

Machine Learning : Lecture -1

Introduction

Topics (Part -1)

1. Linear modelling:

Least square (LS)

Non-linear response from linear models

Generalization versus overfitting

Regularized Least Squares:

L2, L1 regularization (LASSO)

2. Maximum likelihood (ML) approach

3. Classifiers:

Probabilistic Classifier: Bayes classifier, Logistic regression

Non-Probabilistic classifier: K-nearest neighbours

4. Decision Trees

5. Random Forests

6. Gradient Boosting

Tutorials: (Mon 5.00 PM -6.30 PM)

R-102 – Batch 1

R-104 – Batch 2

R-105 – Batch 3

R-110 – Batch 4

Text Books:

1 , 2, 3: A first Course on Machine Learning by Simon Rogers

4,5,6 : Multiple references, will be mentioned on slides.

Reference:

Machine Learning: Foundations, Methodologies, and Applications by Alexander Jung

Assessment Plan: Part 1 (50 Marks)

Class Quiz (After Aug 31) : 10 Marks

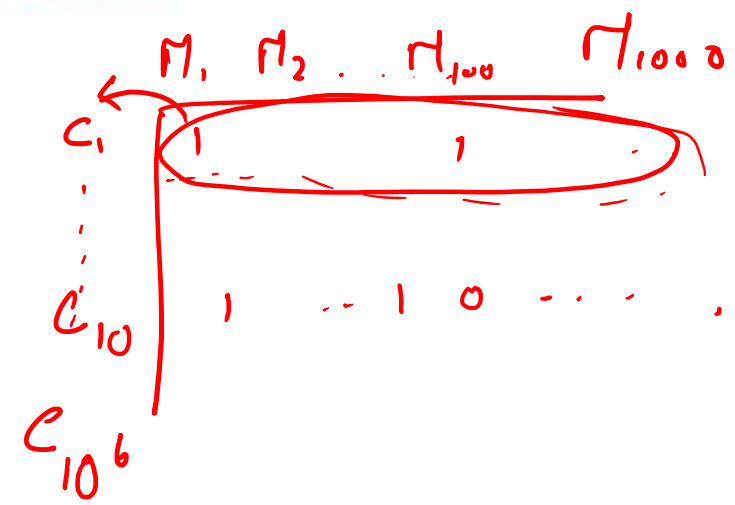
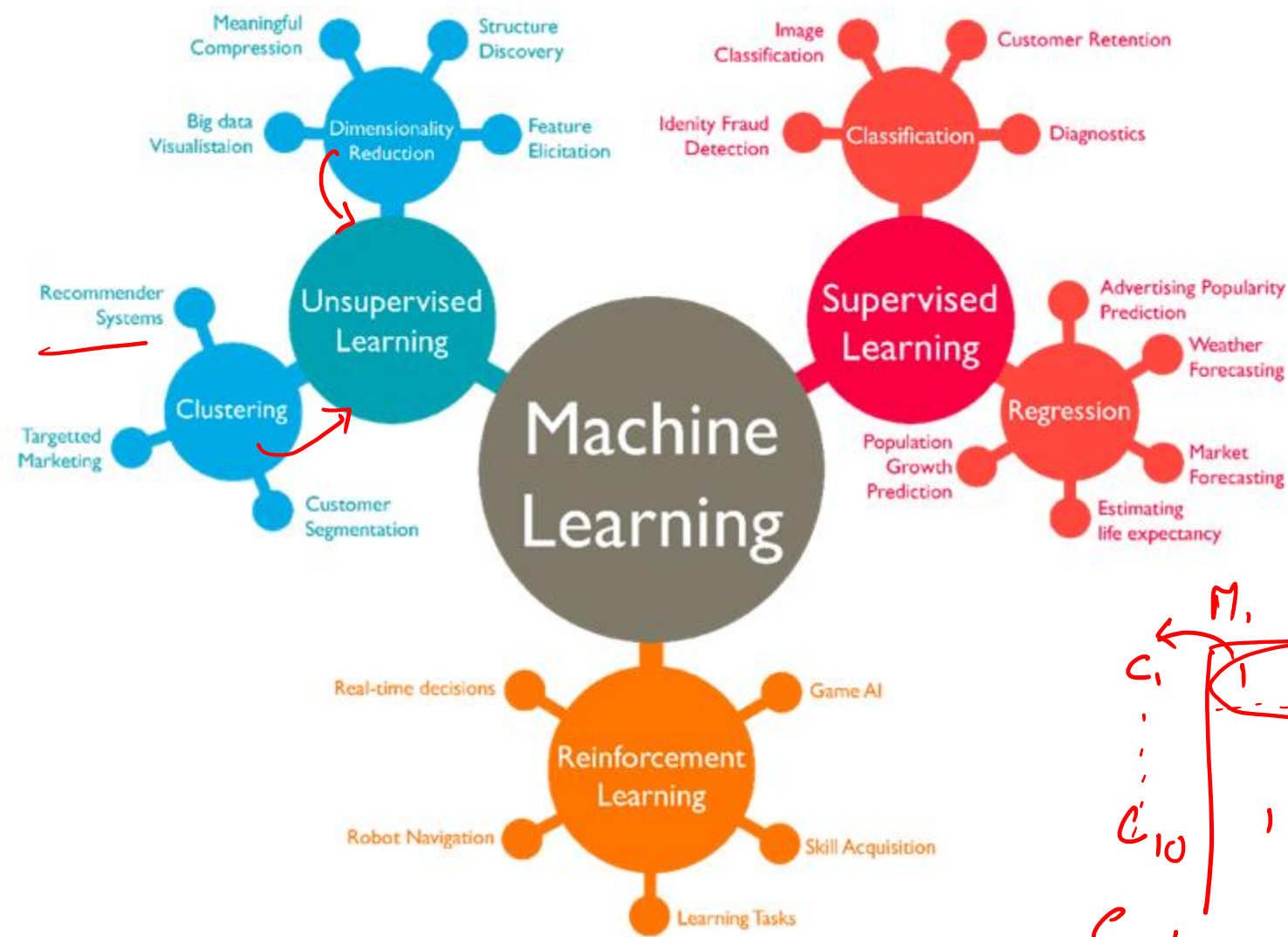
Assignment – 1 (After Sep 15) : 10 Marks

