### SMAI-M20-05: Features, Data and Learning

C. V. Jawahar

IIIT Hyderabad

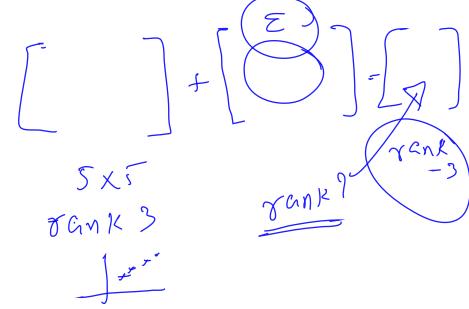
August 19, 2020

#### Recap

- **1** Learn a function  $y = f(\mathbf{W}, \mathbf{x})$  from the data:
  - Representation as a vector in R<sup>d</sup>
  - Learnable parameters W
  - Notion of Training and Testing
- Feature Transformation as a useful trick:
  - $\mathbf{v}' = \mathbf{W}\mathbf{x}'$
  - Dimensionality Reduction
- Three Classification Schemes:
  - Nearest Neighbour Algorithm
  - Linear Classification
  - Decide as  $\omega_1$  if  $P(\omega_1|\mathbf{x}) \geq P(\omega_2|\mathbf{x})$  else  $\omega_2$
- Performance Metrics:
  - Classification: Accuracy, TP/FP etc., Confusion Matrix
  - Ranking: Precision, Recall, F-Score, AP

#### This Lecture: Appreciating Data

- Three Video Clips:
  - Feature Selection and Feature Extraction
  - "Features as a Vector" to "Data as a Matrix"
  - Formulating Supervised Learning Task
- Progressing from "Raw" representation to "Useful features":
  - Examples of "raw" data; Eg. Pixels in an image
  - Subset selection
  - Linear transformation; including Dim. Reduction
  - Non-linear Transformations (more later)
  - Learning the features Eg. Deep Embeddings (more later)
- Data as matrix
  - Data is not "random"
  - "Structure in the data" leading to "data lie in a subspace".
  - Eg. Why are we not seeing samples with any  $\frac{ht}{wt}$ ?



# **Questions?** Comments?

#### Discussions Point - I

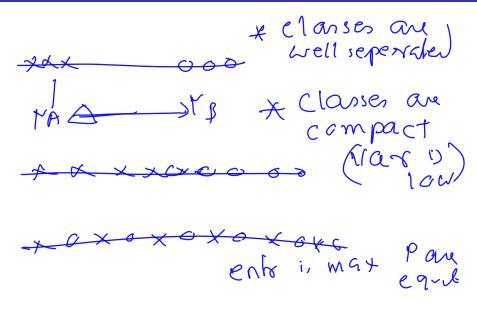
In supervised learning, we use the training data  $\mathcal{D}_{Tr}$  to develop a solution and then evaluate it on the test data  $\mathcal{D}_{Te}$ 

Q: Why is that a solution optimized only on training data as

$$\min_{\boldsymbol{W},} \mathcal{L}(\boldsymbol{W}, \boldsymbol{x}) \ \boldsymbol{x} \in \mathcal{D}_{\textit{Tr}}$$

works reasonably well on the test data also?

- We believe in luck.
- ② Once it is trained on  $\mathcal{D}_{Tr}$ , it then works well on any data.
- **3** Actually, we use  $\mathcal{D}_{Te}$  also for training. Eqn. is incorrect.
- **1** What we care is actually the performance on the  $\mathcal{D}_{Tr}$ ; not  $\mathcal{D}_{Te}$ .
- **1** It is some magic. No explanation found till today.



#### Discussions Point -II

Consider a problem where the original representation d (say 100) and N (say 1000) samples. Problem is binary classification. (Class 'A' Vs. Class 'B'). Assume each of these d features are in the range [0,10].

Q: We want to "select" top 10 features based on a "quality measure" (how good this feature could be for the classification) of the feature.

