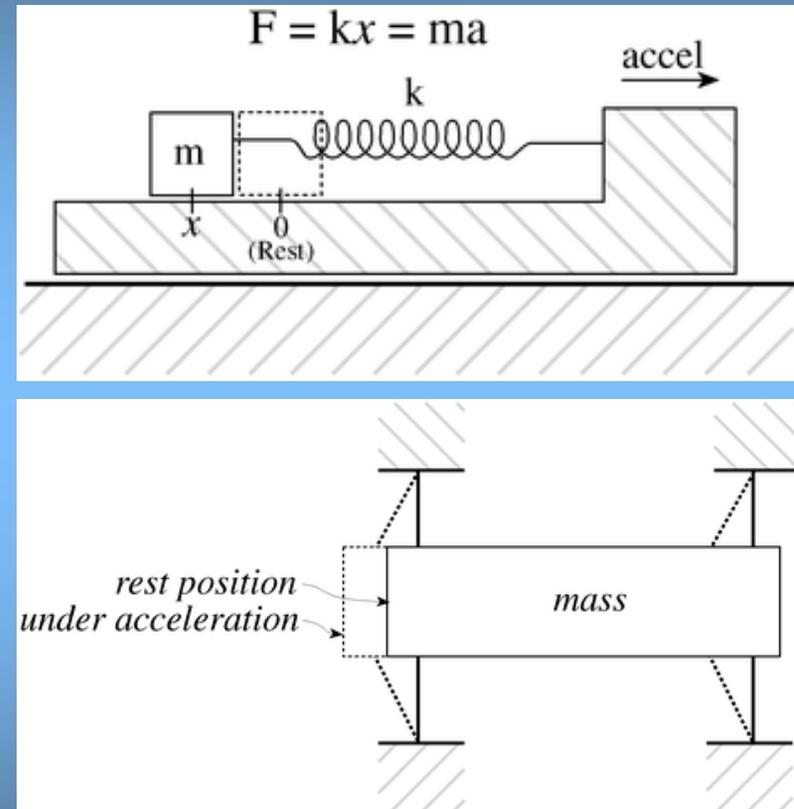
# Changing Things Up: The Parameters

- Lift Velocity
- Down velocity
- Lift Time
- Down time
- Body Height/Angle
- Period
- Position Coordinates
- Hop and Sway



# Accelerometers

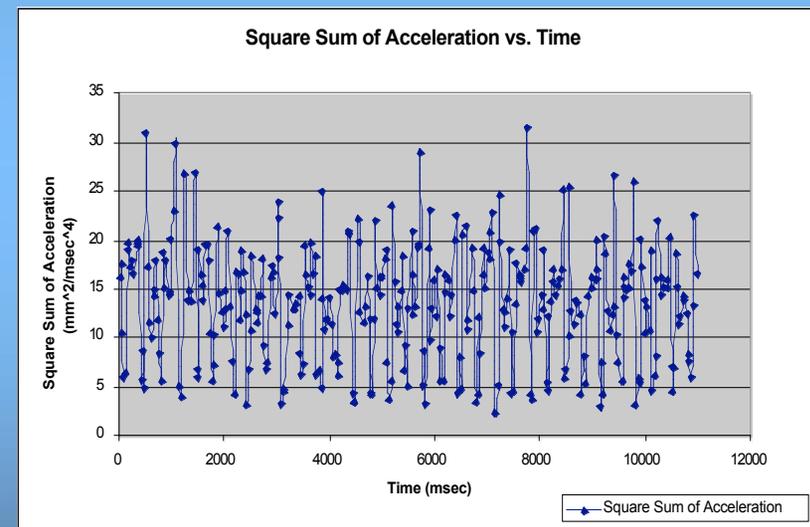
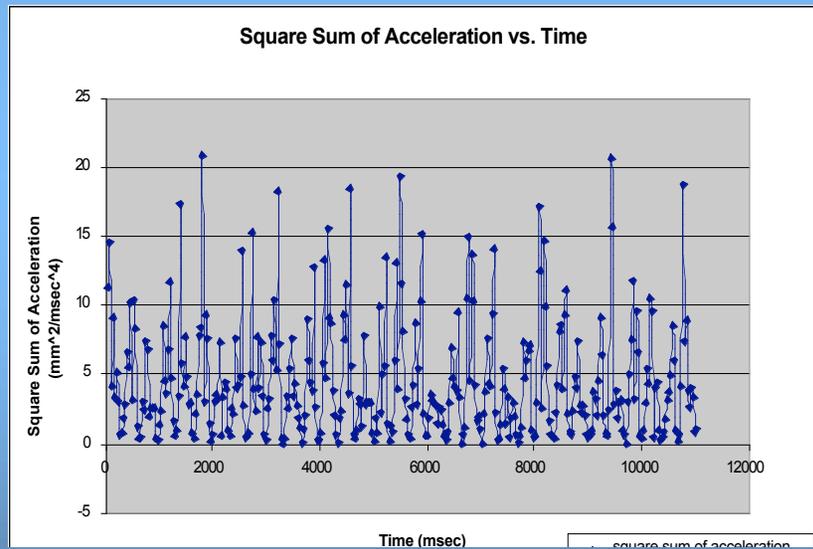
- Each consists of a mass/spring system
- Measure force and displacement on joints as robot walks
- We graphed the way force varies with time to evaluate stability of new walking algorithm



