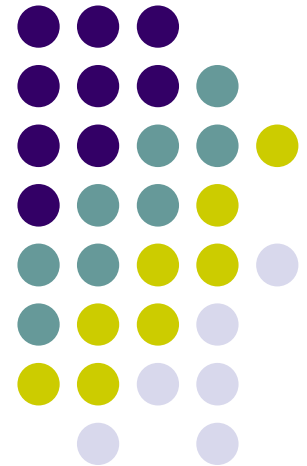


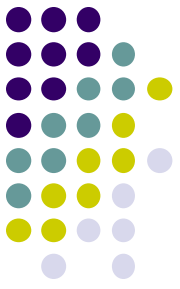
CS 4518 Mobile and Ubiquitous Computing

Lecture 13: Machine Learning for Ubiquitous Computing

Emmanuel Agu



Reminder: 1 Slide of Final Project



- 1-slide from group today, extended till tomorrow Tuesday (2/7):
 - 2/40 of final project grade
- Propose mobile/ubiquitous computing app, solves WPI problem
- Slide should contain 3 bullets
 - 1. Problem you intend to work on**
 - Solve WPI/societal problem (e.g. walking safe at night)
 - Use at least location, 1 sensor or camera
 - If games, must gamify solution to real world problem
 - 2. Why this problem is important**
 - E.g. 37% of WPI students feel unsafe walking home
 - 3. Summary of envisioned mobile app (?) solution**
 1. E.g. Mobile app automatically texts users friends when they get home at night
- Can bounce ideas of me (email, or in person)
- Can change idea any time

Rubric: Grading Considerations



- **Problem (30/100)**

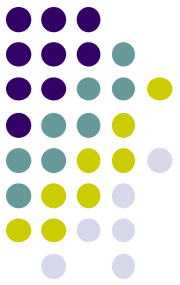
- How much is the problem a real problem (e.g. not contrived)
- Is this really a good problem that is a good fit to solve with mobile/ubiquitous computing? (e.g. are there better approaches?)

- **Importance (30/100)**

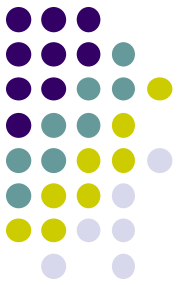
- How useful would it be if this problem is solved?
- What is the potential impact on the community (e.g. WPI students) (e.g. how much money? Time? Productivity.. Would be saved?)
- What is the evidence of the importance? (E.g. quote a statistic)

- **Proposed Solution (40/100)**

- How good/clever is the solution?
- How sophisticated and how many are the mobile/ubiquitous computing components (high level) proposed? (e.g. location, geofencing, activity recognition, face recognition, machine learning, etc)



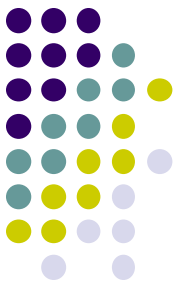
Intuitive Introduction to Machine Learning for Ubiquitous Computing



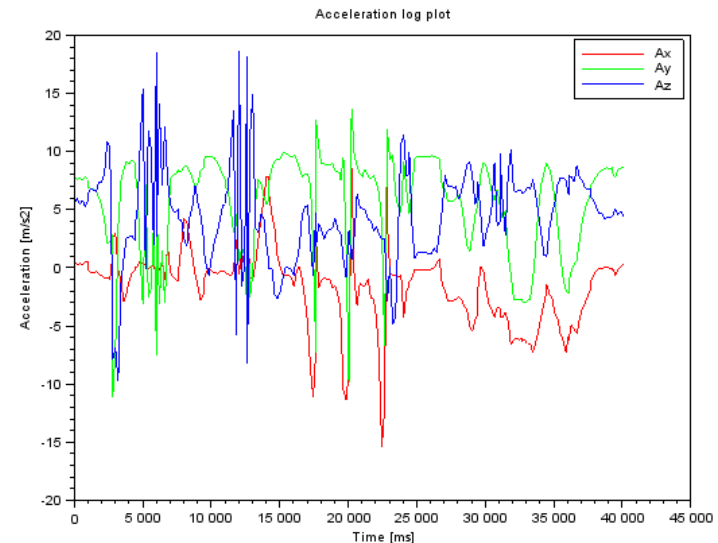
My Goals in this Section

- If you know machine learning
 - Set off light bulb
 - Projects involving ML?
- If you don't know machine learning
 - Get general idea, how it's used
- Knowledge will also make papers easier to read/understand

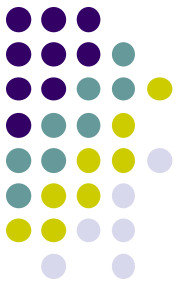
Recall: Activity Recognition



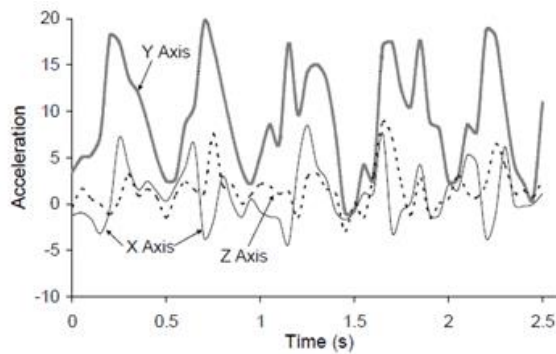
- Want app to detect when user is performing any of the following 6 activities
 - Walking,
 - Jogging,
 - Ascending stairs,
 - Descending stairs,
 - Sitting,
 - Standing



