



Comparision between Accuracy and MSE, RMSE by using Proposed Method with Imputation Technique

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Abstract

Presence of missing values in the dataset leads to difficult for data analysis in data mining task. In this research work, student dataset is taken contains marks of four different subjects in engineering college. Mean, Mode, Median Imputation were used to deal with challenges of incomplete data. By using MSE and RMSE on dataset using with proposed Method and imputation methods like Mean, Mode, and Median Imputation on the dataset and found out to be values of Mean Squared Error and Root Mean Squared Error for the dataset. Accuracy also found out to be using Proposed Method with Imputation Technique. Experimental observation it was found that, MSE and RMSE gradually decreases when size of the databases is gradually increases by using proposed Method. Also MSE and RMSE gradually increase when size of the databases is gradually increases by using simple imputation technique. Accuracy is also increases with increases size of the databases.



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Incomplete data, Missing Values, Imputation, Mean Imputation, Median Imputation and Mode Imputation, MSE (Mean Squared Error), RMSE (Root Mean Squared Error) etc.

Introduction

Missing data imputation techniques can be used to improve the data quality. Missing data imputation techniques replace missing values of a dataset so that data analysis methods can be applied to complete dataset¹.

In this research work, student dataset is taken contains marks of four different subjects in engineering

college. Mean, Mode, Median Imputation were used to deal with challenges of incomplete data. By using MSE and RMSE on dataset using with proposed Method and imputation methods like Mean, Mode, and Median Imputation on the dataset and found out to be values of Mean Squared Error and Root Mean Squared Error for the dataset. Accuracy also found out to be using Proposed Method with Imputation Technique. Experimental observation it was found

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that, MSE and RMSE gradually decreases when size of the databases is gradually increases by using proposed Method. Also MSE and RMSE gradually increase when size of the databases is gradually increases by using simple imputation technique. Accuracy is also increases with increases size of the databases.

The organization of the paper is Section1: Introduction Section 2: Literature Reviews, Section 3: Dataset Used, Section 4: Methodology, Section 5: Experimental Result and Analysis, Section 6: Conclusions

Literature Reviews

lit wise Deletion

This method deletes those instances with missing value for data analysis in the dataset. It is the most common method, it has two drawbacks: a) A substantially decreases the size of dataset available for the data analysis. b) Data are not always missing completely at random.

Mean/Mode Imputation (MMI)

By replacing a missing values with the mean or mode of all attribute which consists missing value. To reduce the influence of exceptional data, median can also be used. This is one of the most commonly used methods.

K-Nearest Neighbor Imputation (KNN)

This method uses k-nearest neighbor algorithms to estimate and replace missing data. The main advantages of this method are a) it can estimate both qualitative attributes and quantitative attributes; b) It is not necessary to build a predictive model for each attribute with missing data².

Median Substitution

Median Substitution is calculated by grouping up of data and finding average for the data. Median can be calculated by using the formula

$$\text{Median} = L + h/f (n/2-c) \quad \dots(1)$$

where L is the lower class boundary of median class h is the size of median class i.e. difference between

upper and lower class boundaries of median class f is the frequency of median class, c is previous cumulative frequency of the median class, n/2 is total no. of observations divided by 2

Standard Deviation

The standard deviation measures the spread of the data about the mean value. It is useful in comparing sets of data which may have the same mean but a different range. The Standard Deviation is given by the formula:

$$S_N = \sqrt{\frac{1}{N} \sum_{i=1}^n (x_i - \bar{x})^2} \quad \dots(2)$$

Where{X₁,X₂,...,X_n} are the observed values of the sample items and is the mean value of these observations, while the denominator N stands for the size of the sample⁷.

Dataset Used

In this work dataset having characteristics is given below.

Number of Instances: 5000,10,000,15,000,20,000
 Number of Attributes: 05
 (Record No., M1, ECE, EM, EE)

Dataset contains marks of four different subjects of engineering college. In dataset randomly distributed the missing values in each attribute to become the incomplete dataset. Record. No. in the Dataset is used are imaginary and generated for the data analysis purpose in data mining process. In dataset M1, ECE, EM, EE are the subject in engineering college and class test marks for each subject is out of twenty marks for each subject repectively.The structure of Dataset as shown in the Table No.1

Table No 1: Dataset

Record No.	Subject 1	Subject 2	Subject 3	Subject 4
01	X1	X2	X3	X4
..
N	XN	XN	XN	XN

Methodology

To found out accuracy by using proposed method with imputation technique like mean, mode and median for five thousand dataset, ten thousand dataset, fifteen thousand dataset and twenty thousand dataset.

To found out MSE (Mean Squared Error) for proposed method with mean imputation technique and simple mean imputation technique by using following equation:

$$MSE = \frac{SSE}{n} \quad \dots(1)$$

Where SSE = Sum of Squared Error N = No of Sample

To found out MSE (Mean Squared Error) for proposed method with mode imputation technique and simple mode imputation technique by using equation no.1.

To found out MSE (Mean Squared Error) for proposed method with median imputation technique and simple median imputation technique by using equation no.1.

To found out RMSE (Root Mean Squared Error) for proposed method with mean imputation technique and simple mean imputation technique by using following equation

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^n (x_i - \hat{x}_i)^2} \quad \dots(2)$$

Where \hat{x}_i = predicted value x_i = Observed value, N = No of Sample

To found out RMSE (Root Mean Squared Error) for proposed method with mode imputation technique and simple mode imputation technique by using equation no.2

To found out RMSE (Root Mean Squared Error) for proposed method with median imputation technique

and simple median imputation technique by using equation no.2

Experimental Result and Analysis

For Experimental Result Student dataset is taken which contains marks of four different subjects of engineering college. Mean, Mode, Median Imputation were used to deal with challenges of incomplete data.

Using proposed method with imputation techniques like Mean, Mode, and Median Imputation on the student dataset and found out to be accuracy for five thousand, ten thousand, fifteen thousand and twenty thousand dataset respectively. Accuracy increase with increase in size of the dataset.

Similarly found out to be MSE by using Proposed Method with Imputation Technique like Mean, Mode, and Median Imputation. MSE decreases with increase in size of the dataset.

Found out to be MSE by using simple Imputation Technique like Mean, Mode, and Median Imputation. MSE increase with increase in size of the dataset.

Similarly found out to be RMSE by using Proposed Method with Imputation Technique like Mean, Mode, and Median Imputation. RMSE decreases with increase in size of the dataset.

Also found out to be RMSE by using simple Imputation Technique like Mean, Mode, and Median Imputation. RMSE increase with increase in size of the dataset.

Accuracy found out to be by proposed method with Imputation Technique like Mean, Mode, and Median Imputation. Also MSE by using Proposed Method with Imputation Technique and MSE by using simple Imputation Technique, RMSE by using Proposed Method with Imputation Technique, RMSE by using simple Imputation Technique. The result is shown in the Table no.2 and Table no.3

Table 2: Comparison of Accuracy and MSE

Dataset	Accuracy by Proposed Method with Mean Imputation	MSE For Proposed Method	MSE For Simple Mean Imputation with Mode Imputation	Accuracy by Proposed Method with Median Imputation	MSE For Proposed Method	MSE For Simple Mode Imputation	Accuracy by Proposed Method with Median Imputation	MSE For Proposed Method	MSE For Simple Median Imputation
Five Thousand	87.72	3.04	3.41	84.00	6.41	7.10	86.99	2.49	4.17
Ten Thousand	88.37	2.35	8.11	86.33	3.65	7.51	87.15	2.10	8.43
Fifteen Thousand	89.88	1.93	8.90	89.34	2.39	7.62	90.45	1.90	8.74
Twenty Thousand	91.25	1.65	9.78	90.80	1.96	9.91	91.34	1.64	9.07