#### We rank features based on:

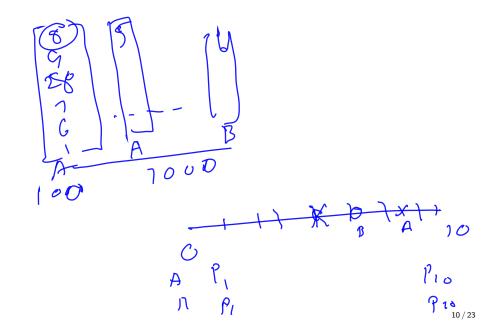
• We rank each feature j based on  $|\mu_A^j - \mu_B^j|$ 

**②** We rank each feature j based on  $\frac{(\sigma_A^j)^2 + (\sigma_B^j)^2}{2}$ 



• We compute the probability of A and B in 10 intervals (say  $[0,1],[1,2]\dots[9,10]$ ) as  $P_i$  and compute entropy for each feature as:

$$H = \sum_{i=1}^{10} P_i^A \log P_i^A + \sum_{i=1}^{10} P_i^B \log P_i^B$$

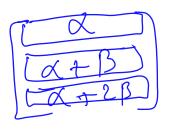


#### Discussion Point - III

Without numerically computing, what is the rank of the matrix? Show:

$$\left[\begin{array}{ccc}
1 & 2 & 3 \\
4 & 5 & 6 \\
7 & 8 & 9
\end{array}\right]$$

- **1**
- **2** 2
- **3** 3
- 4



Hint:

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 1+3 & 2+3 & 3+3 \\ 1+6 & 2+6 & 3+6 \end{bmatrix}$$

### Review Question - I (one, none or more correct)

We know that the rank of a  $3\times 3$  matrix formed by first 9 numbers arranged sequentially is 2.

What is the rank of a  $5\times 5$  matrix formed by first 25 numbers arranged sequentially?

(a) 1 (b) 2 (c) 3 (d) 4 (e) 5 (f) none of the above

### Review Question - II (one, none or more correct)

A certain test for disease is known to have True positive of 0.6 and False Positive of 0.1.

A population of 100 people (where 60 of them are infected) undergoes this test.

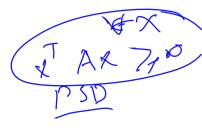
What could be the confusion matrix?

(a) 
$$\begin{bmatrix} 0.6 & 0.4 \\ 0.1 & 0.9 \end{bmatrix}$$
 (b)  $\begin{bmatrix} 0.6 & 0.4 \\ 0.9 & 0.1 \end{bmatrix}$  (c)  $\begin{bmatrix} 0.6 & 0.2 \\ 0.1 & 0.3 \end{bmatrix}$  (d)  $\begin{bmatrix} 0.58 & 0.42 \\ 0.15 & 0.85 \end{bmatrix}$  (e) None of the above

### Review Question - III (one, none or more correct)

#### Covariance Matrix:

- (a) Can neve be diagonal.
- (b) Can be diagonal
- (c) Always Positive Semi Definite
- (d) Always full rank
- (e) Never full rank.
- (f) Always Symmetric
- (g) Not guaranteed to be symmetric





## Review Question - IV (one, none or more correct)

Let 
$$\mu = \sum_{i=1}^{N} \mathbf{x}_i$$
 be the mean of  $\mathbf{x}_1, \dots, \mathbf{x}_N$ .  $\mathbf{x}_i \in R^d$ 

• (a)  $\mu$  is also  $\in R^d$ 

- (b)  $\mu$  is always one of the N samples.
- (c)  $\mu$  can not be one of the N samples
- (d)  $\mu$  can be one of the N samples.
- (e)  $\mu$  is equidistant from all the N samples
- $\bullet$  (f)  $\mu$  has the least sum of square error from all the samples (i.e.,  $\mu$  is  $\underset{\text{arg min}_{\mathbf{v}}}{\text{min}_{\mathbf{v}}} \sum_{i=1}^{N} ||\mathbf{x}_{i} - \mathbf{v}||_{2}^{2}$

### Review Question - V (one, none or more correct)

You are planning a picnic today, but the morning is cloudy

- 50% of all rainy days start off cloudy.
- But cloudy mornings are common (about 40% of days start cloudy)
- This is usually a dry month (only 3 of 30 days tend to be rainy, or 10%)

What is the chance of rain during the day?

(a) 
$$10\%$$
 (b)  $12.5\%$  (c)  $15\%$  (d)  $> 20\%$  (e)  $< 20\%$ 

Hint: Bayes Rule:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

#### What Next: Two Sessions?

- Eigen Values/Vectors, SVD, Rank and Data Matrix
- More into Supervised Learning and the associated issues
- Bayesian Optimal Classification