



# End of Distribution Imputation

Subject: Data Mining

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# Definition

## End of Distribution Imputation

is equivalent to arbitrary value imputation, but it automatically selecting arbitrary values at the end of the variable distribution known as outlier.

If the variable is normally distributed, can use the mean plus or minus 3 times the standard deviation(SD)

If the variable is skewed, can use the IQR( Inter-Quantile Range ) proximity rule.

Which variable is fit for this method?

Suitable numerical variables

# How to use it?

## Skew Distributions

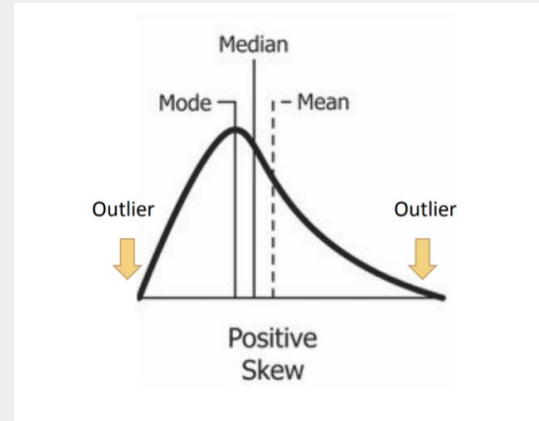
The general approach is to calculate the quantile, and then the inter-quantile range(IQR)

$IQR = 75^{th} \text{ Quantile} - 25^{th} \text{ Quantile}$

Upper limit =  $75^{th} \text{ Quantile} + IQR \times 1.5$

lower limit =  $25^{th} \text{ Quantile} - IQR \times 1.5$

Extreme outliers will time 3 instead of 1.5



# Code Implementation

# Advantages & Disadvantages

## Advantages

- Easy to implement

- Rapid way of obtaining complete dataset

- Can be integrated into production

- Capture the importance of “Missingness” if there is one

## Disadvantages

- Distortion of the original variable distribution

- Distortion of the original variance

- Distortion of the covariance with the remaining variables of dataset

- This technique may mask true outliers in the distribution

## References:

<https://medium.com/analytics-vidhya/feature-engineering-part-1-end-of-tail-imputation-c5069a41869a>

<https://www.kaggle.com/code/rushikeshlavate/end-of-distribution-imputation/notebook>