ECED3901 Design Methods II

LECTURE #10: NAVIGATION #1

What are we covering?

- What is Navigation?
- Dead Reckoning
- Inertial Measurements
- Compasses
- Combining Measurements
- Ideas for Navigational Algorithms

What is Navigation?

Fundamental problem:

Where should we go?

Things to Consider

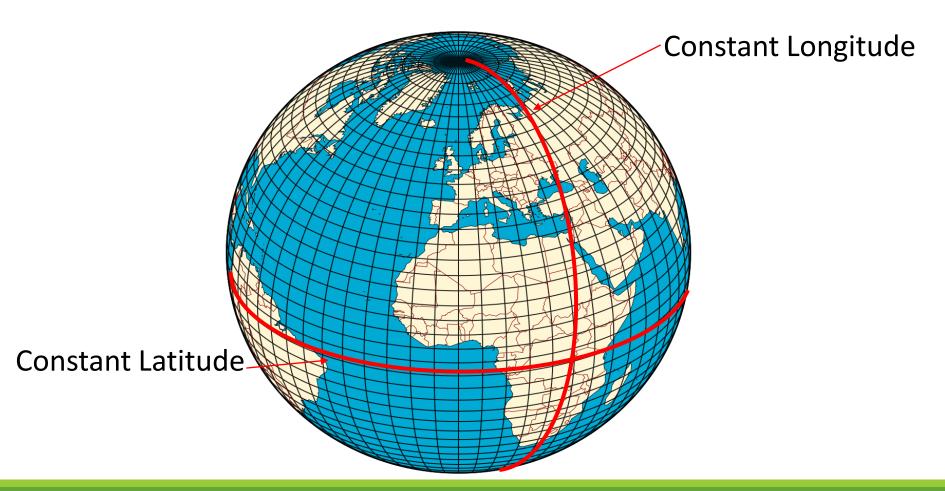
- 1. What is our objective?
 - Light-Seeking robot?
 - Finding an objective?
 - Exploring?
 - Mapping?
- 2. Where are we?
 - o Do we have a map?
 - Are we building a map?
- 3. Where can we go?
 - Limitations of our robot (turn radius, battery life, etc.)?
 - Known obstacles?

Interesting History of Navigation

Just a few Tidbits

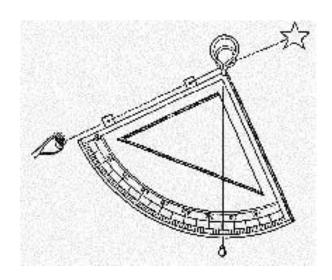
This is not a history class... but might give you some perspective on the problem

Navigating Oceans



Finding your (Rough) Location

Latitude → Possible using Stars or Sun



Quadrant



Mariner's Astrolabe (1645)

Finding your (Rough) Location

Longitude → Much more difficult

"Best" way is to have accurate time-keeper on ship... Earth rotates at fixed rate... if could determine location of reference objective relative to time then can solve problem.

Navigating by Air



Navigating by Air

Charles Lindbergh on an earlier flight:

Over the Straits of Florida my magnetic compass rotated without stopping.... I had no notion whether I was flying north, south, east, or west. A few stars directly overhead were dimly visible through haze, but they formed no constellation I could recognize. I started climbing toward the clear sky that had to exist somewhere above me. If I could see Polaris, that northern point of light, I could navigate by it with reasonable accuracy. But haze thickened as my altitude increased....

Nothing on my map of Florida corresponded with the earth's features I had seen...where could I be? I unfolded my hydrographic chart [a topographic map of water with coastlines, reefs, wrecks and other structures].... I had flown at almost a right angle to my proper heading and it...put me close to three hundred miles off route!

From: http://www.airspacemag.com/history-of-flight/even-lindbergh-got-lost-3381643

Colin O'Flynn

Dead Reckoning

Basic Principles

- Some known starting point
- Measurements are then added to give us our new position
 - Almost always involves integration

• i.e.: Distance travelled in a car = t * v

Core Problem with Dead Reckoning

As we are integrating (adding) up measurements, individual errors will compound over time

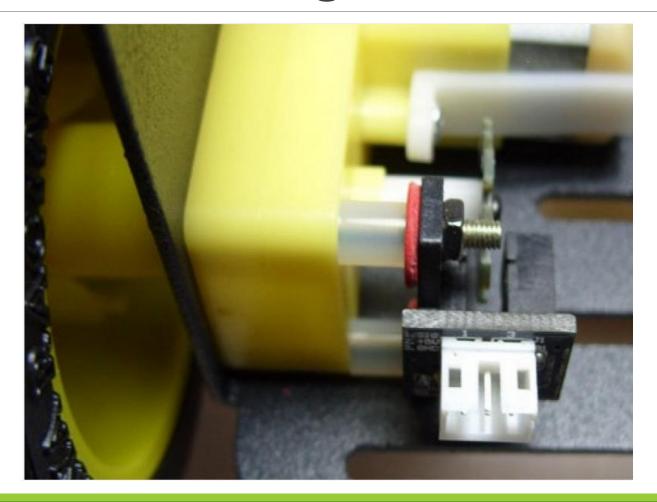
Historic Dead Reckoning...