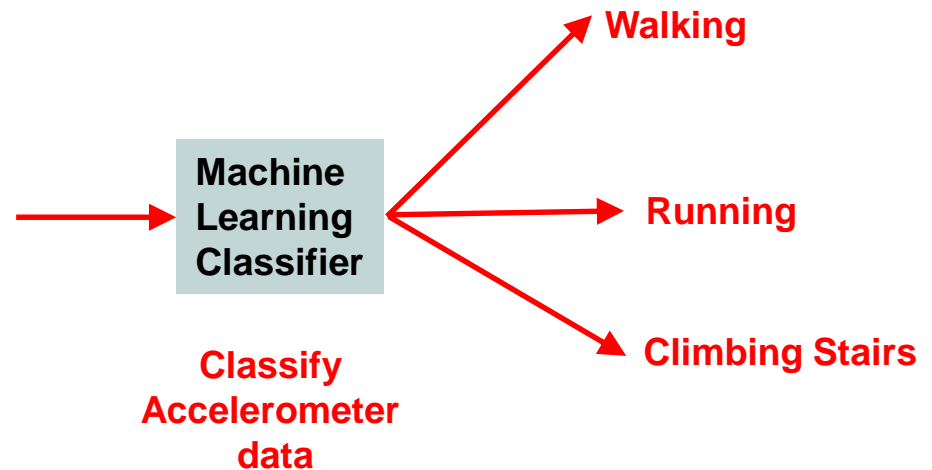
Recall: Activity Recognition Overview



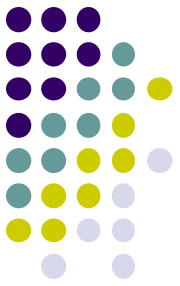
Gather Accelerometer data



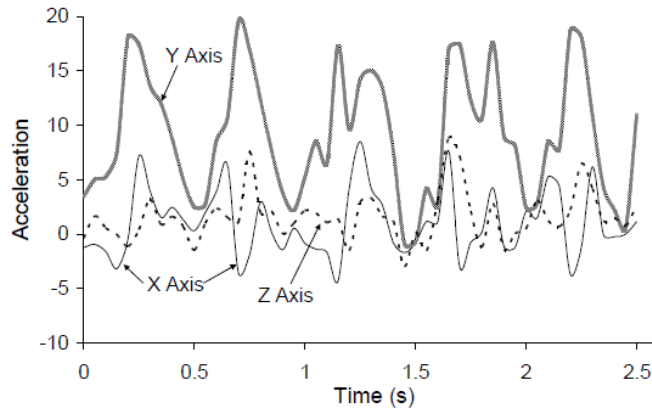
(a) Walking



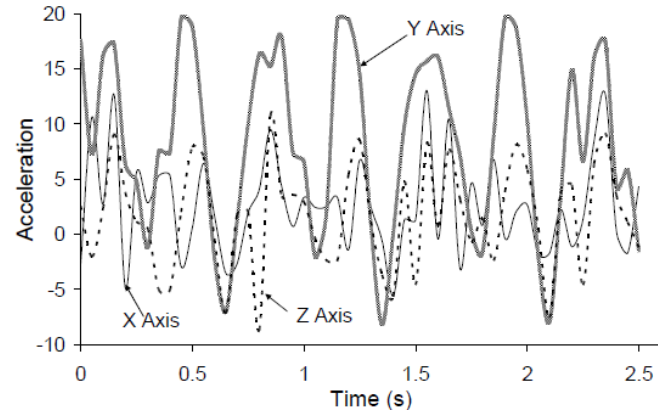
Recall: Example Accelerometer Data for Activities



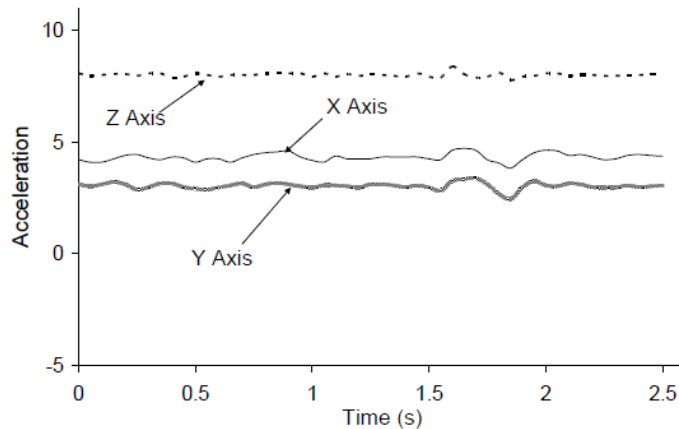
Different user activities generate different accelerometer patterns



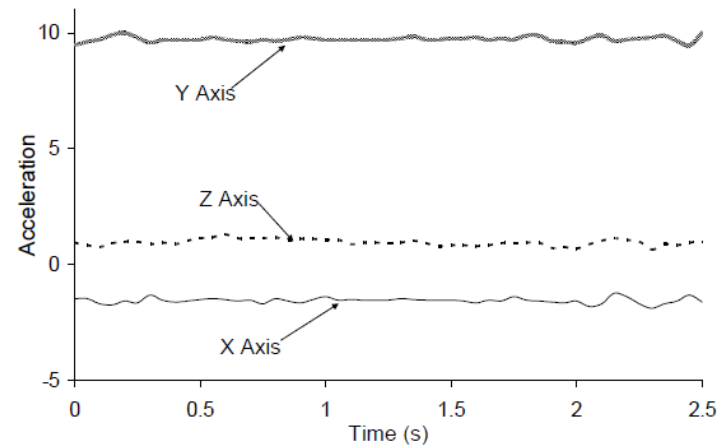
(a) Walking



(b) Jogging

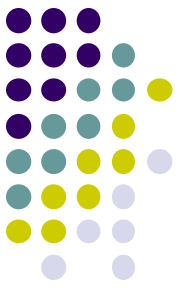


(e) Sitting

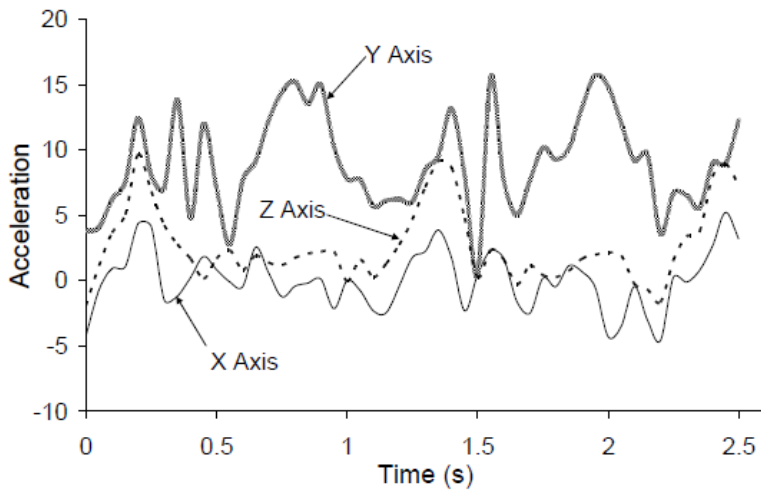


(f) Standing

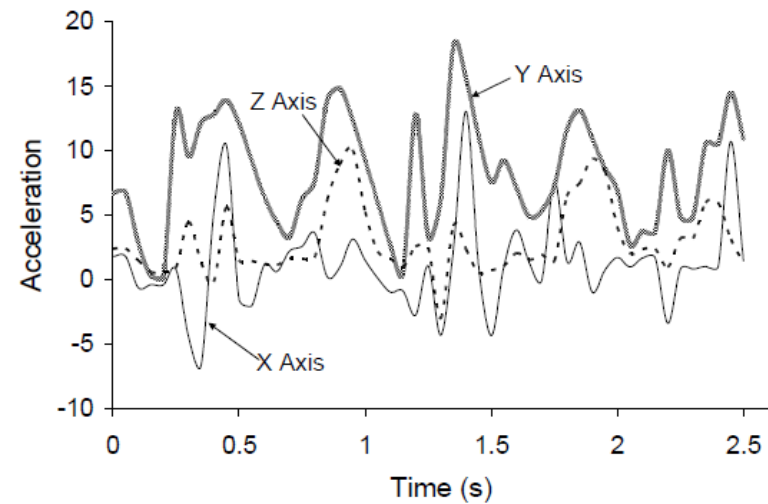
Recall: Example Accelerometer Data for Activities



Different user activities generate different accelerometer patterns

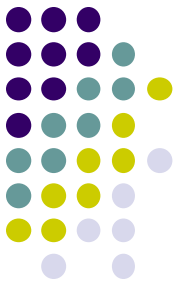


(c) Ascending Stairs

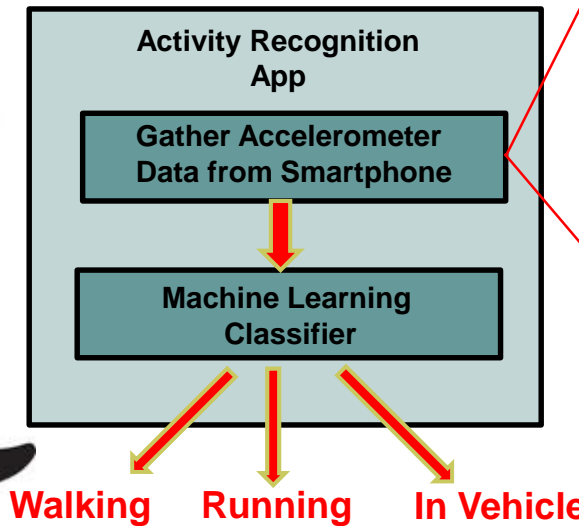
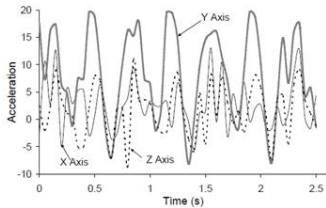