Mid-term: 20 Marks

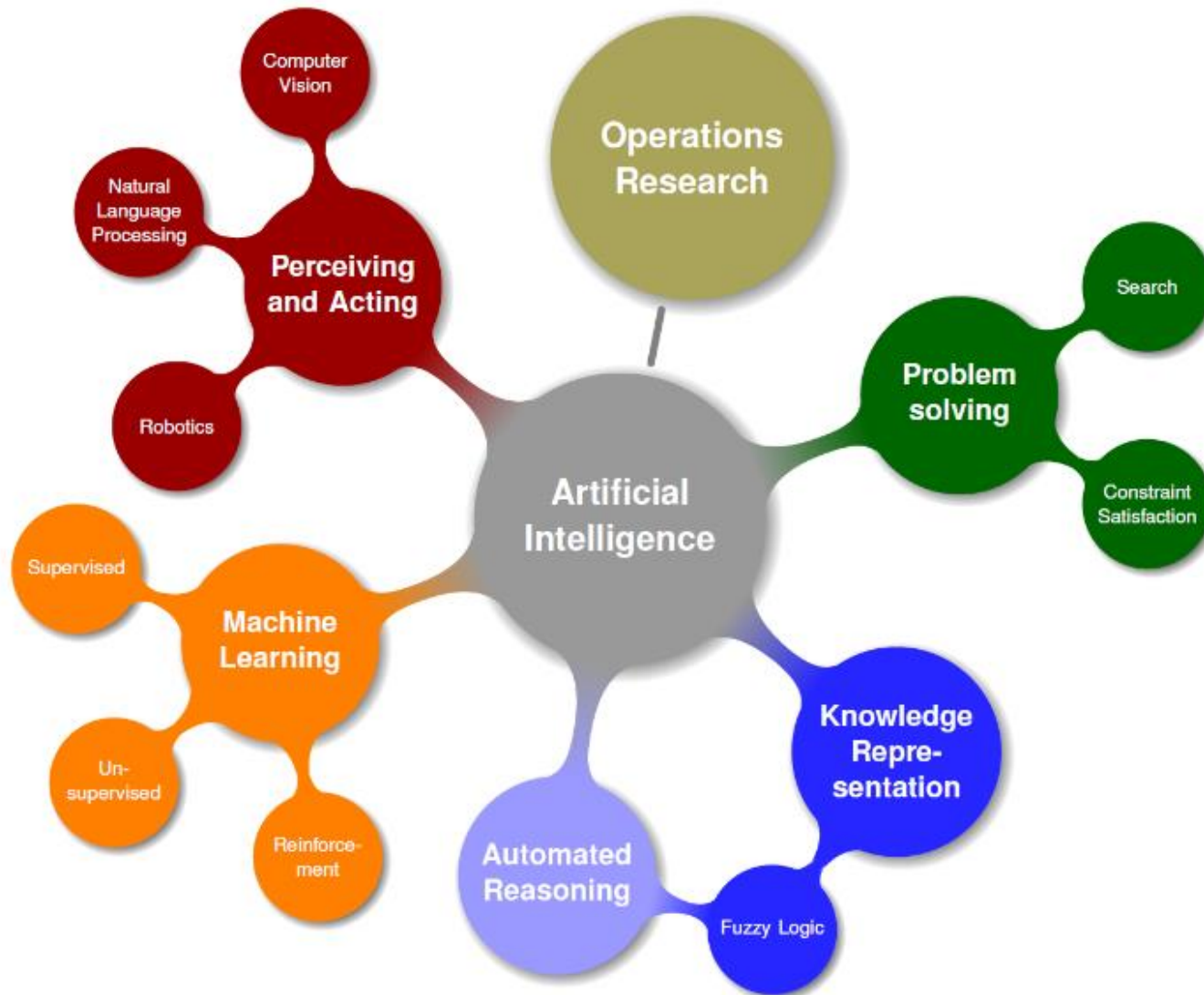
Tutorial Assessment (Multiple assessments spread across random tutorial days): 5 Marks

Class Participation: 5 Marks

Machine Learning



Machine Learning and Artificial Intelligence



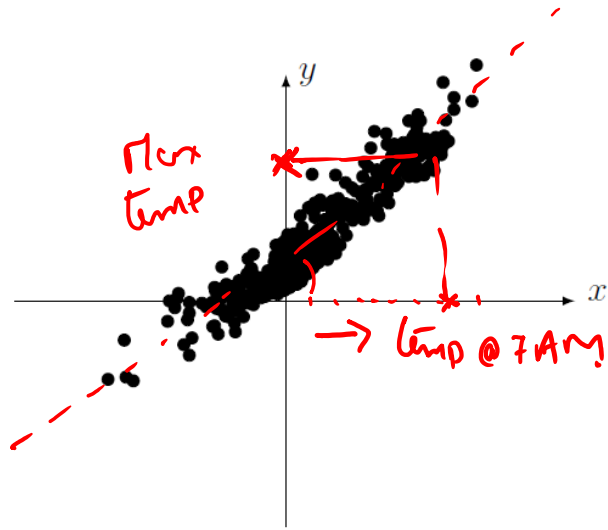
Artificial intelligence is a set of technologies implemented *in* a system to enable it to reason, learn, and act to solve a complex problem.

A Simple Learning Example

$$\underline{w} \in \mathbb{R} \quad \underline{x} \in \mathbb{R}^2$$

Objective: Predict the maximum day temperature after observing the temperature at 7 AM.