6th Century or Earlier (depending on source)





Dead Reckoning in Your Robot



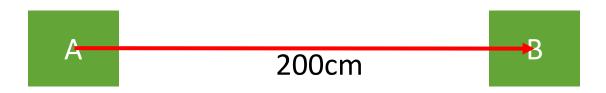
Counting Pulses

- Need to select resistor values for optical device
- Interface pulse output to AVR
- Count pulses to determine rotation of the robot wheels
- Need to calibrate possible errors out

Calibration Method



Calibration Method



Sources of Error

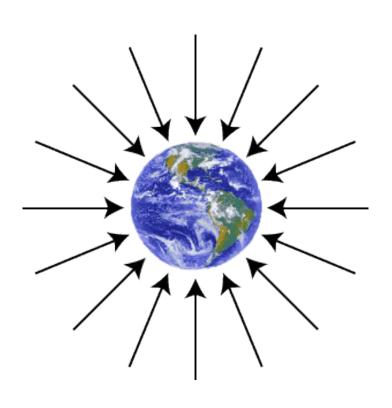
- Robot moving without wheels rotating (slipping)
- Bad calibration data
- Changing wheel diameter

Inertial Measurements

You're stuck in a van...



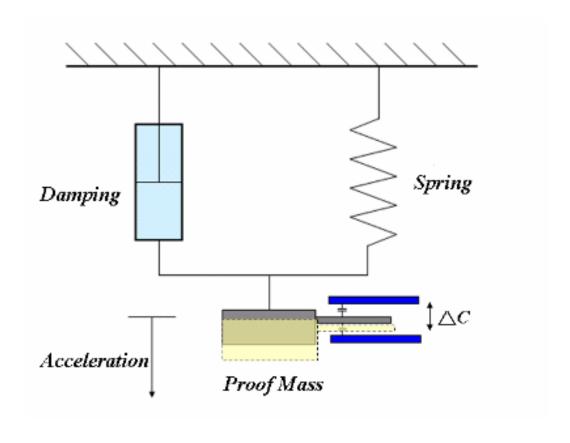
Measurement #1: Acceleration



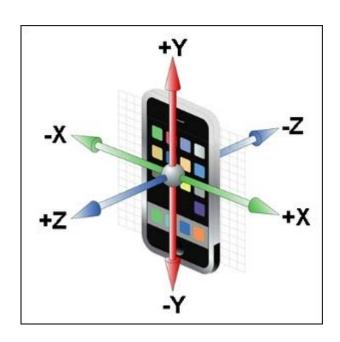
- Gravity Vector points downwards (9.8 m/s²) always pointing downward
- Acceleration due to movement as well



Basic Principle



3-Axis Accelerometers



Mapping of Gravity Vector

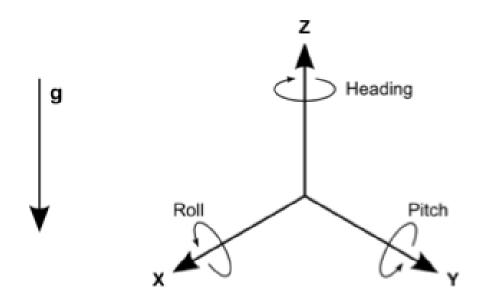
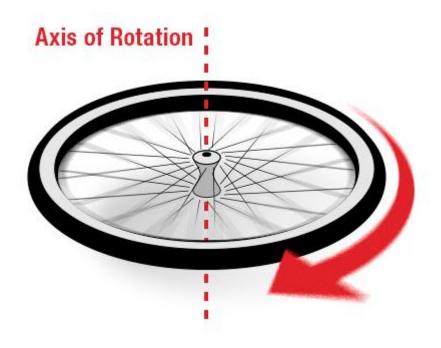


Figure 1: Gravity Vector and Heading, Pitch, & Roll about Axes

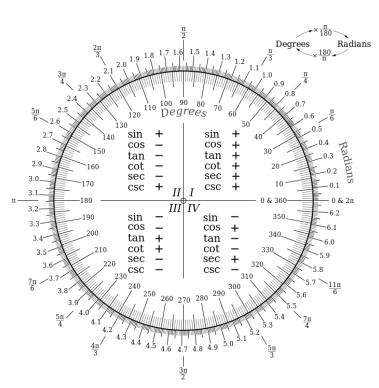
What can it NOT detect

Double-Integration & Errors

Measurement #2: Angular Rate



Rotational Units



Typically specified in deg/s or rad/s

http://en.wikipedia.org/wiki/Radian#/media/File:Degree-Radian_Conversion.svg

Sidenote: Gimballed



http://upload.wikimedia.org/wikipedia/commons/d/d5/Gyroscope_operation.gif

Early IMUs



http://en.wikipedia.org/wiki/Inertial_measurement_unit#/media/File:Centrale-intertielle_missile_S3_Musee_du_Bourget_P1010652.JPG

Peacekeeper Missle (ICBM)





http://en.wikipedia.org/wiki/LGM-118_Peacekeeper

Colin O'Flynn

Strap-down IMUs





VS.