(d) Descending Stairs

DIY Activity Recognition (AR) Android App

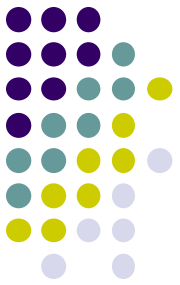


- As user performs an activity, AR app on user's smartphone
 1. Gathers accelerometer data
 2. Uses **machine learning classifier** to determine what activity (running, jumping, etc) accelerometer pattern corresponds to
- **Classifier:** Machine learning algorithm that guesses what activity **class** accelerometer sample corresponds to



```
msensor = (mSensorManager)
            getSystemService(Context.SENSOR_SERVICE)
....
Public void onSensorChanged(SensorEvent event){
....
}
```

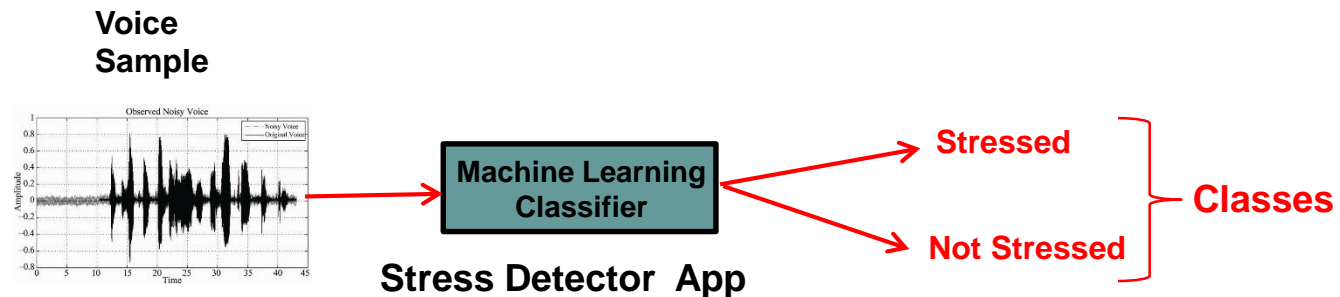
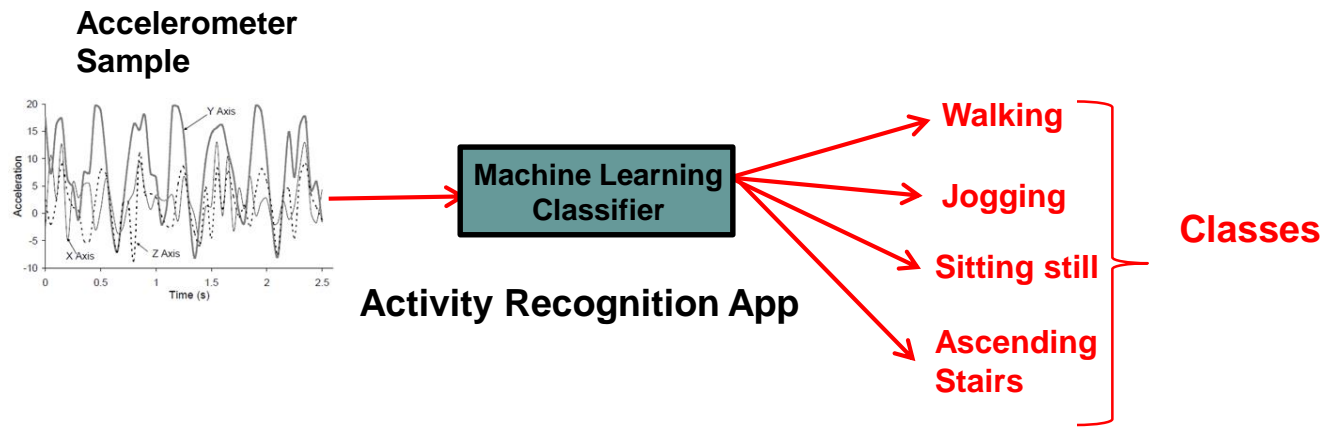
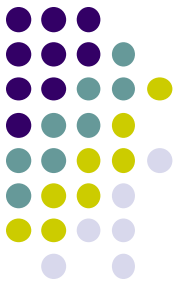
Next: Machine learning Classification



Classification for Ubiquitous Computing

Classification

- **Classification** is type of machine learning used a lot in UbiComp
- Classification? determine which **class** a sample belongs to. Examples:



Classification

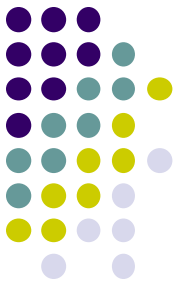
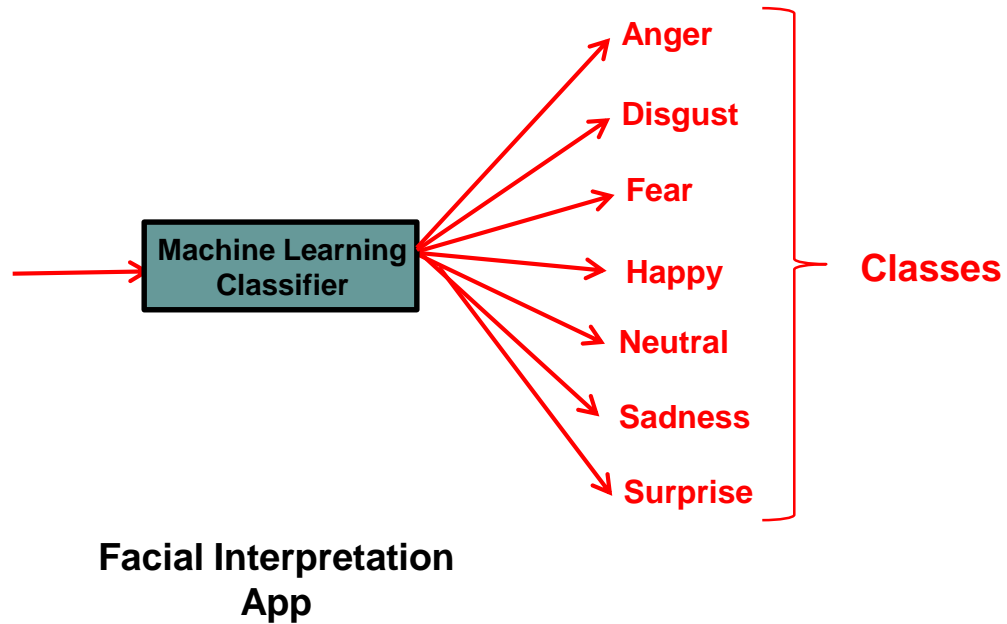
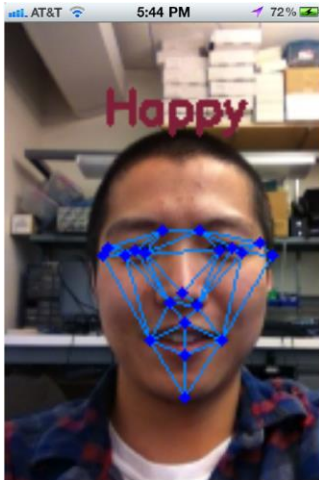
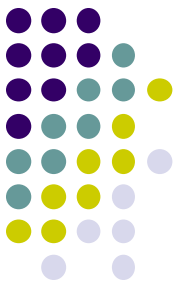


Image showing
Facial Expression

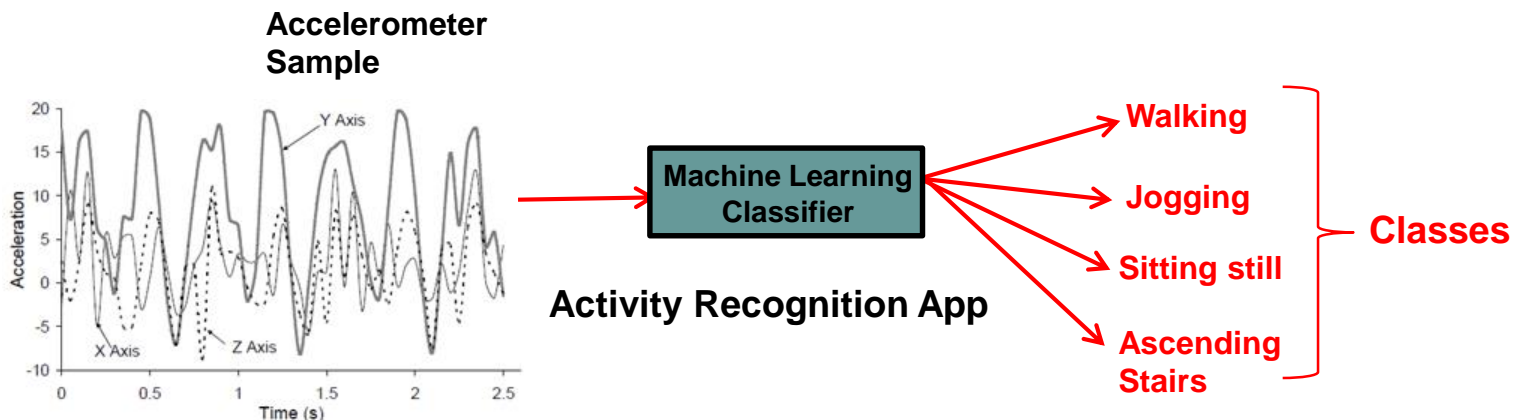


Classifier

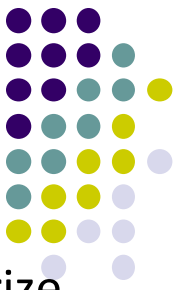


- Analyzes new sample, guesses corresponding class
- Intuitively, can think of classifier as set of rules for classification. E.g.
- Example rules for classifying accelerometer signal in Activity Recognition

```
If ((Accelerometer peak value > 12 m/s)
    and (Accelerometer average value < 6 m/s)){
    Activity = "Jogging";
}
```

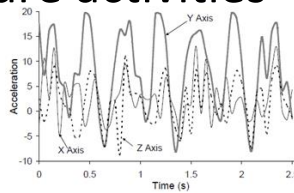


Training a Classifier

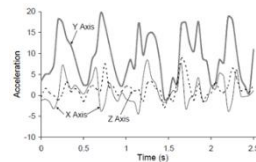


- Created using example-based approach (called training)
- **Training a classifier:** Examples of each class => generate rules to categorize new samples
- **E.g:** Analyze 30+ Examples (from 30 subjects) of accelerometer signal for each activity type (walking, jogging, sitting, ascending stairs) => generate rules (classifier) to classify future activities