# Our Upright Walk vs. CMU RoboCup Walk

- Upright

- RoboCup

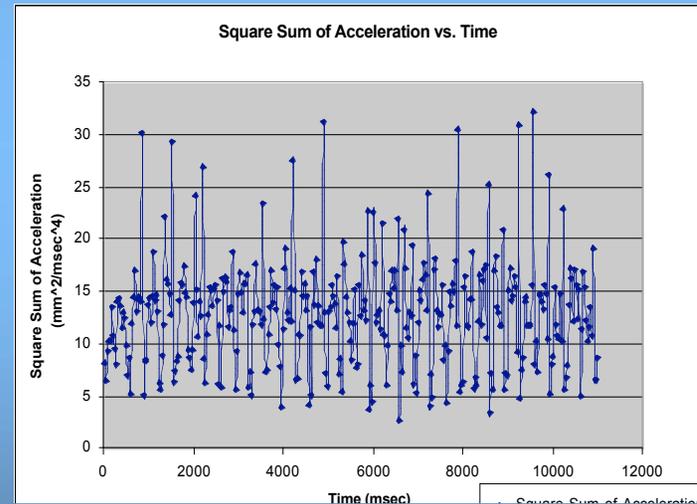
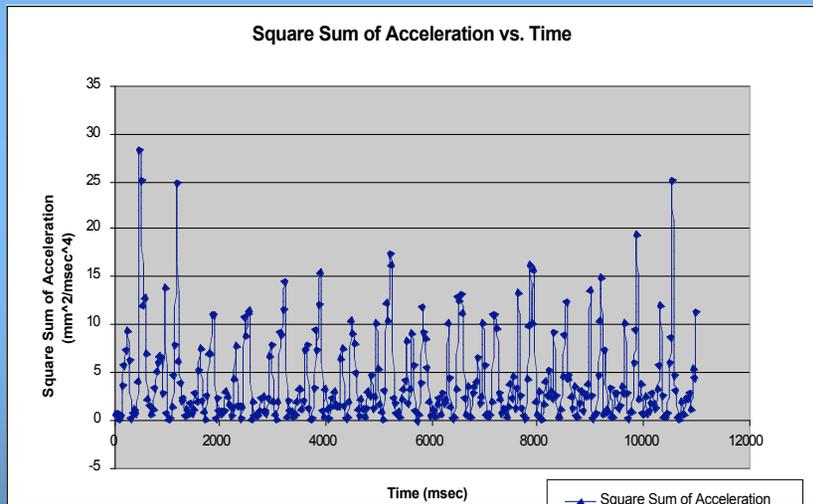


Quarter Speed

# Our Upright Walk vs. CMU RoboCup Walk

- Upright

- RoboCup

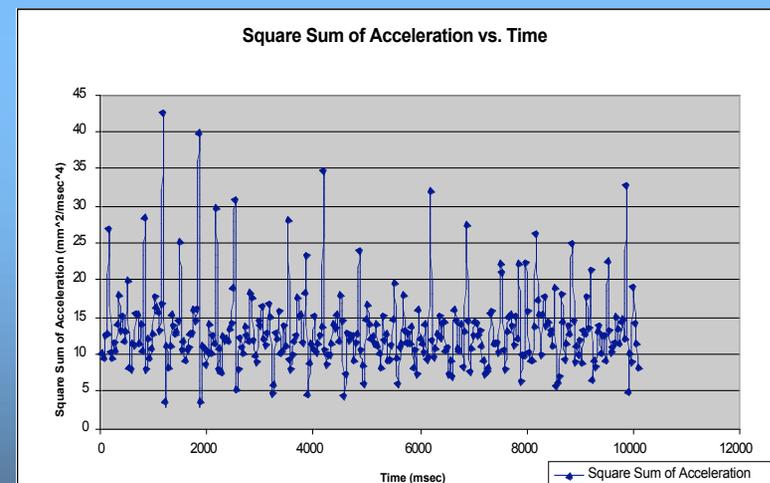
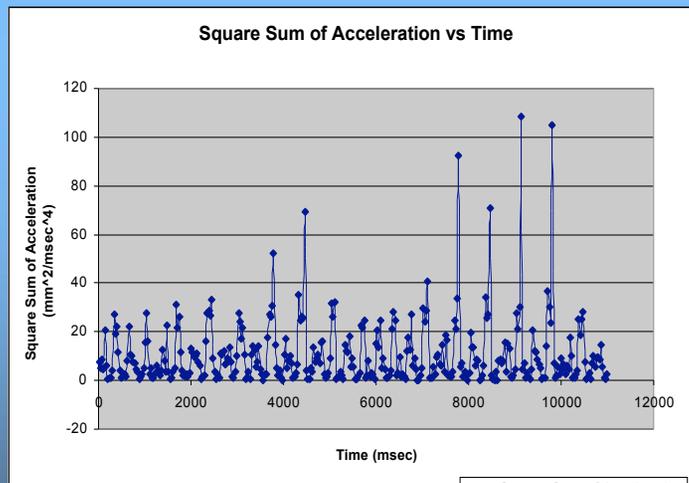