Observations:



x : Temperature at 7 AM

y : Maximum Day Temperature

How can we find suitable w_1 and w_0 ?

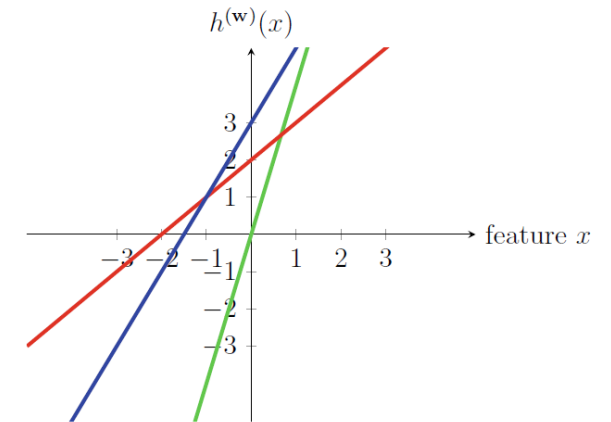
Sample hypotheses:

$$\underline{h(x)} := \underline{w_1 x + w_0} \quad \underline{w_1 \in \mathbb{R}_+}, \underline{w_0 \in \mathbb{R}}.$$

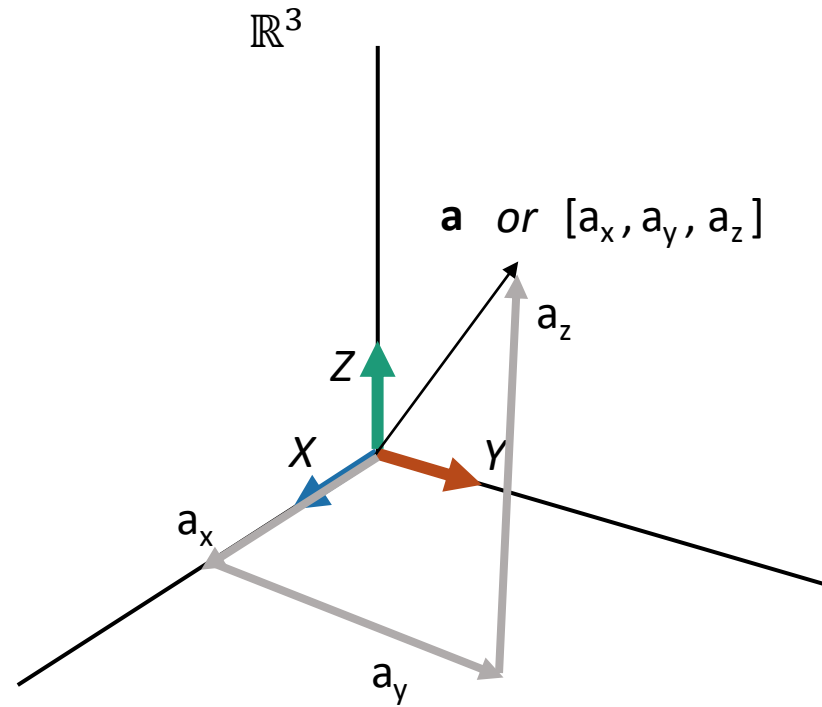
Model:

$$y \approx \underline{w_1 x + w_0}$$

ML Objective: Find the "best" hypothesis from the set of feasible hypotheses



[ML & Linear Algebra] : Vector Representation



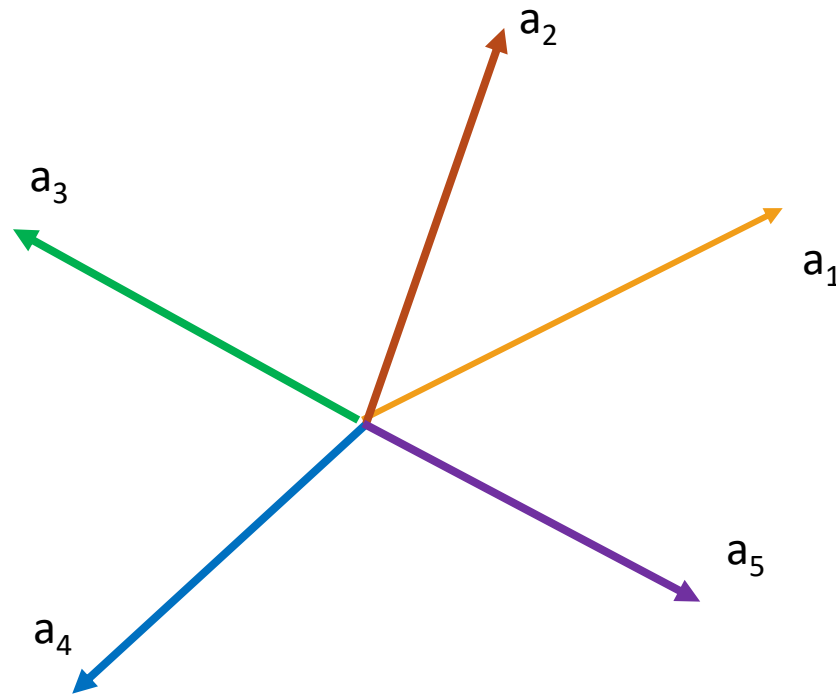
$$\mathbf{a} \in \mathbb{R}^3$$

$[a_x, a_y, a_z]$ ➡ Point in 3- dimensional Space /
3-dimensional Vector / **Data**

[ML & Linear Algebra] : Data in Multi-Dimensional Space

5 –Dimensional Space (\mathbb{R}^5)

(Do not bother to Imagine !
Look only Algebraic way)



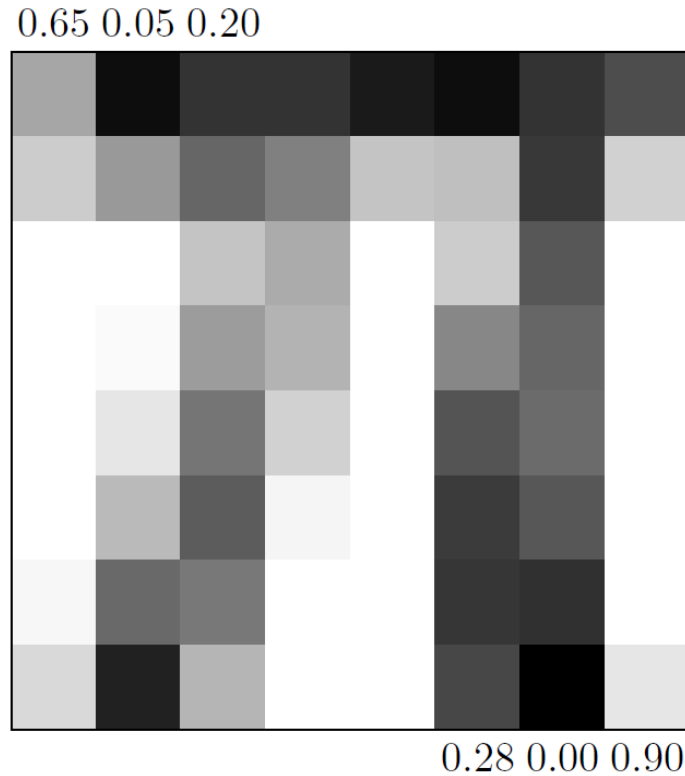
Dimension	Meaning	Value
a_1	Height (ft)	6
a_2	Age	30
a_3	Weight (kg)	70
a_4	Waist-Size(in)	32
a_5	Gender	1

$$a_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad a_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad a_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

$$a_4 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} \quad a_5 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\mathbf{a} = [6, 30, 70, 32, 1] \Rightarrow \mathbf{a} = 6\mathbf{a}_1 + 30\mathbf{a}_2 + 70\mathbf{a}_3 + 32\mathbf{a}_4 + \mathbf{a}_5$$