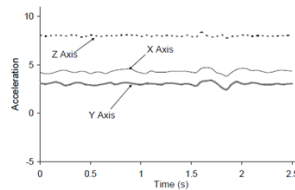
Examples of user jogging



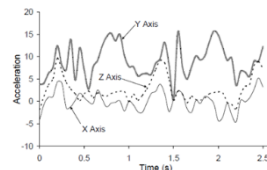
Examples of user walking



Examples of user sitting

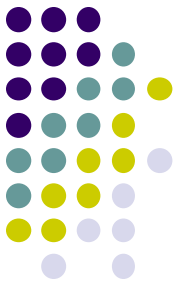


Examples of user ascending stairs

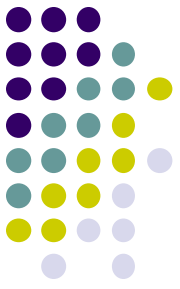


Train Machine Learning Classifier

Activity Recognition Classifier



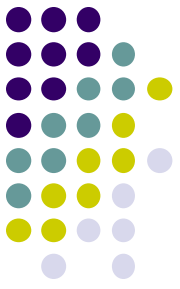
Training a Classifier: Steps



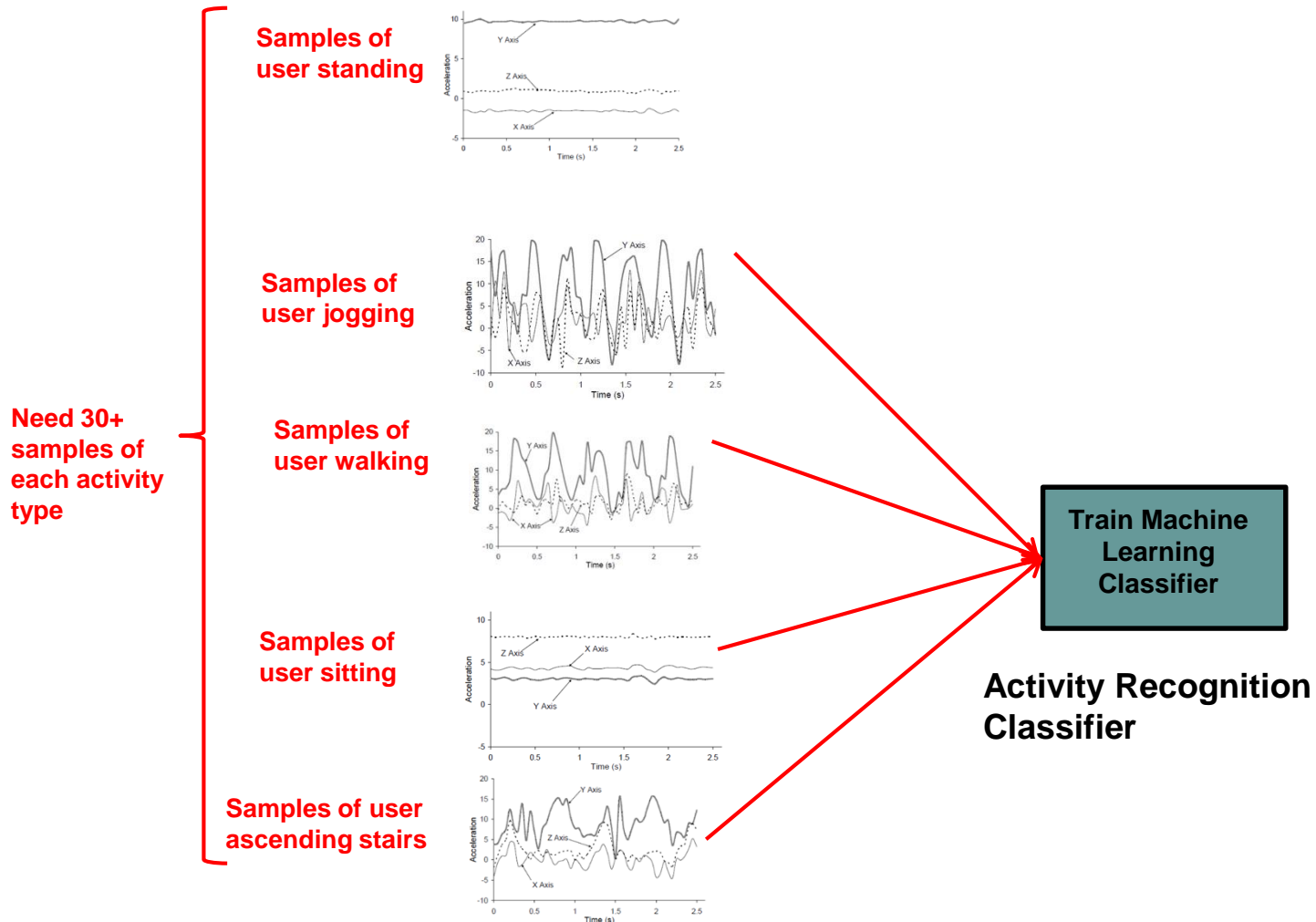
Steps for Training a Classifier

1. Gather data samples + label them
2. Import accelerometer samples into classification library (e.g. Weka, MATLAB)
3. Pre-processing (segmentation, smoothing, etc)
4. Extract features
5. Train classifier
6. Export classification model as JAR file
7. Import into Android app

Step 1: Gather Sample data + Label them



- Need many samples of accelerometer data corresponding to each activity type (jogging, walking, sitting, ascending stairs, etc)



Step 1: Gather Sample data + Label them

- Run a study to gather sample accelerometer data for each activity class
 - Recruit 30+ subjects
 - Run program that gathers accelerometer sensor data on subject's phone
 - Make subjects perform each activity (walking, jogging, sitting, etc)
 - Collect accelerometer data while they perform each activity (walking, jogging, sitting, etc)
 - Label data. i.e. tag each accelerometer sample with the corresponding activity
- Now have 30 examples of each activity

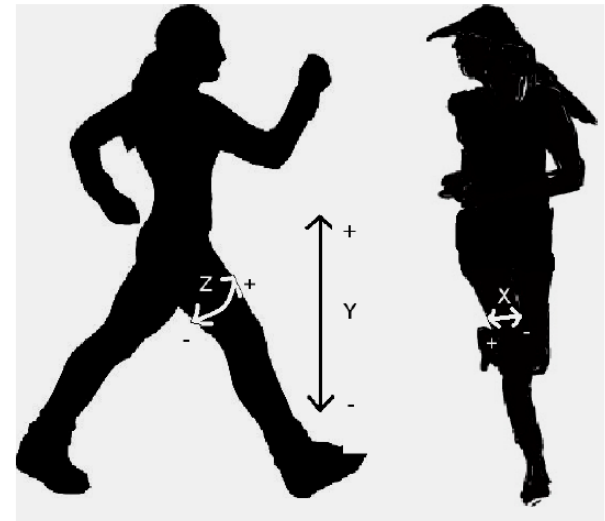
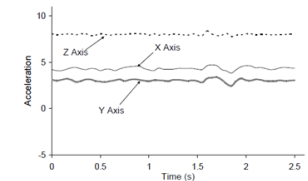
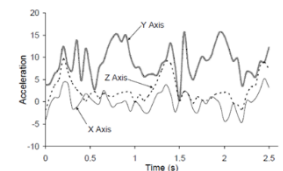


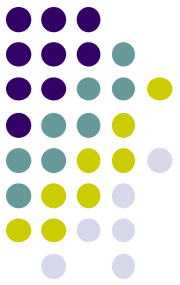
Figure 1: Axes of Motion Relative to User

**30+
Samples of
user sitting**



**30+ Samples of
user ascending
stairs**





Step 1: Gather Sample data + Label them

Program to Gather Accelerometer Data

- **Option 1:** Can write sensor program app that gathers accelerometer data while user is doing each of 6 activities (1 at a time)

```
mSensor = (mSensorManager)
    getSystemService(Context.SENSOR_SERVICE)
....

Public void onSensorChanged(SensorEvent event){
....
}
```



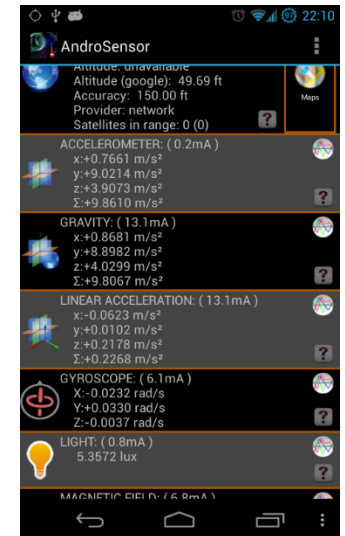
Step 1: Gather Sample data + Label them

Program to Gather Accelerometer Data

- **Option 2:** Use 3rd party app to gather accelerometer
 - 2 popular ones: **Funf** and **AndroSensor**
 - Just download app,
 - Select sensors to log (e.g. accelerometer)
 - Continuously gathers sensor data in background
- **FUNF** app from MIT
 - Accelerometer readings
 - Phone calls
 - SMS messages, etc
- **AndroSensor**

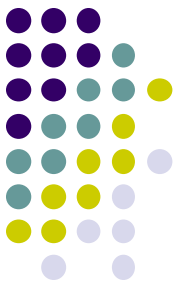


Funf

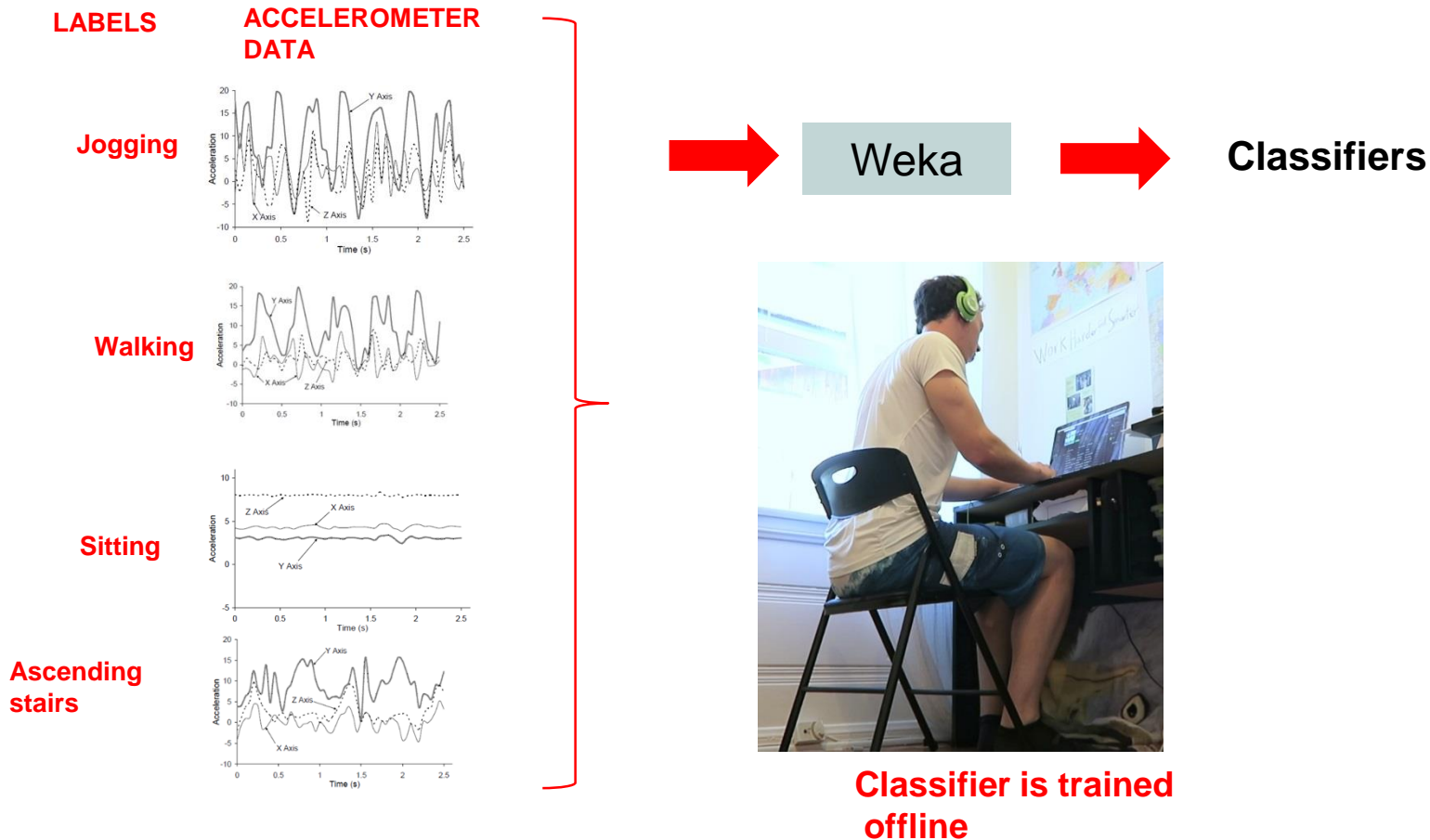


AndroSensor

Step 2: Import accelerometer samples into classification library (e.g. Weka, MATLAB)

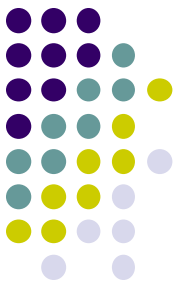


- Import accelerometer data (labelled with corresponding activity) into Weka (or other Machine learning Framework)

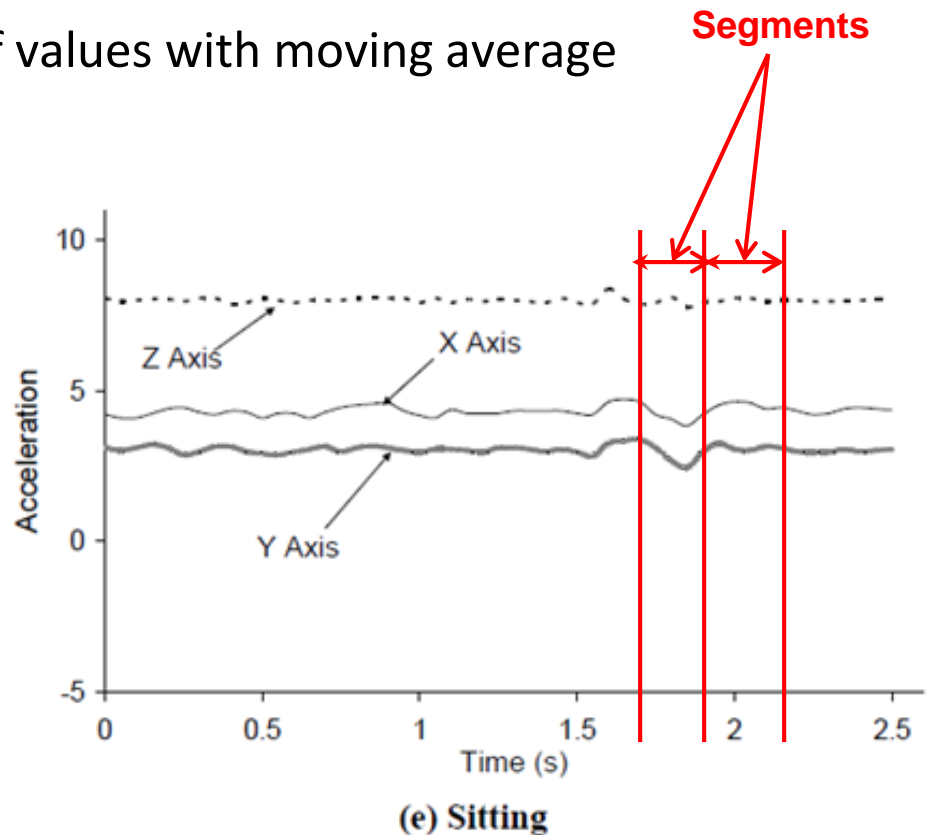


Step 3: Pre-processing (segmentation, smoothing, etc)

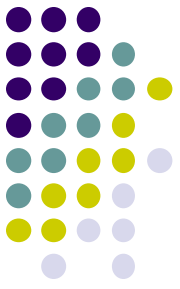
Segment Data (Windows)



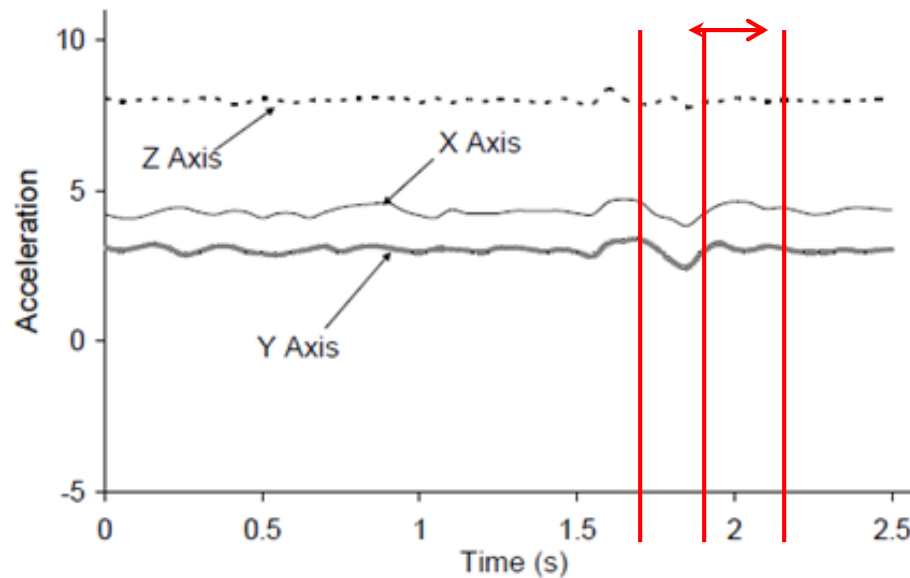
- Pre-processing data (in Weka) may include segmentation, smoothing, etc
 - **Segment:** Divide 60 seconds of raw time-series data divided into chunks(e.g. 10 seconds)
 - **Smoothing:** Replace groups of values with moving average



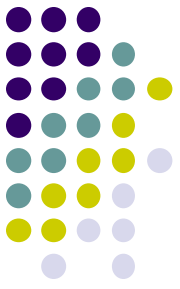
Step 4: Compute (Extract) Features



- For each segment (batch of accelerometer values) compute features (in Weka)
- **Features:** Functions computed on accelerometer data, captures important accelerometer characteristics
- **Examples:** min-max of values, largest magnitude within segment, standard deviation



