Feature Selection for Machine Learning

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Introduction

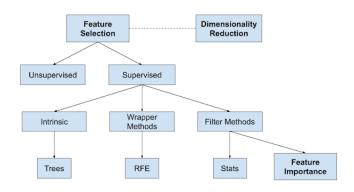


Figure 1: Overview of feature selection techniques



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- Language and Libraries:
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 - Scikit-Learn
 - Xgboost
- Feature selection Methods:
 - Random Forest
 - Recursive Feature Elimination
 - Xgboost



	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583	NaN	Q
6	7	0	1	McCarthy, Mr. Timothy J	male	54.0	0	0	17463	51.8625	E46	S
7	8	0	3	Palsson, Master. Gosta Leonard	male	2.0	3	1	349909	21.0750	NaN	S
8	9	1	3	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	female	27.0	0	2	347742	11.1333	NaN	S
9	10	1	2	Nasser, Mrs. Nicholas (Adele Achem)	female	14.0	1	0	237736	30.0708	NaN	С
10	11	1	3	Sandstrom, Miss. Marguerite Rut	female	4.0	1	1	PP 9549	16.7000	G6	S

Figure 2: Overview of dataset



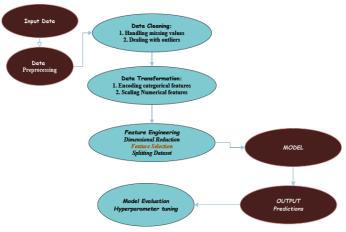


Figure 3: Machine Learning Workflow



Results and Discussions

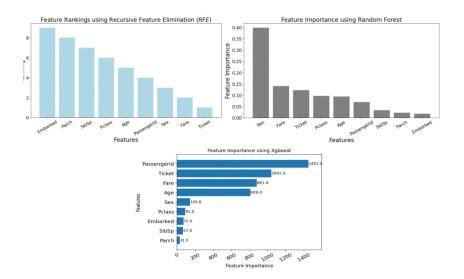


Figure 4: Comparing the three feature selection methods used

Results and Discussions

Feature Selection Method	Features	Accuracy	
Without Feature Selection	Passengerld, Pclass, Sex, Age, SibSp, Parch, Ticket, Fare, Embarked	74.30%	
RFE Feature Selection	Embarked, Parch, SibSp, Pclass, Age, Passengerld, Sex	79.89%	
Random Forest	Sex, Fare, Ticket, Pclass, Age, Passengerld, SibSp	80.44%	
Xgboost	Passengerld, Fare, Age, SibSp, Pclass, 'Sex', Ticket	80.44%	

Table 1: Different feature selection methods and accuracy.



 In this study, we reviewed the different feature selection methods and investigated the impact of three methods (random forests, recursive feature elimination, and Xgboost) on the accuracy of the model.



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- Using Feature selection increased the accuracy of the model in making predictions.
- Although the different methods gave different feature rankings, they all gave a higher accuracy than the model built without any feature selection.
- The random forest and Xgboost methods give the highest accuracy for this dataset.



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