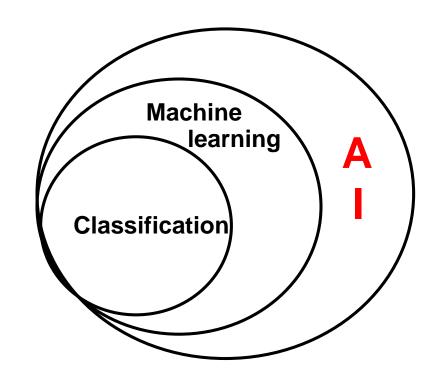
# LAB for Classification

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### How to Build Data Models

- □Split data into: training and testing set
- ☐ Build the classifier from the training set
  - Clean the data: checking missing values
  - Be careful about the categorical data
  - Standardize the data range
- ☐ Inference on the testing data set
- ■Visualize the data and classification results

## LAB 1: Diabetes

### Data sets

□Data/diabetes.csv □768 instances ■ 8 features(x) (no.1- no.8): Number of times pregnant Plasma glucose concentration a 2 hours in an oral glucose tolerance test Diastolic blood pressure (mm Hg) 4. Triceps skin fold thickness (mm) 5. 2-Hour serum insulin (mu U/ml) \ Body mass index (weight in kg/(height in m)^2) Diabetes pedigree function 8. Age (years) ☐ Response variable (y) (no.9): binary

## **Apply Classification Algorithms**

Algorithms

Linear Discrimination Analysis (LDA)

Linear Discrimination Analysis on Expanded Basis (ELDA)

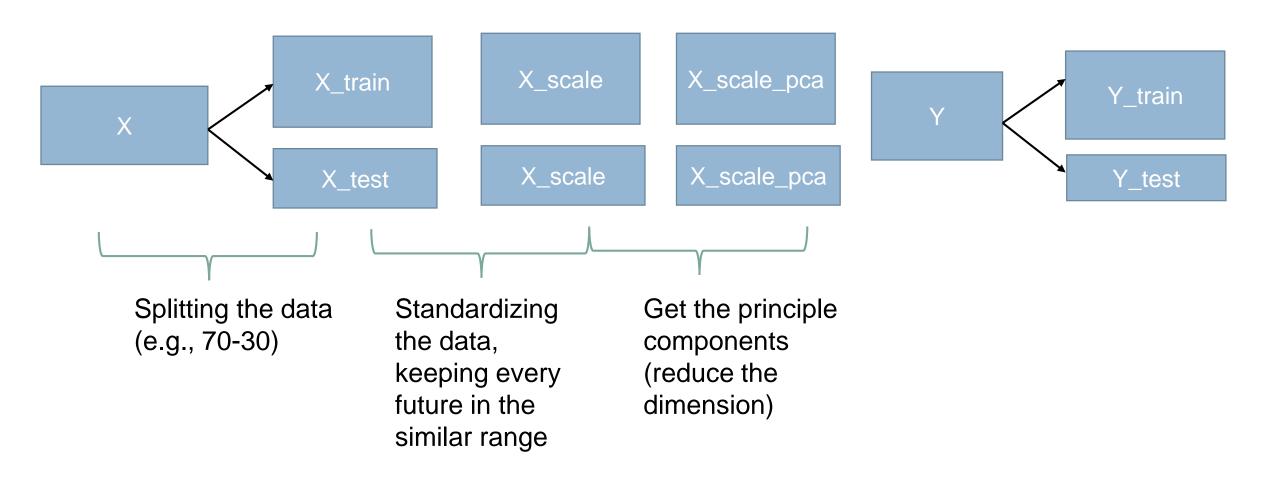
Quadratic Discrimination Analysis (QDA)

Logic Regression (LR)

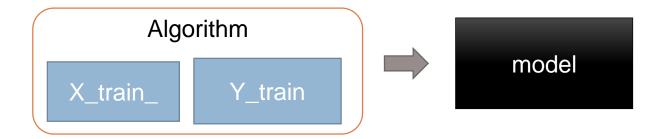
Models

Data (x,y)

#### 1. Creating training and testing data



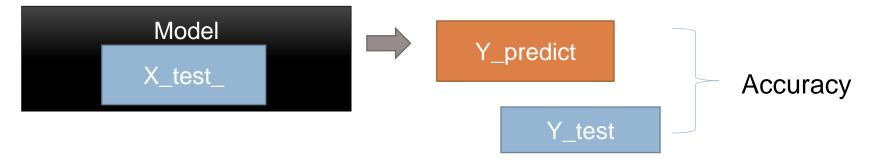
#### 2. Build the model on training set using a specific algorithm and training data



Algorithm: Ir, Ida, qda, elda, svm, dt, rf

Input: x\_train, x\_train\_scale, x\_train\_pca

#### 3. Check the accuracy on the testing set



Input: x\_test, x\_test\_scale, x\_test\_pca

### Basic structure of functions

```
☐ model= modelfunctionname().fit(X_train, Y_train)

☐ Y_model=model.predict(X_test)

Implication
□accuracy_score(Y_test, Y_model)*100))
Dompute and plot cm(Y test Y model.
     data class_labels, title="model")
Inlot 2d(X test, Y model, data class labels,
     title="model")
```

## LAB 2: Google Trace

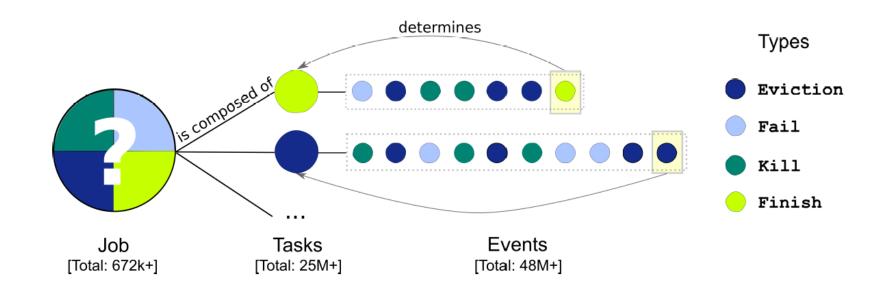
## Big data from performance logs



### Look into performance big data

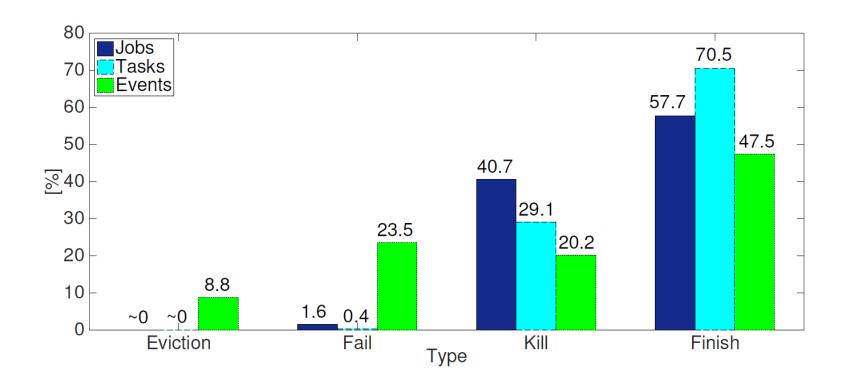


### A study on Google trace



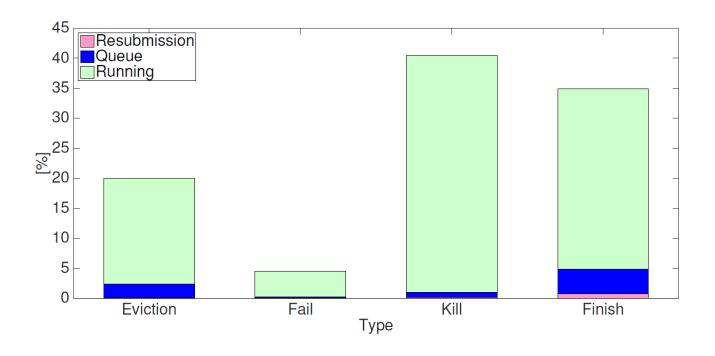
- Heterogeneous system and applications (multiple priorities)
- Multi-layer structure: job, task and event

#### Statistics on unsuccessful executions



- Many unsuccessful executions: job (42.3%), task (29.5%), and events (52.5%)

### The impact of unsuccessful events: wasted time



Resubmission: termination to sub. Queue: sub. to scheduling. Running: scheduling to termination

- More than 65% of time is wasted
- Events fail after a *short* while but are evicted after a *long* while

#### Relevant Features

Туре	group	Feature List	
Static	Event (5)	Task priority Task requested CPU/RAM/DISK	Task scheduling class
	Job (10)	Job size AVG/STD task priority AVG/STD lower/same/higher priority arrivals	Job scheduling class AVG/STD task requested CPU/RAM/DISK AVG/STD lower/same/higher priority variation of arrivals
	System(36)	AVG/STD lower/same/higher priority throughput AVG/STD lower/same/higher priority n. of tasks	AVG/STD lower/same/higher priority variation of throughput AVG/STD lower/same/higher priority variation of n. of tasks
Historical	Previous event type (4)	Number of previous finish events Number of previous fail events Previous event priority	Number of previous eviction events Number of previous kill events Previous event scheduling class
	Previous event data (15)	Previous event priority Previous event requested CPU/RAM/DISK Previous event queue/running time Previous event machine concurrency	Previous event type Previous event machine CPU/RAM capacity Previous event machine CPU/RAM reservation/utilization

#### **Key features:**

- Static feature (known upon arrival): system load, priority
- Historical feature: previous event

### Putting it together: machine learning

Linear Discrimination Analysis (LDA)

Linear Discrimination Analysis on Expanded Basis (ELDA)

Quadratic Discrimination Analysis (QDA)

Logic Regression (LR)

Decision Tree (DT)

Support Vector Machine (SVM)

**Neural Networks** 

#### **Questions:**

- How to predict event (more difficult), task, and job type?
- How to minimize the waste of resources/time?

### Data sets

- □Data/job.csv
- □1541720 instances
- □ 28 features (x) (no.1- no.28)
- ☐ Response variable (y): 4 classes