Half Speed

# Our Upright Walk vs. CMU RoboCup Walk

- Upright

- RoboCup

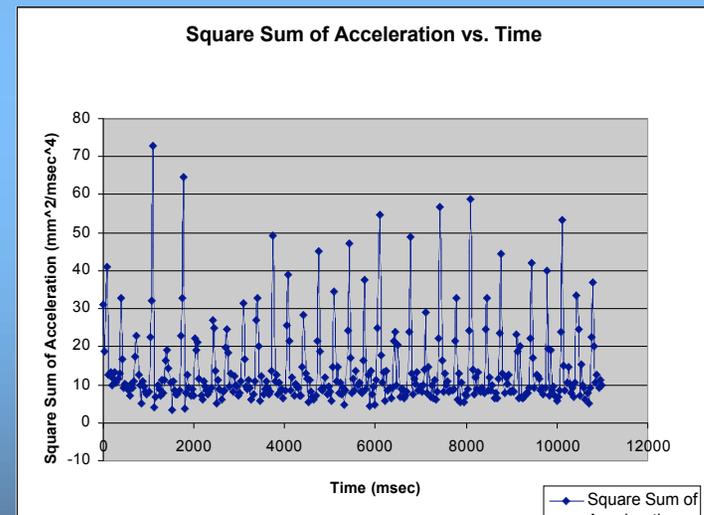
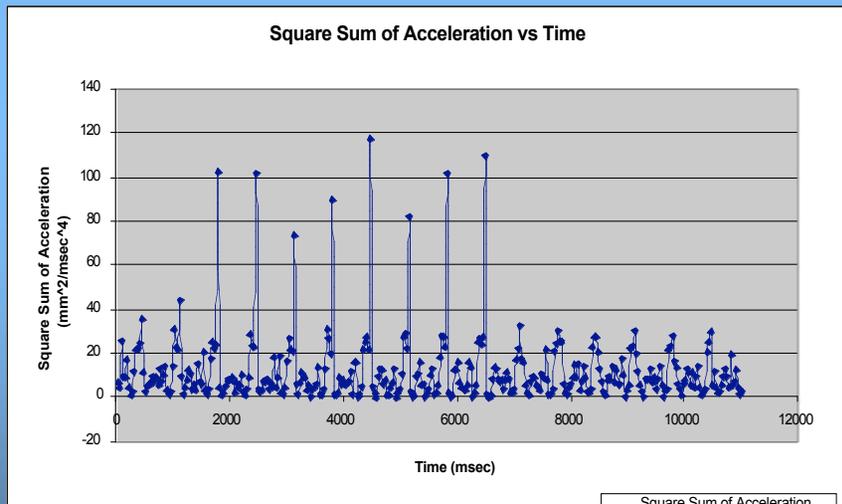


Three Quarter Speed

# Our Upright Walk vs. CMU RoboCup Walk

- Upright

- RoboCup



Full Speed

# See, AIBO, See!

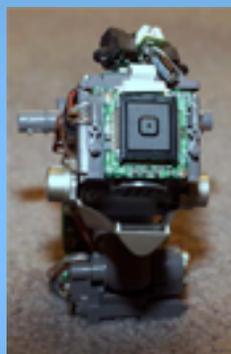
- Developed an algorithm that allows AIBO to follow a pink line
- Gradually improved algorithm based on perceived weaknesses



# How does AIBO see?



**AIBO**



**CCD Camera  
(YUV)**



**Raw Image  
Translated  
on  
Computer  
(RGB)**

# Segmented Vision

- AIBO camera captures YUV format
- Bitmapped images are too large to send over network efficiently, so the images must be compressed

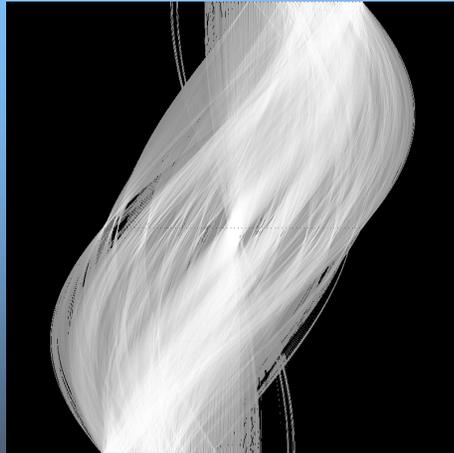
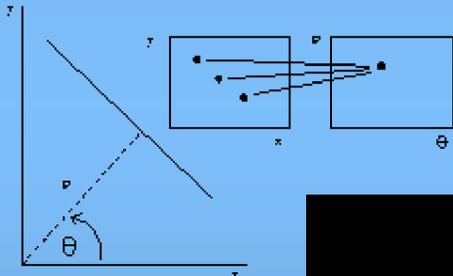


