

Date: \_\_\_\_\_

## Conceptual part.

Covariance: used to identify the relationship between two features i.e positive trend, negative trend, no trend.

### PCA : Principal component Analysis

- ① find origin of data
  - ② draw a random line crossing through that origin
  - ③ project all data points onto that line
  - ④ find the line that maximizes the distances from the projected points to the origin OR find a line that minimizes the distances of each data point to the origin.
  - ⑤ sum of squared distances, minimize/maximize that value accordingly.
  - ⑥ After we have our principal component no. 1 or first eigen vector we find our second eigen vector which is perpendicular to the PC 1.
  - ⑦ Total number of PCs depend on number of variables/features
  - ⑧ After we have all our PCs we check which PC/s can accumulate most ~~that relationship~~ of variables, we choose them and plot the data on a decreased ~~2~~ dimensioned plane.
- Tips: ① Scale the data ② Centre the data ③



Date: \_\_\_\_\_

## correlation and convolution on image:

\* convolution combines two functions and shows how they convolve together. Let the first function be an image and the second be the gaussian blur function. If you combine both, convolutional filter would apply the second function i.e. the gaussian blur function on each pixel of the image passed as the first function. As an end result, we will receive a smoother picture.

\* cross-correlation, is usually used for object detection or similarity detection. We give two functions, an image and another image which may be smaller. The cross-corr slides the smaller image across the larger one and calculates on each slide how similar ~~an~~ that part is to the small image i.e. correlation.



Date: \_\_\_\_\_

## Bag of Words:

① we make vectors of each sentence in a corpus using the 0s and 1s as unrepresent and present word of a corpus in that specific sentence.

## TF-IDF:

① we make a table of ~~the~~ number of each term in each document

② use formula:

$$tfidf = tf \times \log \left( \frac{N}{df} \right)$$

term frequency

number of documents

number of documents containing that term.

## n-grams:

→ set of words containing n elements.

→ calculate probability of a word with the history of the word. (predict the next word)

→ calculate probability of a sentence

$$P(w|h) = \frac{\text{count}(w) + \text{count}(h)}{\text{count}(w)}$$

Date: \_\_\_\_\_

## Data Refinement techniques:

we can clean/refine our data in following ways:

- ① Remove duplicate tuples / Nulls / irrelevant
- ② Fix structural errors (spellings, typos etc)
- ③ Filter unwanted outliers.
- ④ Handle missing data

## Graph & when to use them:

\* tabular data.

- ① Bar Chart : use when \* more than 10 features in data  
\* show trends \* relationships of features
- ② Line Chart : \* show relationship between variables
- ③ Scatter plot : \* show part-to-whole comparison
- ④ Pie Chart : \* identify smallest, largest parts
- ⑤ Bubble Chart : \* show relation between more than 2 features



Date: \_\_\_\_\_

\* Convolutional Neural Networks: CNNs are often used for image classification, they can identify objects in images by learning and recognizing important features.

→ CNN can be used for image data, classification prediction problems and regression prediction problem.

\* Multi-layer perceptrons: can be used for ~~text, image, time series, and other types of~~ <sup>tabular data</sup> data. It is the classical neural network with hidden layers.

\* Recurrent Neural network: used for text, speech, classification, regression, generative models. They don't work on image or tabular data.

### Types of Gradient descent:

\* Stochastic: one <sup>example</sup> at a time

\* mini-batch: apply GD with many examples at a time but still less than the total dataset of examples.

\* batch: → if we input entire dataset at a time and update weight, that batch GD.

epoch → go through all examples once.

Date: \_\_\_\_\_

### \* Types of activation functions in neural networks:

- An activation function decides whether a single neuron in a network should be activated / should participate or not.
- The ultimate outcome of using activation functions is non-linearity in our decision boundary.

① Sigmoid :  $f(x) = \frac{1}{1 + e^{-x}}$  range  $(0, 1)$

② Hyperbolic tangent / tanh function :

slowest  $f(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$  output range  $(-1, 1)$

③ Softmax : a combo of ~~diff~~ multiple sigmoids.

$$f(x_i) = \frac{e^{x_i}}{\sum_j e^{x_j}}$$

$$z_1, z_2, \dots, z_n$$
$$f(z_1) = \frac{e^{z_1}}{e^{z_1} + e^{z_2} + \dots + e^{z_n}}$$

$$f(z_i) = \frac{e^{z_i}}{e^{z_1} + e^{z_2} + \dots + e^{z_n}}$$



Date: \_\_\_\_\_

## Image Filtering:

① Lowpass filters (Smoothing)

→ removes high spatial frequency noise

② High pass filters (edge detection, sharpening)

→ make image sharp

→ emphasize details

\* Mean filter / Average

→ used to smooth images, reduce noise

$$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

Gaussian

\* ~~Median~~ filter

Gaussian noise

→ used to remove noise (especially ~~salt & pepper~~)

$$\begin{bmatrix} 2 & 1 & 2 \\ 1 & 4 & 1 \\ 2 & 1 & 2 \end{bmatrix} \quad \frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

linear

\* Median filter

→ removes noise, salt & pepper

→ arrange elements

into order & choose middle

element & replace

non-linear