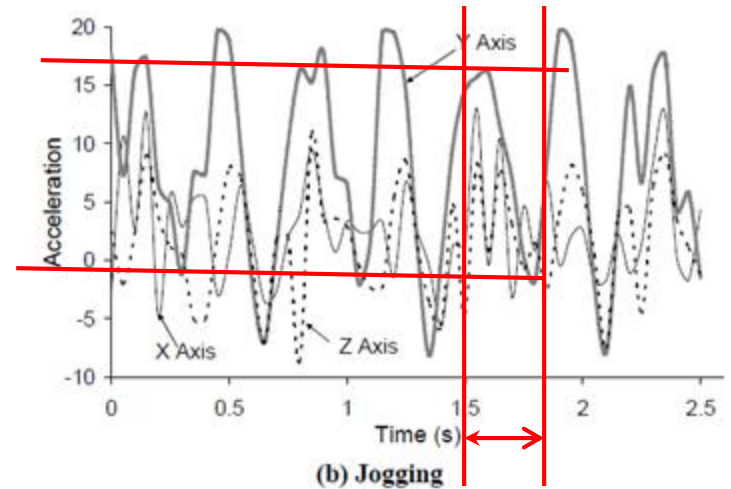
(e) Sitting



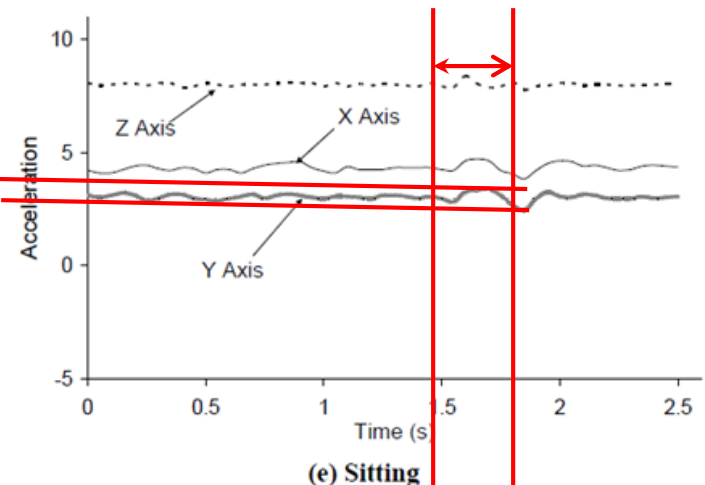
Step 4: Compute (Extract) Features

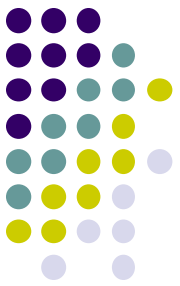
- **Important:** Ideally, values of features different for each activity type
- **E.g:** Min-max range feature

Large min-max
for jogging



Small min-max
for sitting



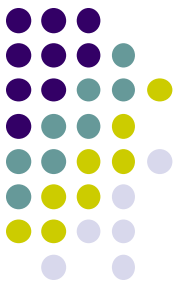


Step 4: Compute (Extract) Features

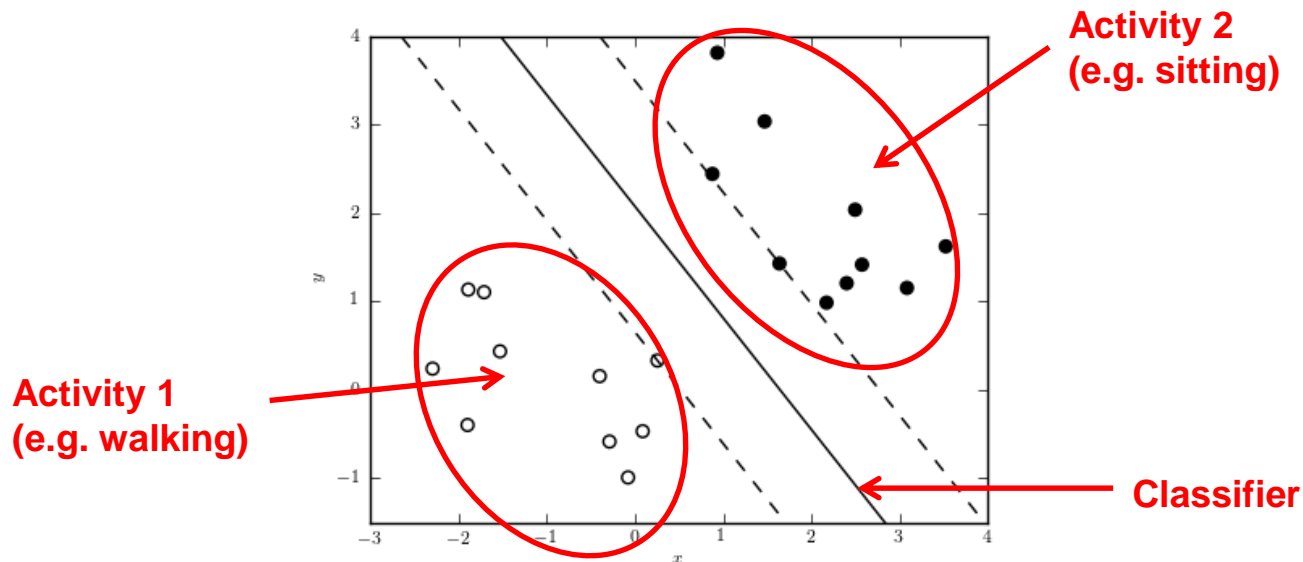
Calculate
many
different
features

- Average[3]: Average acceleration (for each axis)
- Standard Deviation[3]: Standard deviation (for each axis)
- Average Absolute Difference[3]: Average absolute difference between the value of each of the 200 readings within the ED and the mean value over those 200 values (for each axis)
- Average Resultant Acceleration[1]: Average of the square roots of the sum of the values of each axis squared $\sqrt{(x_i^2 + y_i^2 + z_i^2)}$ over the ED
- Time Between Peaks[3]: Time in milliseconds between peaks in the sinusoidal waves associated with most activities (for each axis)
- Binned Distribution[30]: We determine the range of values for each axis (maximum – minimum), divide this range into 10 equal sized bins, and then record what fraction of the 200 values fell within each of the bins.