>\$30 000 USD Difficult to purchase.

Probably going to jail if I travel with it.

>\$10 USD Widely available Not going to jail.

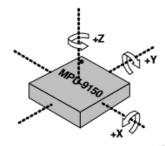
Performance Differences

| OVER ASSET HOUSE PERFORMANCE | E0 0E1 0 | | | |
|----------------------------------|------------------------|-----|-----|--|
| ZRO Variation Over Temperature | -40°C to +85°C | ±20 | º/s | |
| Initial ZRO Tolerance | Component level (25°C) | ±20 | º/s | |
| GYROSCOPE ZERO-RATE OUTPUT (ZRO) | | | | |
| | I. | | | |

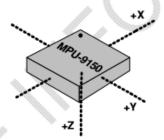
LN-200 Core IMU

| Performance | | | | |
|-----------------------|---|--|--|--|
| Acc | Accelerometer | | | |
| Bias Repeatability | 300 μg to 3.0 milli-g, 1σ | | | |
| Scale Factor Accuracy | 300 to 5,000 ppm, 1 _o | | | |
| Gyro | | | | |
| Bias Repeatability | 1º/hr to 3º/hr, 1σ | | | |
| Scale Factor Accuracy | 100 to 500 ppm, 1σ | | | |
| Random Walk | 0.07° to 0.15°/√hr Power Spectral Density (PSD) level | | | |

Using your MPU 9150



Orientation of Axes of Sensitivity and Polarity of Rotation for Gyroscopes and Accelerometers



Orientation of Axes of Sensitivity for Compass

Example of Sensor Readings

Calculating Pitch & Roll

```
if ((Ysqacc + Zsqacc) == 0) {
          pitch = M_PI/2;
} else {
          pitch = atan(Xacc / (sqrt(Ysqacc + Zsqacc)));
}

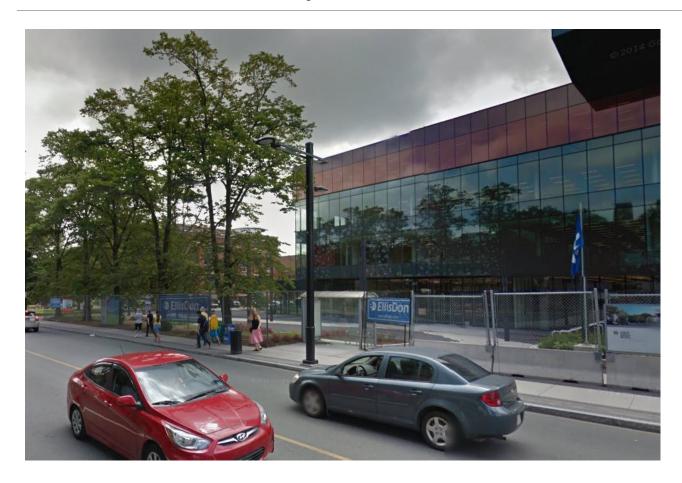
if ((Xsqacc + Zsqacc) == 0) {
          roll = M_PI/2;
} else {
          roll = atan(Yacc / (sqrt(Xsqacc + Zsqacc)));
}
```