[ML & Linear Algebra] : Image Representation



8 × 8 image

$$x \in \mathbb{R}^{64 \times 32}$$

$$x \in \mathbb{R}^{32 \times 64}$$

8 × 8 image can be represented as 64 dimensional vector

$$x = [0.65 \ 0.05 \ 0.20 \ \dots \ 0.28 \ 0.00 \ 0.90]$$

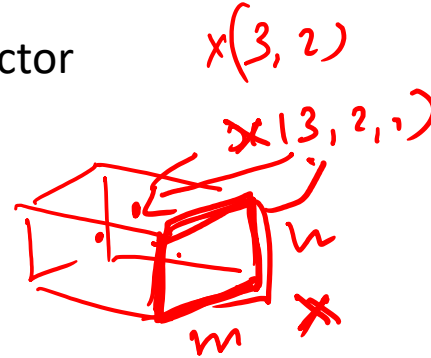
x is a point in 64-dimension space, $x \in \mathbb{R}^{64}$

What does $x \in \mathbb{R}^{8 \times 8}$ mean ?

Videos:

k frames in a video of resolution $m \times n$ can be represented as as $m \times n \times k$ vector

$$x \in \mathbb{R}^{m \times n \times k}$$



$$x = x(3, 2, 1)$$
$$x(:, :, 1)$$

Poll : $x \in \mathbb{R}^{m \times n \times k}$ & $x \in \mathbb{R}^{m \times k \times n}$ represents the same video ?

[ML & Linear Algebra] : Document Representation

Documents:

D1 : I like deep learning

D2 : I like NLP

D3 : I enjoy flying

$D1 \in \mathbb{R}^{3000}$ $D2 \in \mathbb{R}^{3000}$ $3K$

$D1: [1 \dots 0 \dots 1 \dots 0 \dots]$

$D2: [1 \dots 0 \dots 1 \dots 0 \dots]$

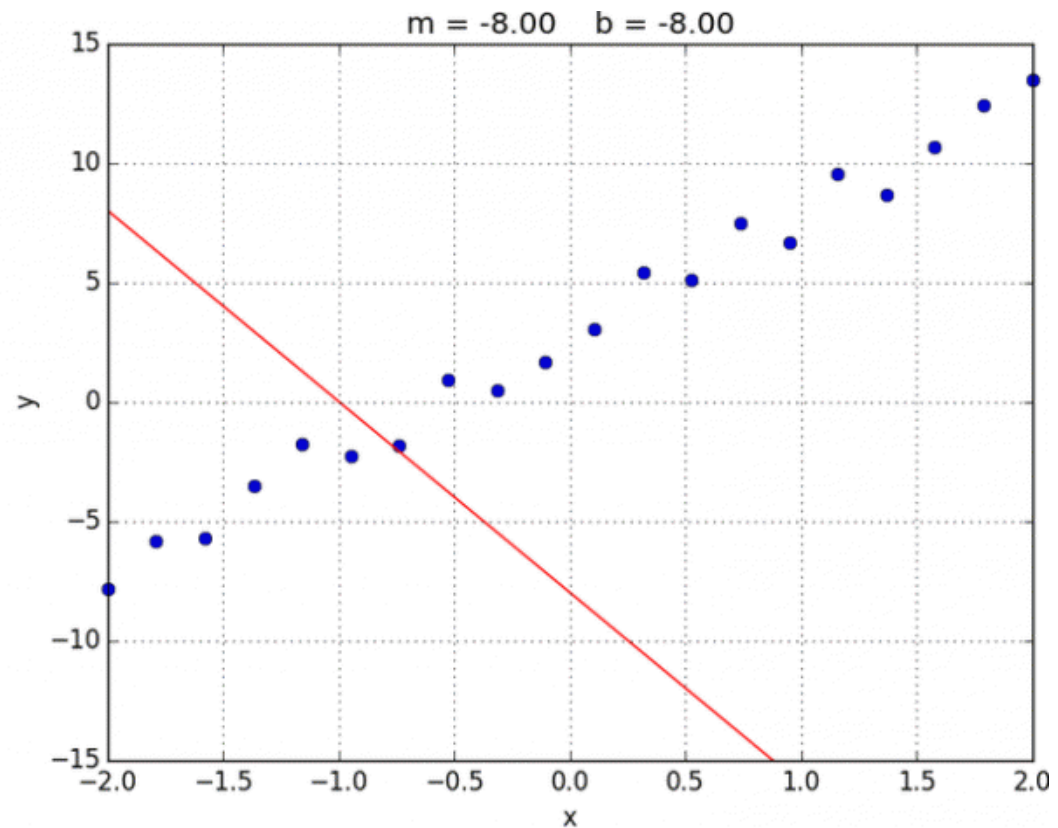
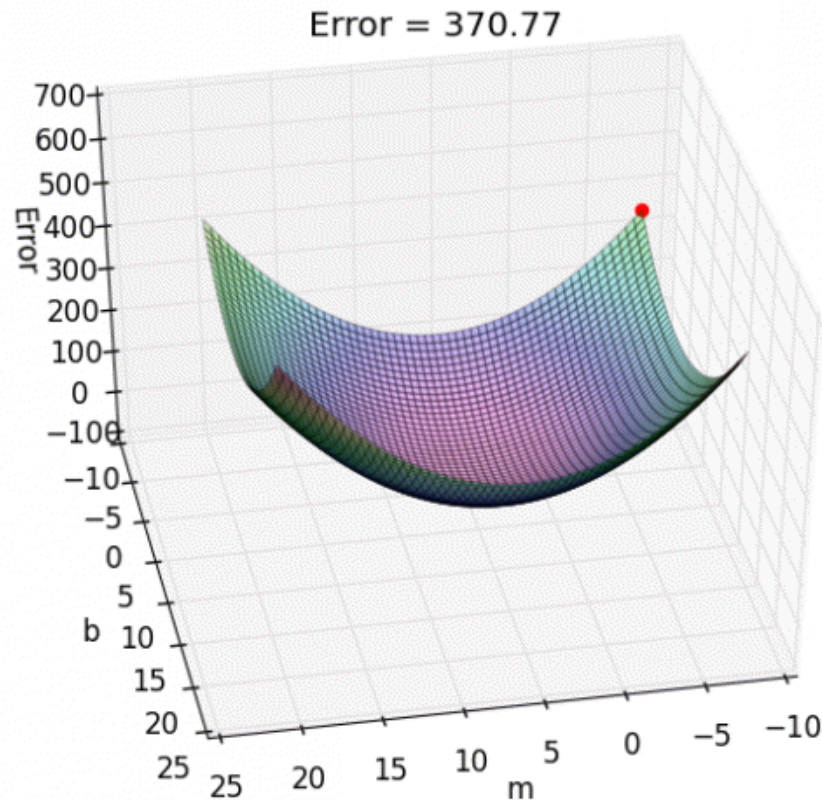
(V)

Word Count Histogram:

counts	D1	D2	D3
I	1 ✓	1	1
like	1 ✓	1	0
enjoy	0	0	1
deep	1 ✓	0	0
learning	1 ✓	0	0
NLP	0	1	0
flying	0	0	1
.	1	1	1

[ML & Optimization] Cost Function Minimization

$$J(y|x)$$

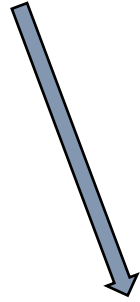


Line Fitting: Loss function minimization with two parameters

[ML & Probability]

What if we can model the joint probability distribution of data and labels?

What if we can additionally sample new data from this joint distribution?



Generative Models

7 AM Temperature vs. Max Day temperature:

