



Getting to Know Your Data-II

IT- 303 I (Cr-3)

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Data Normalization



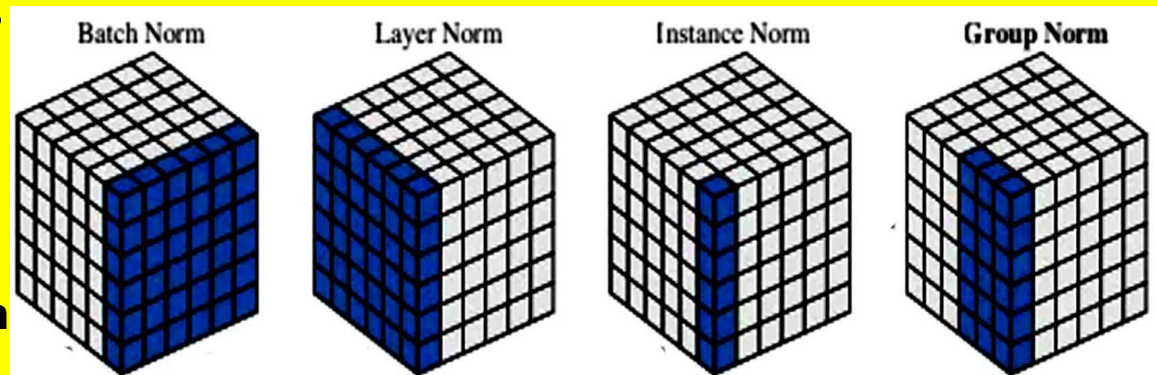
- Normalization is generally required when multiple attributes are there but attributes have values on different scales, this may lead to poor data models while performing data mining operations.
- Otherwise, it may lead to a dilution in effectiveness of an important equally important attribute(on lower scale) because of other attribute having values on larger scale.
- Heterogenous data with different units usually needs to be normalized. Otherwise, data has the same unit and same order of magnitude it might not be necessary with normalization.
- Unless normalized at pre-processing, variables with disparate ranges or varying precision acquire different driving values.

Methods of Data Normalization

- Normalization is normally done, when there is a distance computation involved in our algorithm.
- Methods of Data Normalization:
 - **Decimal Scaling**
 - **Min-Max Normalization**
 - **z-Score Normalization(zero-mean Normalization)**

- There are several approaches in normalisation which can be used in deep learning

- **Batch Normalization**
- **Layer Normalization**
- **Group Normalization**
- **Instance Normalization**
- **Weight Normalization**



Methods of Data Normalization...



◦ Decimal Scaling Method For Normalization

- It normalizes by moving the decimal point of values of the data.
- To normalize the data by this technique, we divide each value of the data by the maximum absolute value of data.
- The data value, v_i , of data is normalized to v'_i by using the formula

$$v'_i = \frac{v_i}{10^j}$$

where j is the smallest integer such that $\max(|v'_i|) < 1$.

In this technique, the computation is generally scaled in terms of decimals. It means that the result is generally scaled by multiplying or dividing it with $\text{pow}(10, k)$.

Example:

- Let the input data is: -15, 121, 201, 421, 561, 601, 850
- To normalize the above data,
- **Step 1:** Maximum absolute value in given data (m): 850
- **Step 2:** Divide the given data by 1000 (i.e $j=3$)
- Result: The normalized data is: -0.015, 0.121, 0.201, 0.421, 0.561, 0.601, 0.85

Methods of Data Normalization...



◦ Min-Max Normalization

- In this technique of data normalization, linear transformation is performed on the original data.
- Minimum and maximum value from data is fetched and each value is replaced according to the following formula.

$$v' = \frac{v - \min(A)}{\max(A) - \min(A)} (\text{new_max}(A) - \text{new_min}(A)) + \text{new_min}(A)$$

- Where A is the attribute data,
- **min**(A), **max**(A) are the minimum and maximum absolute value of A respectively.
- v' is the new value of each entry in data.
- v is the old value of each entry in data.
- $\text{new_max}(A)$, $\text{new_min}(A)$ is the max and min value of the range(i.e boundary value of range required) respectively.

Roll No	Marks
1	10
2	15
3	50
4	60

Example

If we were to normalize it between the ranges of 0 to 1 we would get the following



Roll No	Marks
1	0
2	0.1
3	0.8
4	1

Methods of Data Normalization...



- **z-Score Normalization (zero-mean Normalization)**

- In this technique, values are normalized based on mean and standard deviation of the dataA.
- It is also called **Standard Deviation method**.
- So, the unstructured data can be normalized using z-score parameter, the formula for z-score is as below;

$$v' = \frac{v - \bar{x}}{s}$$

where, \bar{x} is the mean and s is the standard deviation.

v is the old value of each entry in data.

v' is the Z-score normalized of each entry in data.

Example

Roll No	Marks
1	10
2	15
3	50
4	60

Mean is 33.75 and Standard Deviation is 24.95

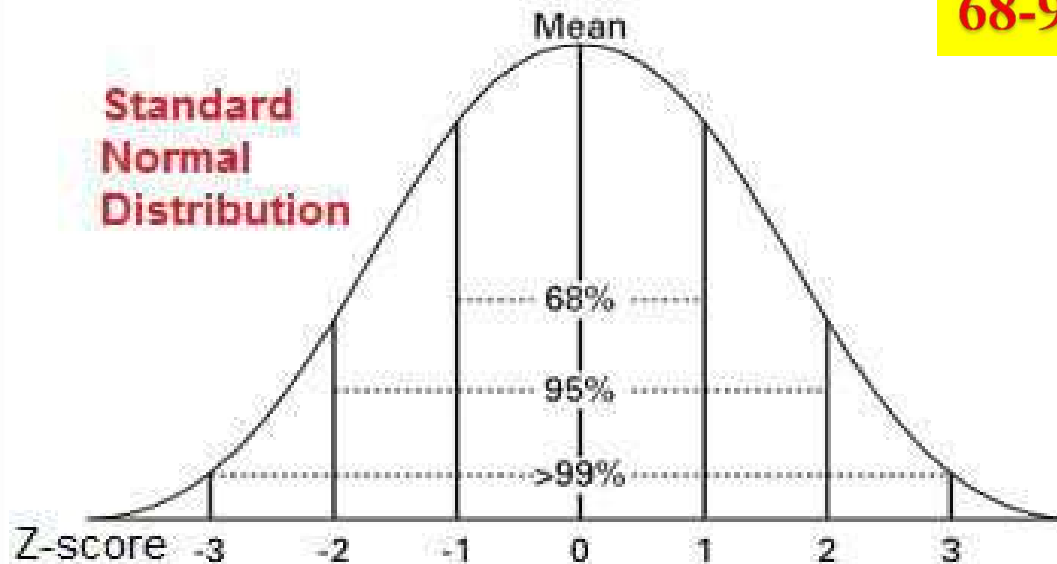


Roll No	Marks
1	-0.951587303
2	-0.751253134
3	0.651086049
4	1.051754387

Methods of Data Normalization...

The normal distribution is a probability function that describes how the values of a variable are distributed.

No matter what μ and σ are,
the area between $\mu - \sigma$ and $\mu + \sigma$ is about 68%;
the area between $\mu - 2\sigma$ and $\mu + 2\sigma$ is about 95%; and the
area between $\mu - 3\sigma$ and $\mu + 3\sigma$ is about 99.7%.
Almost all values fall within 3 standard deviations.



68-95-99.7 Rule, in Math terms...

$$\int_{\mu - \sigma}^{\mu + \sigma} \frac{1}{\sigma \sqrt{2\pi}} \cdot e^{-\frac{1}{2} \left(\frac{x - \mu}{\sigma} \right)^2} dx = .68$$

$$\int_{\mu - 2\sigma}^{\mu + 2\sigma} \frac{1}{\sigma \sqrt{2\pi}} \cdot e^{-\frac{1}{2} \left(\frac{x - \mu}{\sigma} \right)^2} dx = .95$$

$$\int_{\mu - 3\sigma}^{\mu + 3\sigma} \frac{1}{\sigma \sqrt{2\pi}} \cdot e^{-\frac{1}{2} \left(\frac{x - \mu}{\sigma} \right)^2} dx = .997$$

**THANK
YOU!**