# RLE (Run-Length Encoding)

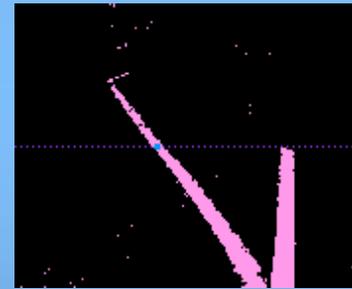
- The data is then converted into a series of color run triplets
- Triplets are sent to the computer, where they are placed into an array
- Vision segmentation and run-length encoding is performed on board the AIBO

# Line Following Algorithms

## Hough Transformation



VS.



## Hack Algorithm

# Hough Transformation

- Represent a line in an image in a different way: Parameter Space
- Example – a line in image space can be represented as:

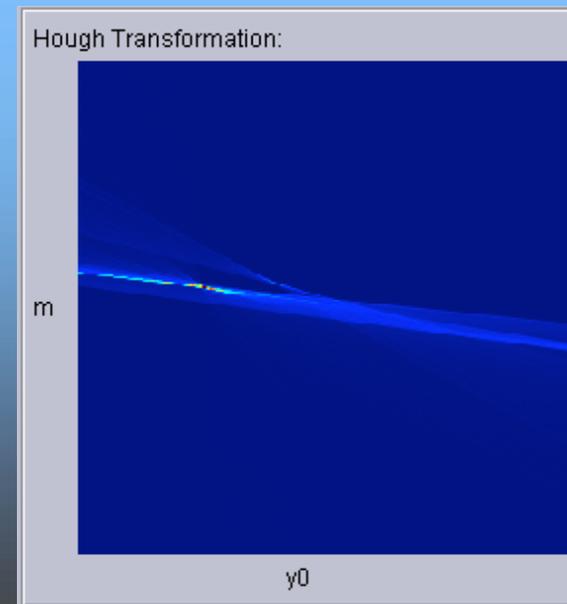
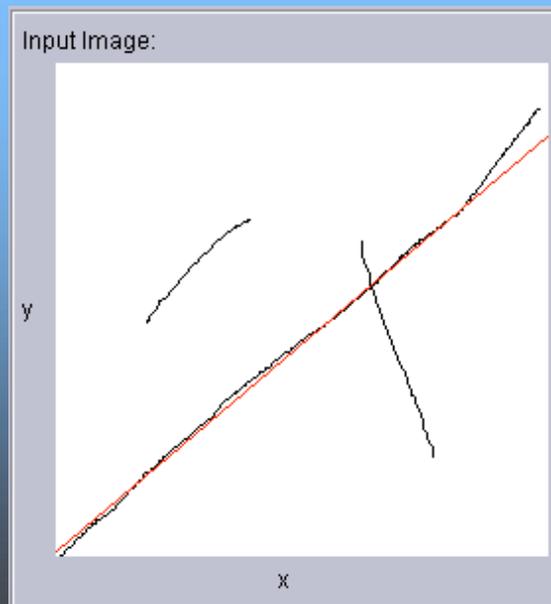
$$y = mx + y_0$$

- The Hough method “transforms” this equation to parameter space:

$$y_0 = y - mx$$

# Hough Transformation (cont'd)

- A discrete parameter space called an *accumulator* is created. All points in the original image are converted using Hough Transform into this parameter space.
- Points with the same slope and y-intercept are accumulated into the same cell of the accumulator.
- The highest cell is found, thus finding the most prominent line (marked in red).



# Hough Transformation (cont'd)

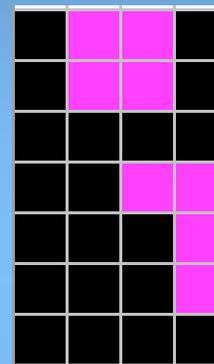
- **Advantages:**
  - Will almost always find the most prominent line
  - Ignores static and foreign objects (unless they have defined edges)
- **Disadvantages:**
  - Implementation in a robot is too computationally expensive
  - Processes too slow for a real-time image, like the AIBO's

# A Hack Gains Greatness

- A *hack* is defined as an inelegant and usually temporary solution to a problem
- Ironically, the hack line following algorithm we developed (with the help of Alok Ladsariya) became our best solution.

# The Basic Line Following Algorithm

- Take in decoded segmented vision
- Create RegionMap
  - First each pixel is set to pink or not-pink
  - Give each pink region a unique number... remember PaintBucket?
- Turn toward largest region on horizontal center line



-1	0	0	-1
-1	0	0	-1
-1	-1	-1	-1
-1	-1	0	0
-1	-1	-1	0
-1	-1	-1	0
-1	-1	-1	-1

-1	A	A	-1
-1	A	A	-1
-1	-1	-1	-1
-1	-1	B	B
-1	-1	-1	B
-1	-1	-1	B
-1	-1	-1	-1

Legend	
-1	Not Pink
0	Pink
A	1st Region
B	2nd Region

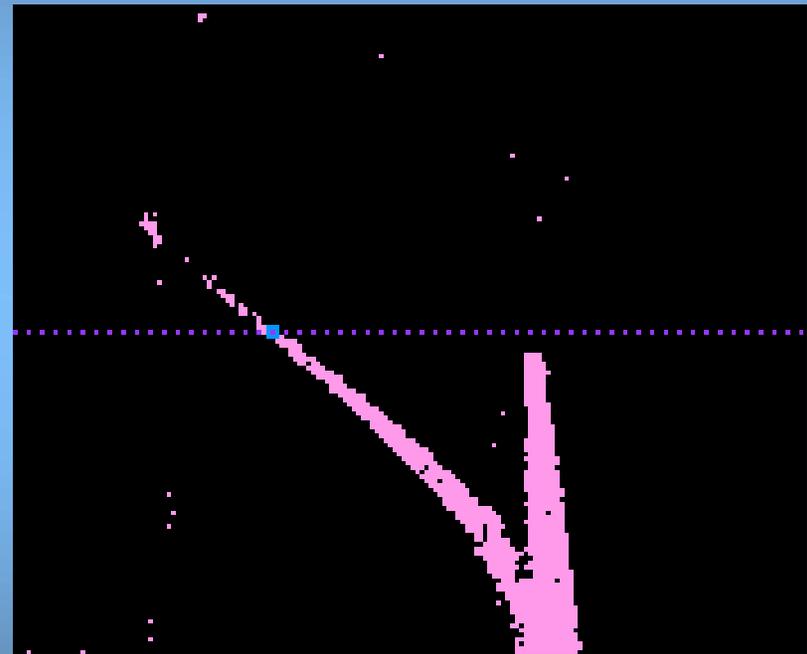
# The Basic Line-Following Algorithm

- **Primary target** – displayed as blue dot
  - at intersection of largest region & center row
- **Direction adjustment**
  - Turns L if blue dot to L of center column
  - Turns R if blue dot to R of center column
- **A few fundamental assumptions allow us to keep the algorithm simple**
  - The line will be the largest region
  - The line will cross the center row once

# Improvements

**Definition: Lost = no regions exist in center row**

- **If lost → stop forward motion & rotate**
  - determine rotation direction by which side of image contained dot more commonly
    - In last 15 frames
- **Adjust direction with speed proportional to blue dot's distance from center column**

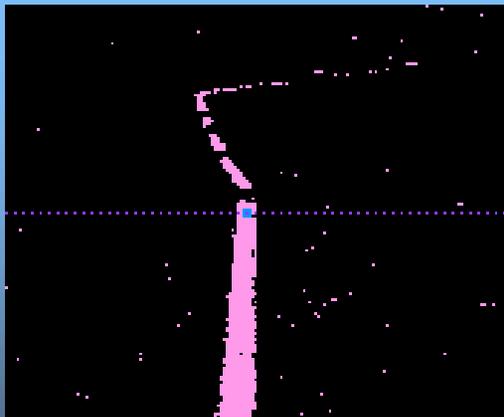


# Algorithm Issues

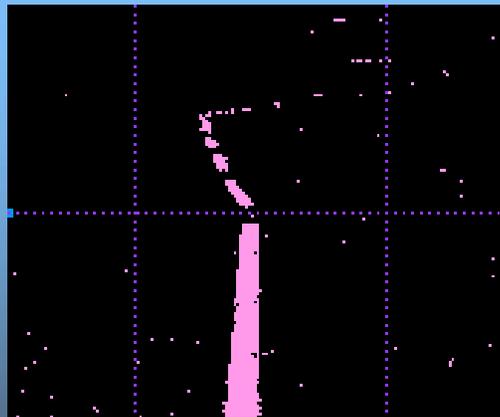


# Algorithm Issues

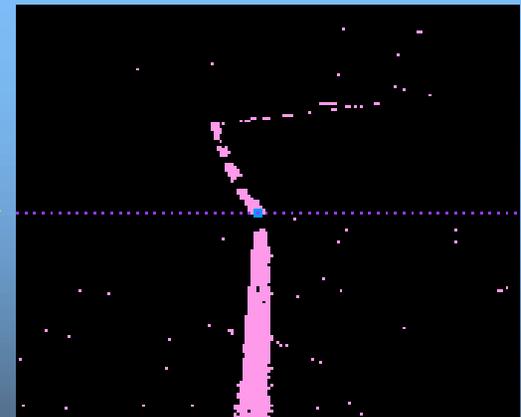
- **A Dashed Line**
  - The breaks in the line trigger the 'lost' behavior briefly
  - But even so, the break is not long enough to lose the dog completely.



Walking along happily...



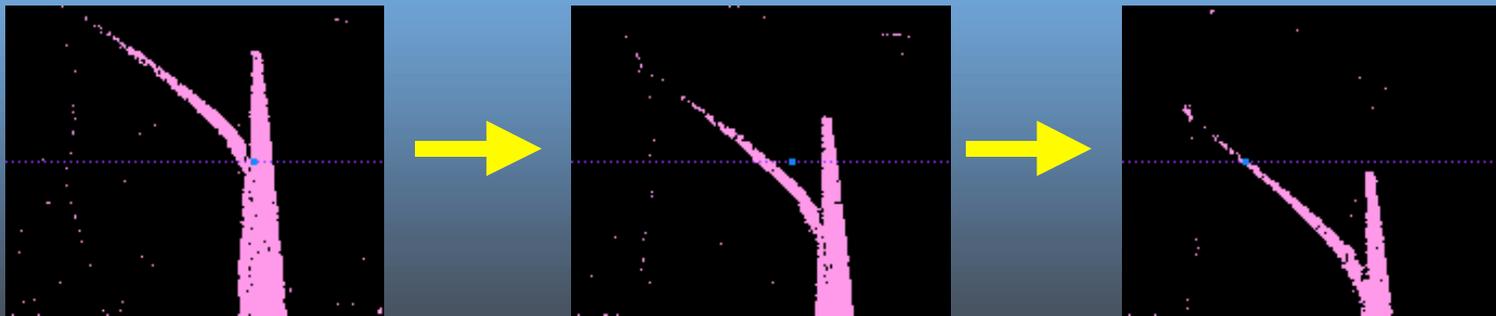
AHH!! BREAK IN THE LINE!!



Found line, walking along happily...

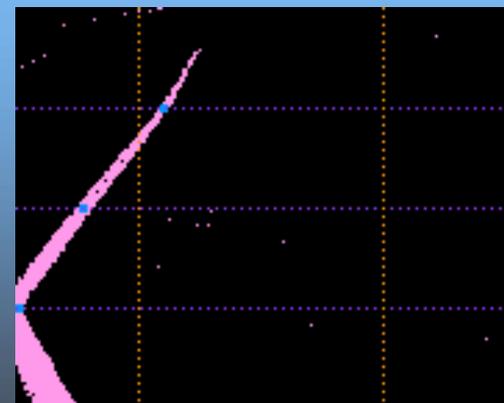
# Algorithm Issues

- **Branching lines:**
  - branch recognized as one region
  - the average taken of all x values of region's pixels in center row
  - once the shorter branch gets below center row, target jumps to longer branch (line)



# Algorithm Issues

- Dog can lose line when
  - Special case: line slopes toward center of image
- Slope-Opposite-Direction Method
  - Two more blue dots
    - Above and below center
  - used to find slope of the line
  - If special case true:
    - Compromise and go *straight!*



# Unsolved Algorithm Issues

- Extremely sharp curves (< 90 deg.)
- Similar to branching issue (but more troublesome)
  - Current algorithm averages X values
  - Heads for center between two intersections of line with center row



# Problem: Follow the Line!

- ...that means don't follow the square.
- A possible solution: Shape Recognition
  - Feature Extraction
  - Describe regions



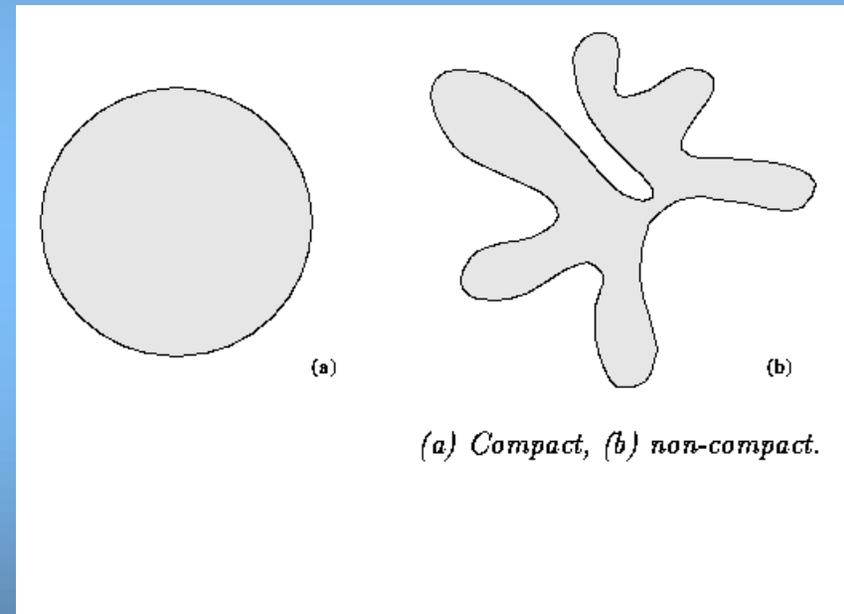
**FAILURE!**

# In the Future: Feature Extraction

- Features easy to measure:
  - Area
  - Perimeter
  - Granularity Measurement
- Hypothesis: AI/BO should choose
  - Largest
  - Most elongated
  - Least grainy region in image.
- And how can their measurements be combined → decision?

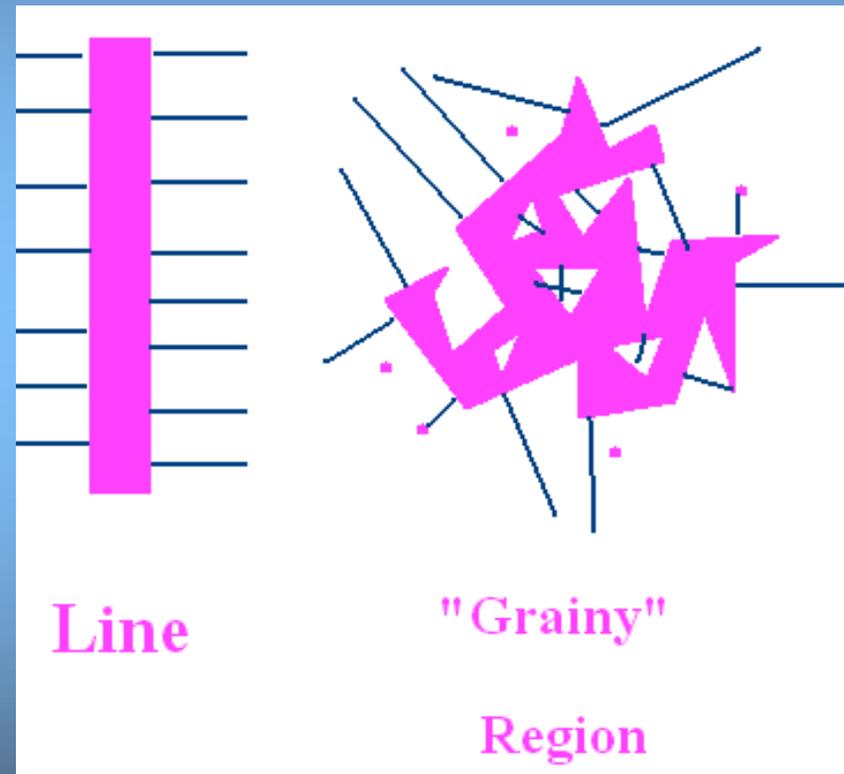
# Elongation

- Ratio of Perimeter to Area of region
- In computer vision, called “Compactness”
- $C = (P^2) / A$
- Circle = most compact
- Way to differentiate between more compact shapes and the line.



# Granularity Measurement

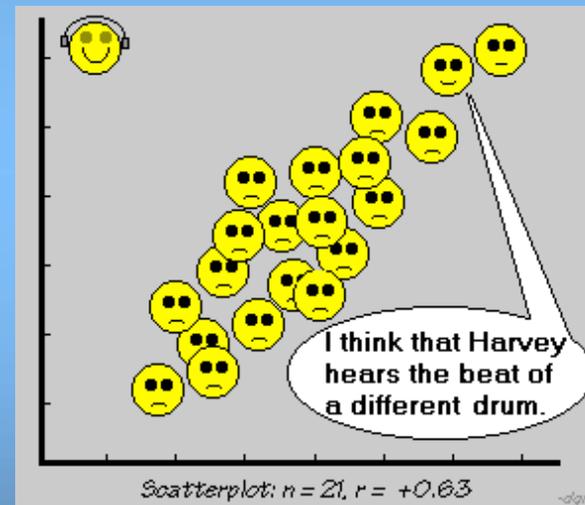
- Purpose: distinguish between
  - clear lines and grainy regions
- From each pixel at the boundary of a region
  - a ray extends until another pink pixel is encountered
- Granularity Measurement = sum of all rays' lengths



# Standardization

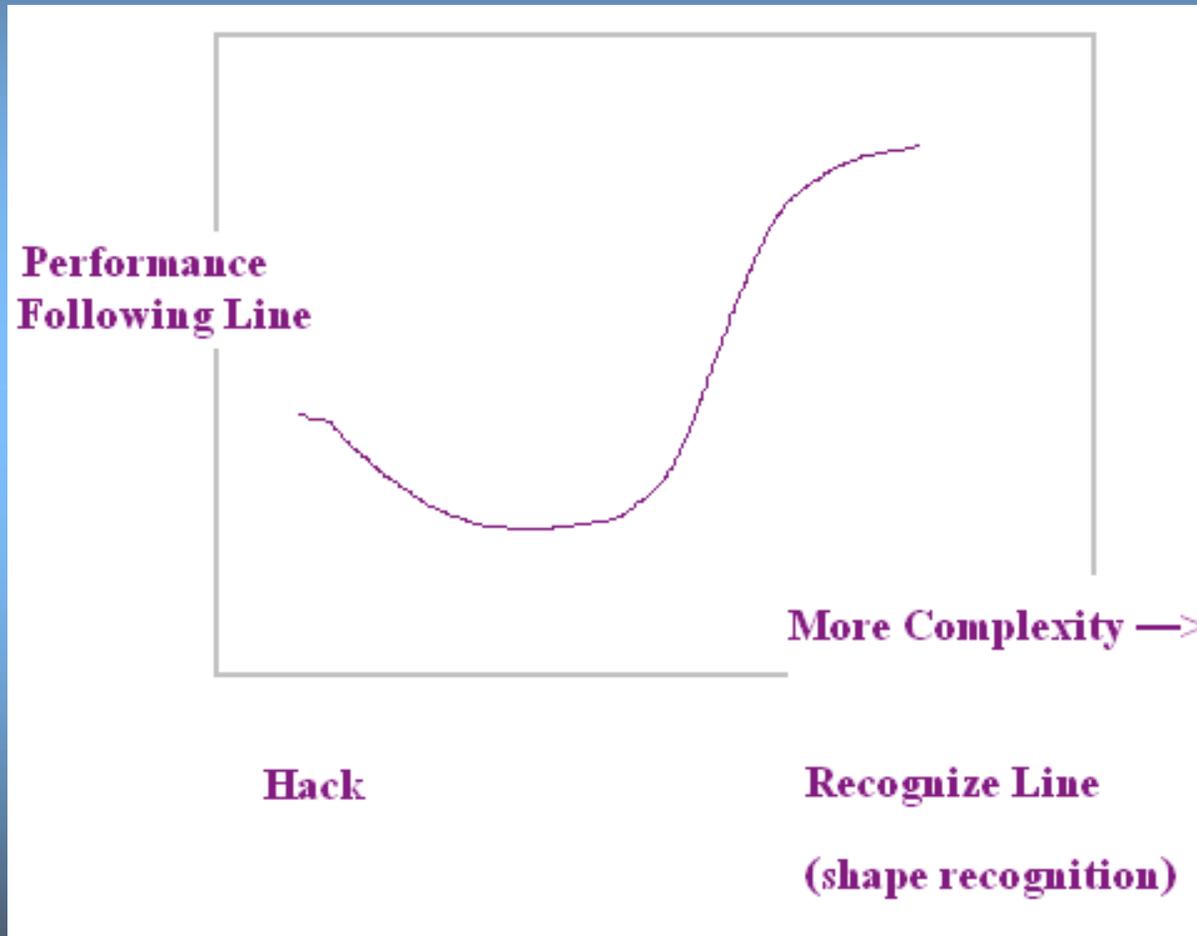
- To pick the best region, one can combine the three measurements
  - Largeness, Compactness, & Grainy Measure
- All have their own scales → standardization necessary
- Relevant → measurement for a region relative to all other regions.

• How different is this point?



[www.neiu.edu/~lruecker/smrm.htm](http://www.neiu.edu/~lruecker/smrm.htm)

# Performance vs. Complexity



# Live Demo

# How Different Walks Affect Line-Following

- Low velocity vs. high velocity
- Center of gravity affects slip potential
- Camera level affects:
  - Ease of getting lost
  - Precision of line following

# Special Thanks to:

- Ethan
- Alok Ladsariya
- Dr. David Touretzky
- Greg Kesden
- Sony
- CMU School of  
Computer Science  
and Robotics  
Institute
- Manuela Veloso &  
CMU RoboCup
- PMatt



# Bye!



# Image Sources

[www.dai.ed.ac.uk/HIPR2/hough.htm](http://www.dai.ed.ac.uk/HIPR2/hough.htm)

[www.dai.ed.ac.uk/CVonline/LOCAL\\_COPIES/MARBLE/medium/contours/feature.htm](http://www.dai.ed.ac.uk/CVonline/LOCAL_COPIES/MARBLE/medium/contours/feature.htm)

[www.physik.uni-osnabrueck.de/nonlinop/Hough/LineHough.html](http://www.physik.uni-osnabrueck.de/nonlinop/Hough/LineHough.html)

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[www003.upp.so\\_net.ne.jp/studio\\_mm/kameo/kameo.html](http://www003.upp.so_net.ne.jp/studio_mm/kameo/kameo.html)

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[www.seaple.icc.ne.jp/~somari/aibo.htm](http://www.seaple.icc.ne.jp/~somari/aibo.htm)

[www-2.cs.cmu.edu/afs/cs/project/robosoccer/www/legged/legged-team.html#publications](http://www-2.cs.cmu.edu/afs/cs/project/robosoccer/www/legged/legged-team.html#publications)

[ccrma-www.stanford.edu/CCRMA/Courses/252/sensors/node9.html](http://ccrma-www.stanford.edu/CCRMA/Courses/252/sensors/node9.html)

<http://www.icaen.uiowa.edu/~dip/LECTURE/Shape3.html#scalar>