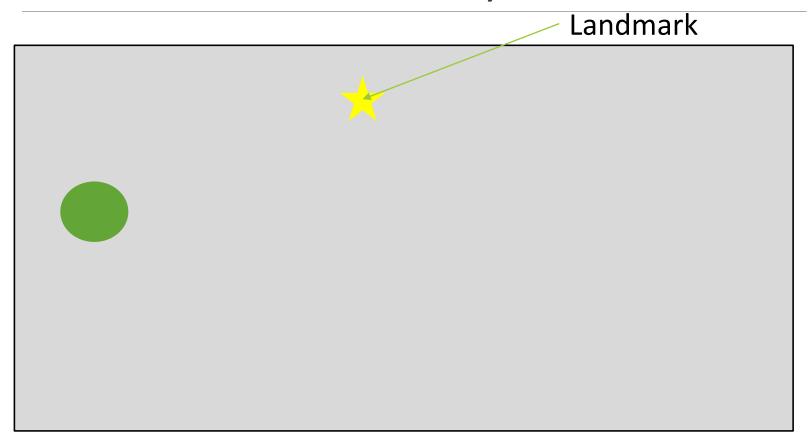
Magnetic Sensor to Compass

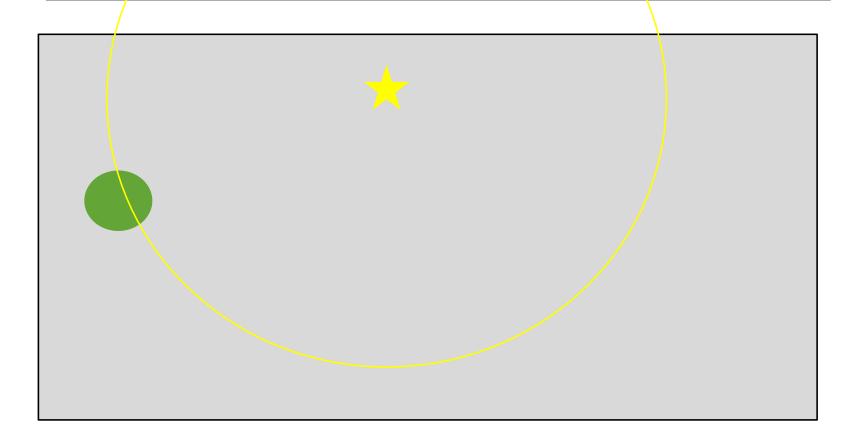
Tilt Compensation

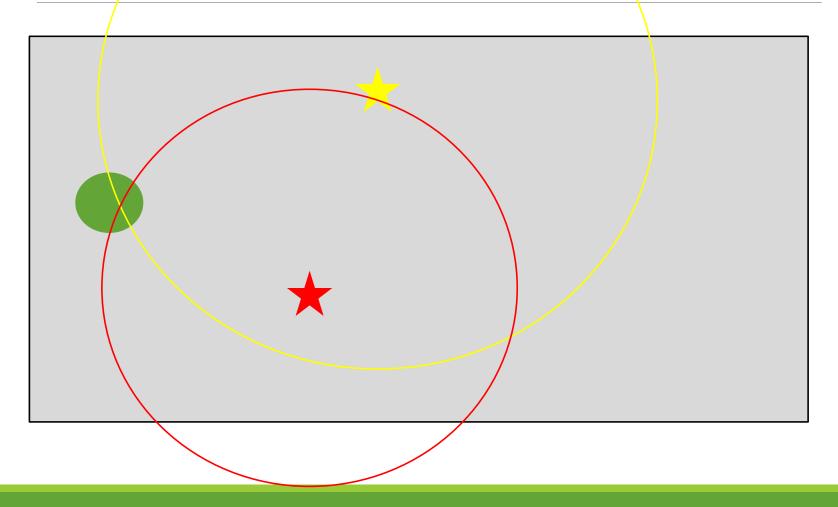
Landmarks

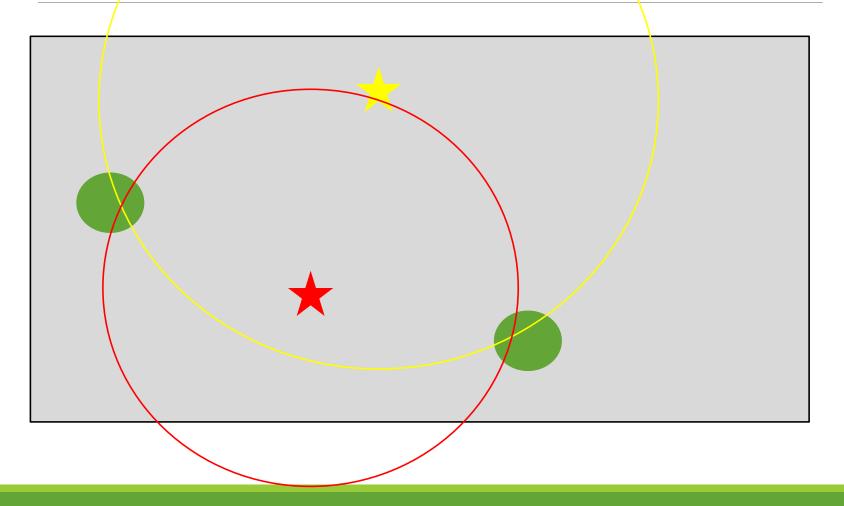
Where are you?