Table 3: Comparison of Accuracy and RMSE

Dataset	Accuracy by Proposed Method with Mean Imputation	RMSE For Proposed Method	RMSE For Simple Mean Imputation with Mode Imputation	Accuracy by Proposed Method with Median Imputation	RMSE For Proposed Method	RMSE For Simple Mode Imputation	Accuracy by Proposed Method with Median Imputation	RMSE For Proposed Method	RMSE For Simple Median Imputation
Five Thousand	87.72	3.04	1.84	84.00	2.53	3.01	86.99	1.879	2.04
Ten Thousand	88.37	1.53	2.84	86.33	1.91	3.74	87.15	1.57	2.50
Fifteen Thousand	89.88	1.38	2.92	89.34	1.54	3.95	90.45	1.38	2.72
Twenty Thousand	91.25	1.28	3.12	90.80	1.40	3.14	91.34	1.28	3.01

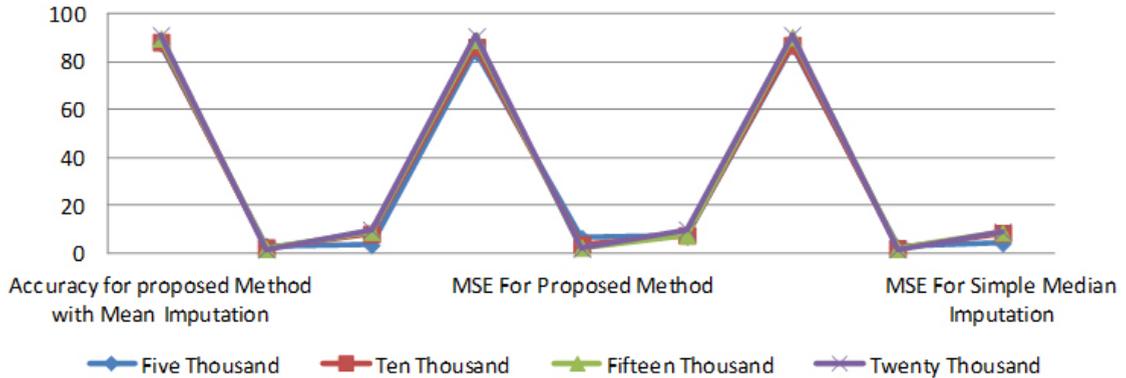


Fig. 1: Graphical Representation of Accuracy and MSE

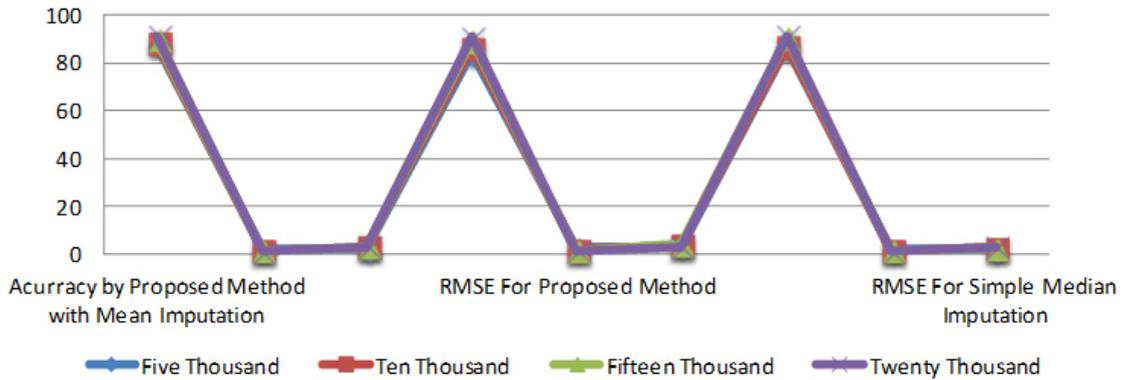


Fig. 2: Graphical Representations of Accuracy and RMSE

Conclusions

In this research paper by using proposed method with imputation technique like Mean, Mode and Median Imputation on the student dataset and found out to be accuracy. Also found out to be MSE and RMSE on dataset using with proposed Method and imputation methods like Mean, Mode, and Median Imputation on the dataset. Experimental observation it is found that, MSE and RMSE gradually decreases when size of the databases is gradually increases by using proposed Method. Also MSE and RMSE

gradually increase when size of the databases is gradually increases by using simple imputation technique. Accuracy increase with increase in size of the dataset.

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