Step 5: Train classifier



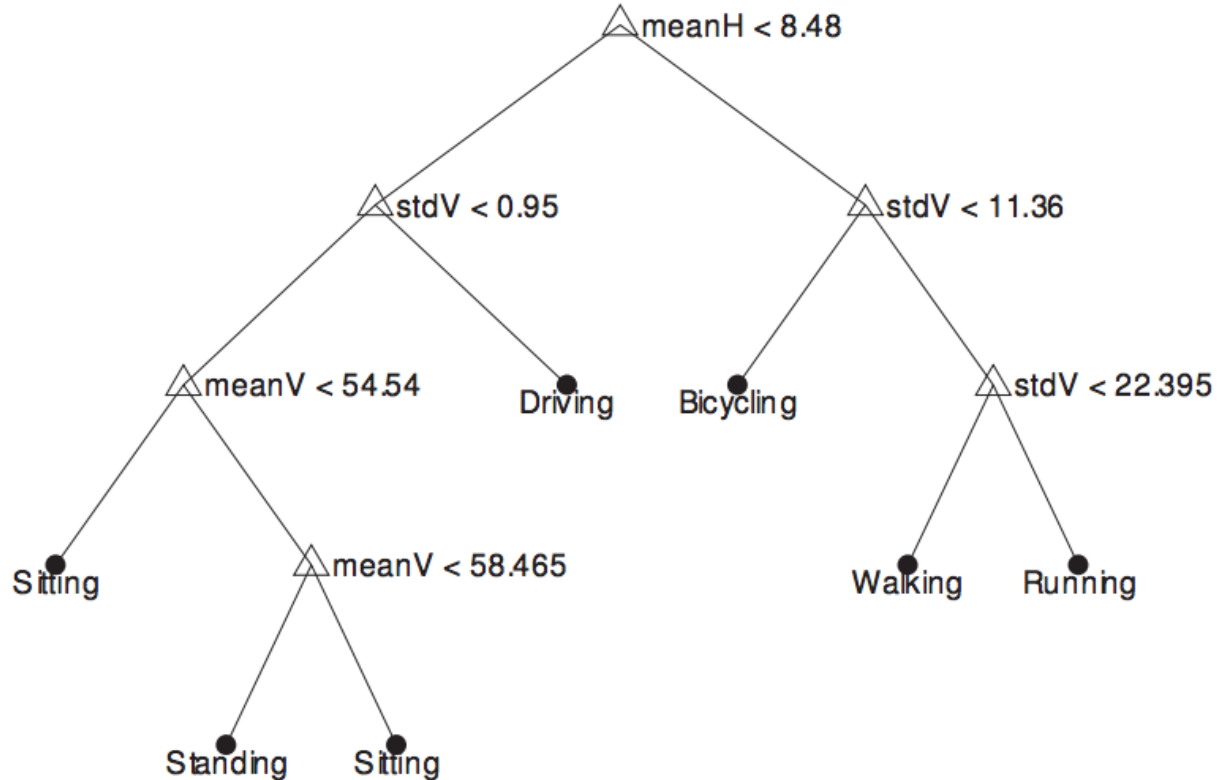
- Features are just numbers
- Different values for different activities
- **Training classifier:** figures out feature values corresponding to each activity
- Weka already programmed with different classification algorithms (SVM, Decision Trees, Naïve Bayes, Random Forest, J48, logistic regression, SMO, etc)
- Try different classification algorithms, compare accuracy
- **SVM example**





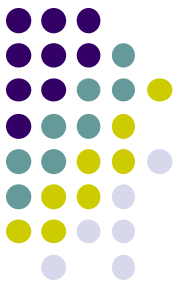
Step 5: Train classifier

- **Example:** Decision Tree Classifier
- Feature values compared against learned thresholds at each node



Step 5: Train classifier

Compare Accuracy of Classifier Algorithms



- Weka also reports accuracy of each classifier type

Table 2: Accuracies of Activity Recognition

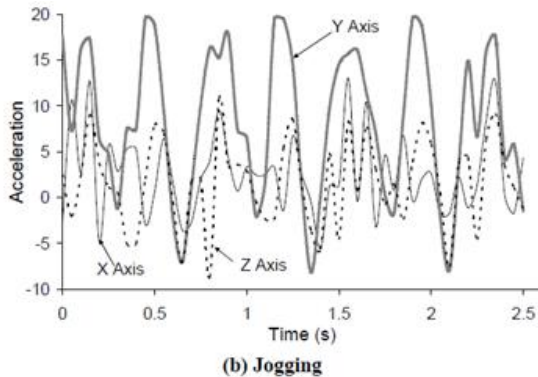
	% of Records Correctly Predicted			
	J48	Logistic Regression	Multilayer Perceptron	Straw Man
Walking	89.9	<u>93.6</u>	91.7	37.2
Jogging	96.5	98.0	<u>98.3</u>	29.2
Upstairs	59.3	27.5	<u>61.5</u>	12.2
Downstairs	<u>55.5</u>	12.3	44.3	10.0
Sitting	<u>95.7</u>	92.2	95.0	6.4
Standing	<u>93.3</u>	87.0	91.9	5.0
Overall	85.1	78.1	<u>91.7</u>	37.2

Pick most accurate
classification algorithm
for all classes

Step 6: Export classification model as JAR file

Step 7: Import into Android app

- Export classification model (most accurate classifier) as Java JAR file
- Import JAR file into Android app
- In app write Android code to
 - Gather accelerometer data, segment, extract feature, classify using classifier in JAR file
- Classifies new accelerometer patterns while user is performing activity => Guess (infer) what activity



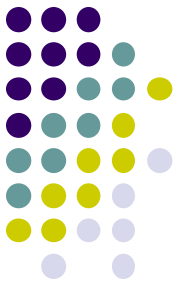
**New accelerometer
Sample in real time**



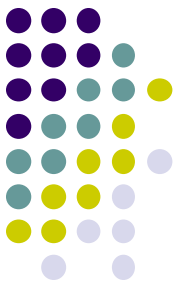
**Classifier in
Android app**



**Activity
(e.g. Jogging)**



Context Sensing

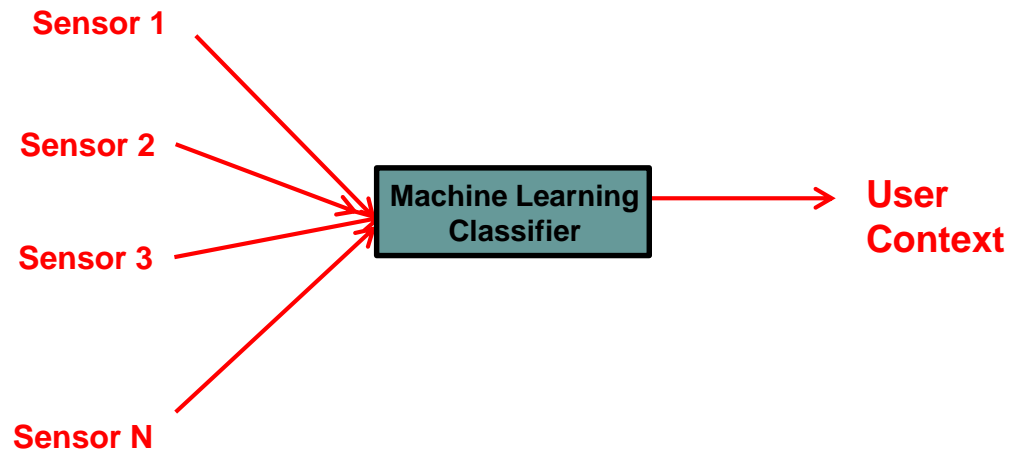
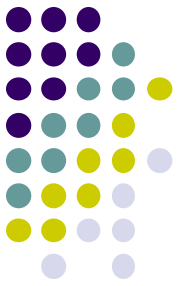


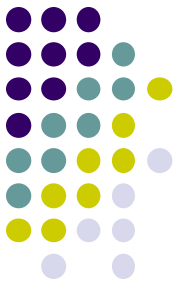
Recall: Ubicomp Senses User's Context

- Context?
 - *Human*: motion, mood, identity, gesture
 - *Environment*: temperature, sound, humidity, location
 - *Computing Resources*: Hard disk space, memory, bandwidth
 - *Ubicomp example*:
 - *Assistant senses*: Temperature outside is 10F (environment sensing) + Human plans to go work (schedule)
 - *Ubicomp assistant advises*: Dress warm!
- Sensed **environment + Human + Computer resources = Context**
- *Context-Aware* applications adapt their behavior to context

Context Sensing

- Activity Recognition uses data from only accelerometer (1 sensor)
- Can combine multiple sensors, use machine learning to sense **user context**
- More later





References

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