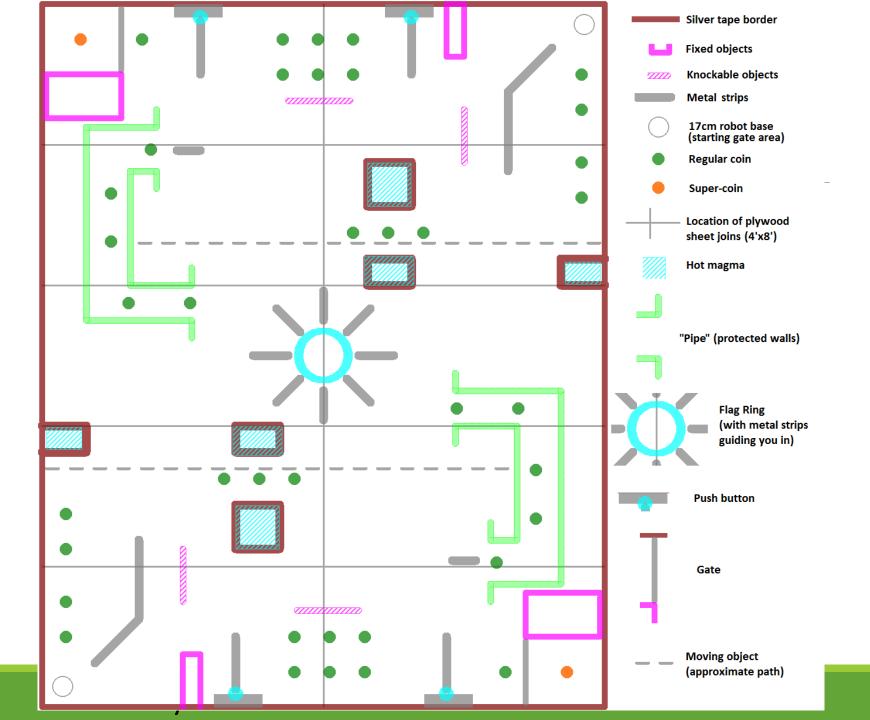












Combining Measurements

Measurement Example on Car

GPS:

Inertial:

Speedometer:

Kalman Filtering

Has idea of estimated state... state being speed, position, etc.

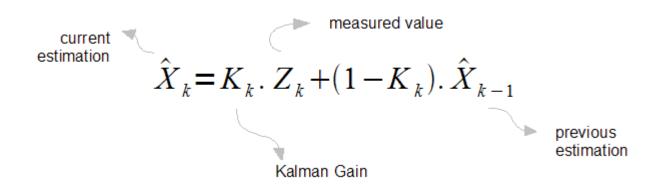
Uses two steps:

- 1. A prediction step which predicts new state
- 2. A measurement step which uses noisy measurements

http://bilgin.esme.org/BitsBytes/KalmanFilterforDummies.aspx

http://blog.tkjelectronics.dk/2012/09/a-practical-approach-to-kalman-filter-and-how-to-implement-it/

Kalman Filtering



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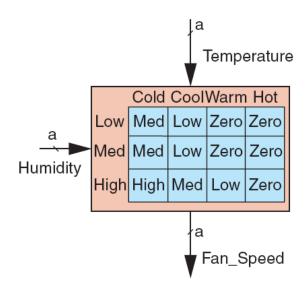
Colin O'Flynn



Fuzzy Logic

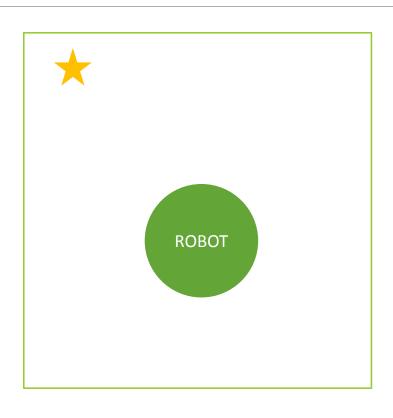
Not this type of fuzzy.

Fuzzy Controller



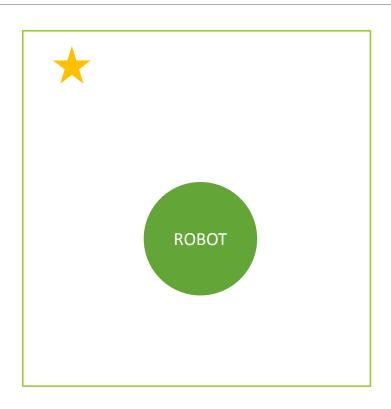
Simple Light-Seeking Robot

```
left = read_light(LEFT);
right = read_light(RIGHT);
if (left < right){</pre>
    motorR(ON);
    motorL(OFF);
}
if (right > left){
    motorL(OFF);
    motorR(ON);
}
```



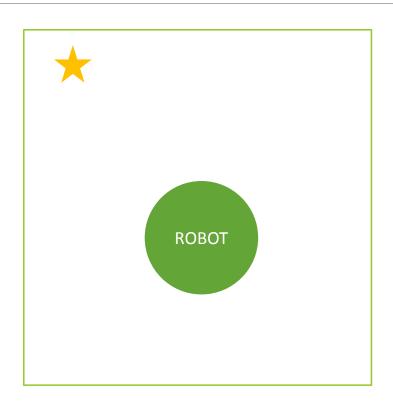
A Quazi-Fuzzy Solution...

```
left = read_light(LEFT);
right = read_light(RIGHT);
motorL(right * ARB_CONST1);
motorR(left * ARB_CONST1);
```



A Quazi-Fuzzy Solution...

```
left = read_light(LEFT);
right = read_light(RIGHT);
motorL(right * ARB_CONST1);
motorR(left * ARB_CONST1);
```



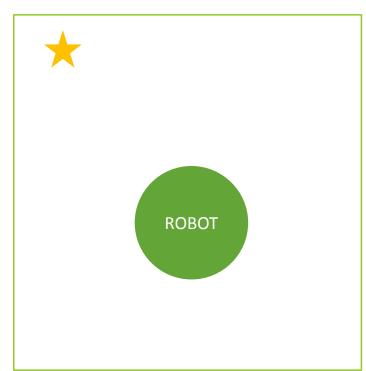
http://www.societyofrobots.com/programming fuzzy logic.shtml

Adding a Touch Sensor

```
left = read_light(LEFT);
right = read_light(RIGHT);

motorL(left_bumper()*-100 +
right * ARB_CONST1);

motorR(right_bumper()*-100 +
left * ARB_CONST1);
```



http://www.societyofrobots.com/programming fuzzy logic.shtml

What is Fuzzy Logic?

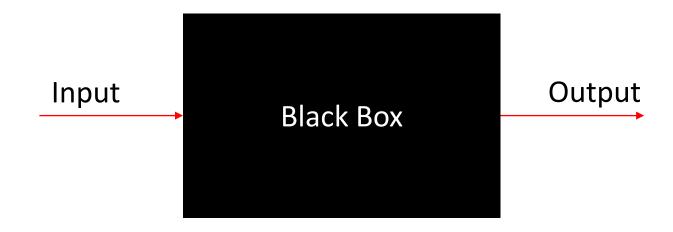
NOTE: Previous example is *not* a good example classic fuzzy logic, and in fact could be argued is instead a form of analog computer...

Interested in playing around more with fuzzy sets? I'd suggest starting with using

http://jfuzzylogic.sourceforge.net/html/index.html

Neural Networks

Pattern Recognition

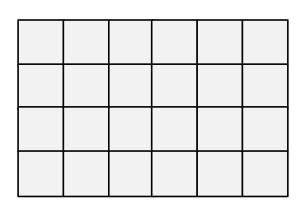


Potential Problems

- 1) They are *not* magic. You cannot use them for lottery number prediction for example.
 - Also difficult if not clear there *is* a pattern (i.e. stock number prediction)

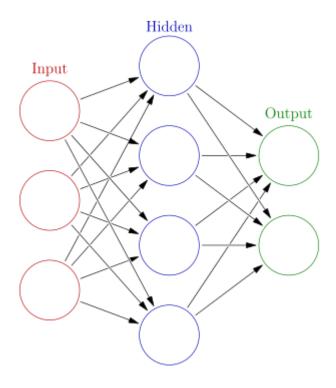
- 2) Requires diverse training set.
 - For problems which you already have a good model, they will probably perform worse.
- 3) May learn the wrong pattern if your training set has errors.

Using a Neural Network



Neural Network

Network Functions

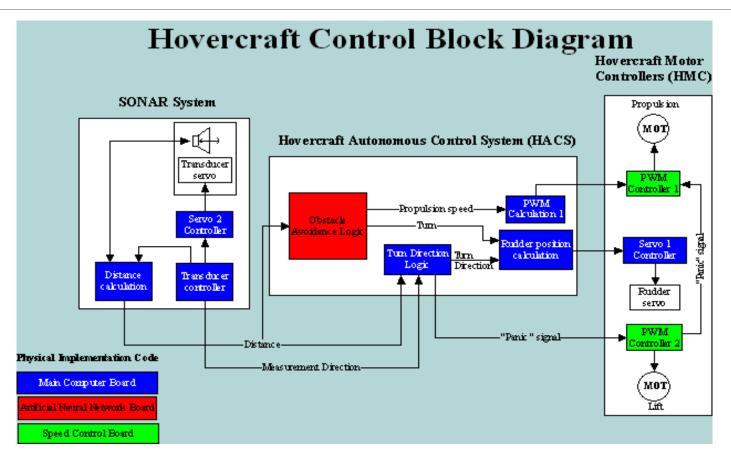


Neural Network Example

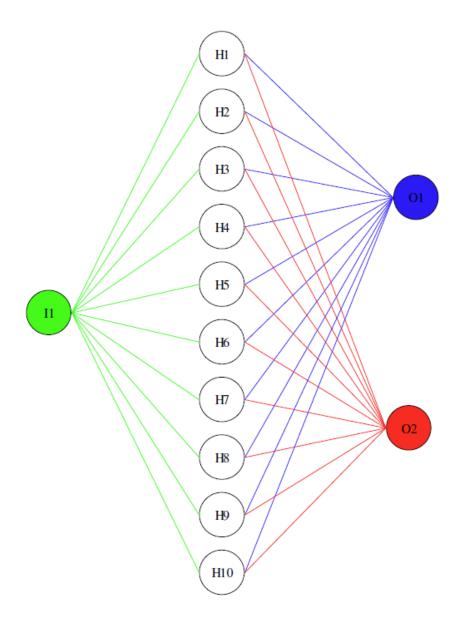


http://colinoflynn.com/oldsite/tiki-index.php?page=ProjectHovercraft

Simple Block Diagram



ANN Example

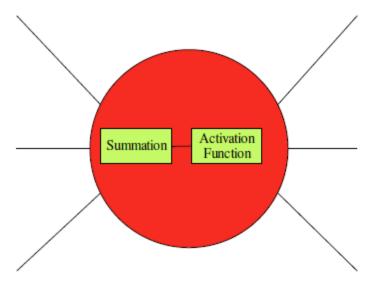


ANN Weights

| Neuron | Neuron | Weight |
|--------|--------|-----------|
| I1 | H1 | -0.704894 |
| I1 | H2 | -6.526604 |
| I1 | H3 | 0.568346 |
| I1 | H4 | 4.448008 |
| I1 | H5 | 2.076094 |
| I1 | H6 | 2.238839 |
| I1 | H7 | 2.374001 |
| I1 | H8 | 7.698660 |
| I1 | H9 | 2.429901 |
| I1 | H10 | 2.426541 |
| H1 | O1 | 2.511209 |
| H1 | O2 | -2.440218 |
| H2 | O1 | 13.311865 |
| H2 | O2 | 13.178253 |
| H3 | O1 | 3.122861 |

| Neuron | Neuron | Weight |
|--------|--------|------------|
| H3 | O2 | -1.016410 |
| H4 | O1 | 1.125672 |
| H4 | O2 | -2.862084 |
| H5 | O1 | -3.051573 |
| H5 | O2 | 1.669890 |
| H6 | O1 | -3.647110 |
| H6 | O2 | 3.076672 |
| H7 | O1 | -2.520317 |
| H7 | O2 | 4.681576 |
| H8 | O1 | 3.402647 |
| H8 | O2 | -11.532937 |
| H9 | O1 | -2.229351 |
| H9 | O2 | 5.593643 |
| H10 | O1 | -1.643753 |
| H10 | O2 | 5.045427 |

Neuron Structure

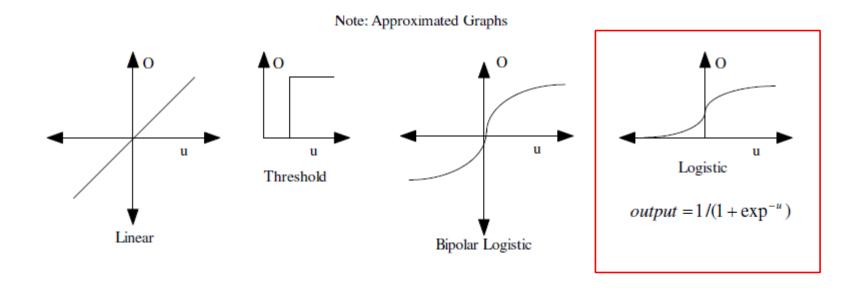


The inputs to the neuron are first passed through a summation function. The summation function is simply:

$$u = (o_1 \bullet w_1) + (o_2 \bullet w_2) + (o_n \bullet w_n)$$

Where O is the output from the connected neuron that is presented as the input to this neuron, and w is the weight of that connection. Once all the weights have been calculated, the output of that is passed into the activation function. The activation function is the key to the neural network, and there are several different types.

Activation Functions



Training Example

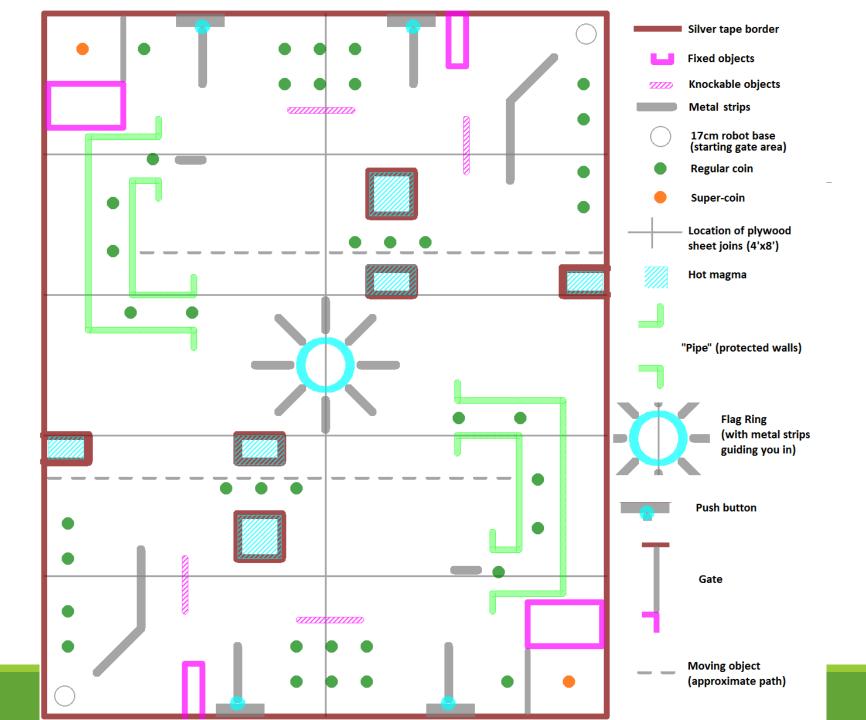
```
7 1 2
.1 1 1
.2 .8 .9
.3 .4 .7
.4 .2 .6
.5 .1 .6
.7 0.05 .8
1 0 .9
```

<distance> <speed> <turn>

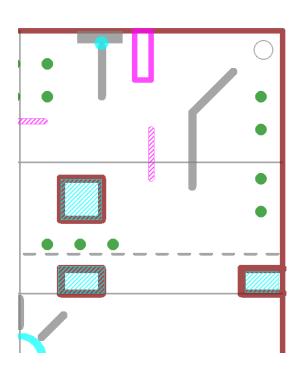
Notes:

- Training algorithm adjusts weights such that expected response occurs for the known inputs
- •This project just has turn as a magnitude separate logic determines direction (left/right) to turn

Robot Navigational Techniques

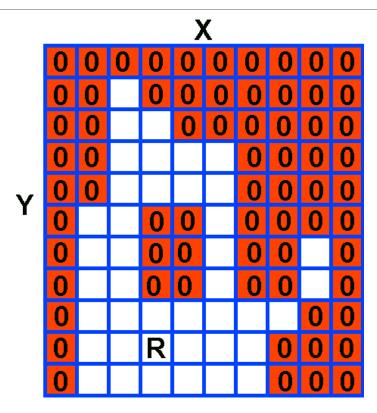


Map Representation



| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|---|---|---|---|---|---|---|---|---|---|---|---|---|
| | | | | | | 0 | 0 | | | | | 0 |
| | | | | | | 0 | 0 | | | | | 0 |
| | | | | | | | | | | | | 0 |
| | | | | | | | | | | | | 0 |
| | | | | | | | 0 | | | | | 0 |
| | | | 0 | 0 | 0 | | 0 | | | | | 0 |
| | | | 0 | 0 | 0 | | 0 | | | | | 0 |
| | | | 0 | 0 | 0 | | | | | | | 0 |
| | | | | | | | | | | | | 0 |
| | | | | | | | | | | | | 0 |
| | 0 | 0 | 0 | | | | | | | 0 | 0 | 0 |
| | 0 | 0 | 0 | | | | | | | 0 | 0 | 0 |
| | | | | | | | | | | | | 0 |
| | | | | | | | | | | | | 0 |
| | | | | | | | | | | | | 0 |

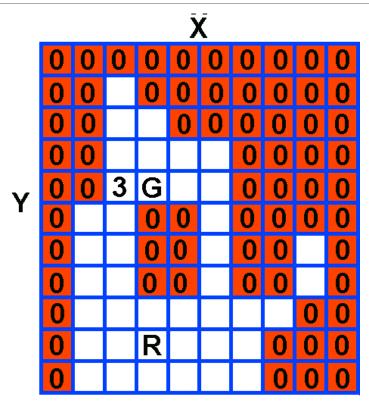
map[x][y]

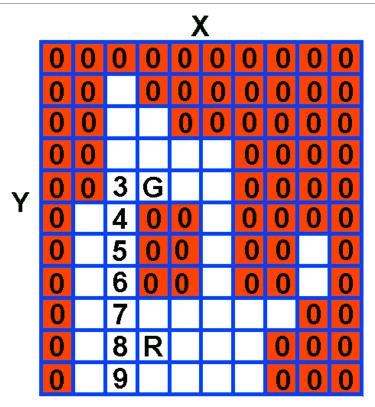


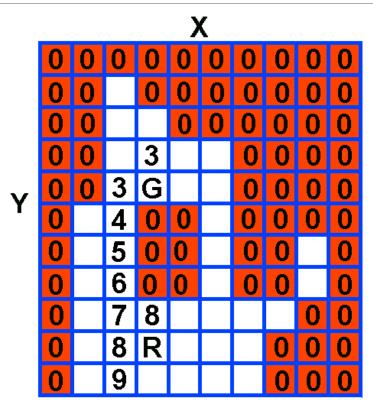
Wavefront Code

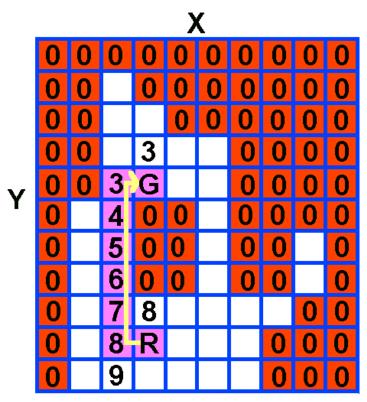
Start at node x=0 y=0:

- 1. If boundry node is wall \rightarrow Ignore, goto 5
- 2. If boundary node is robot location && has number \rightarrow path found
- 3. If boundary node has a goal \rightarrow Mark node with 3, goto 5
- If boundary node is marked with number → mark node with number+1, goto 5
- 5. If no path found \rightarrow node = y+1
- 6. If no path found after full scan of matrix \rightarrow node = (x=0, y=0) but do not clear matrix
- 7. If no path found and matrix full \rightarrow fail.









Conclusions

- Many options for navigation
- Don't get overloaded lots of options and things to consider, but making the most robust solution is the best option!
- Use of magnetic compasses *might* be useful... needs experimentation on your surface.
- Use of accelerometers might give you bump detection.