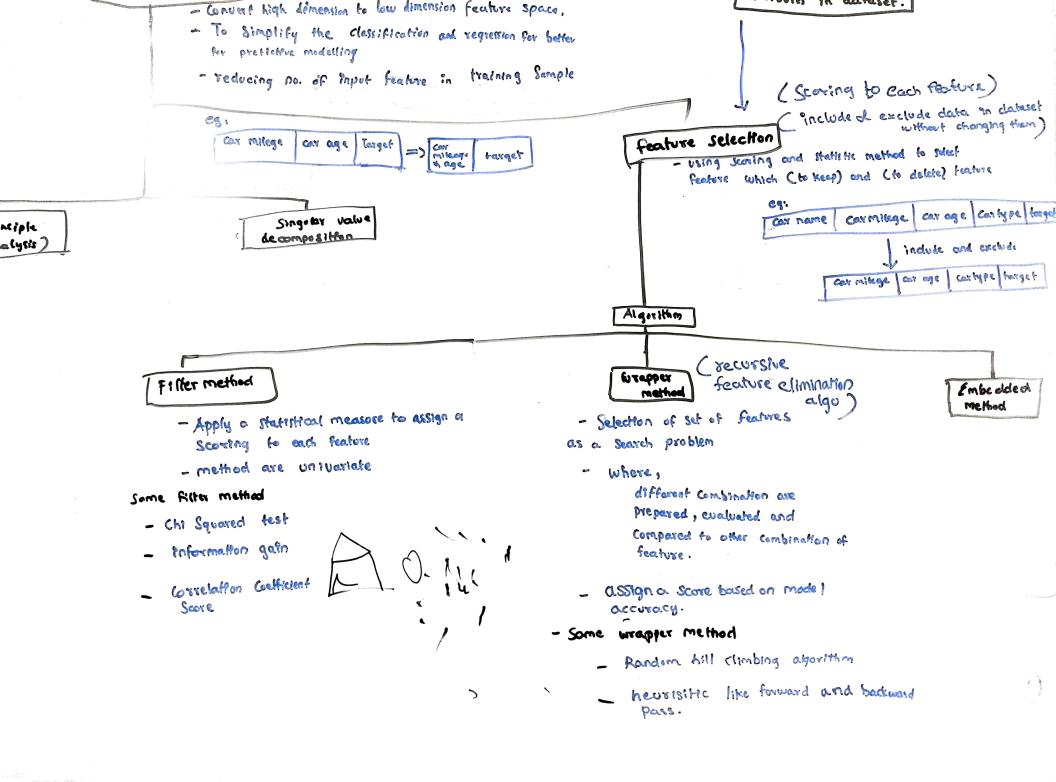


fter, (new | Combination of attributes) Different from each other but a cleaning, D; mensionality Common in to freduce the no. of ta preparation seduction attributes in dataset. - Convert high dimension to low dimension feature space. - To Simplify the classification and regression for better for prefictive modelling Feature Selection (Include of exclude data in dataset - reducing no. of input feature in training Sample Cg. Can milege can age Target - using scoring and statistic method to solut rarget Feature which (to keep) and (to datate) feature Ton name | cormittege | car age | Car by Me. for Singolar value PCA Covinciple de composition I include and eachdi component analysis) Car mitega | car age | cartype thorget Algerithm E Lobber Piller method Embe delect method PRE Photo



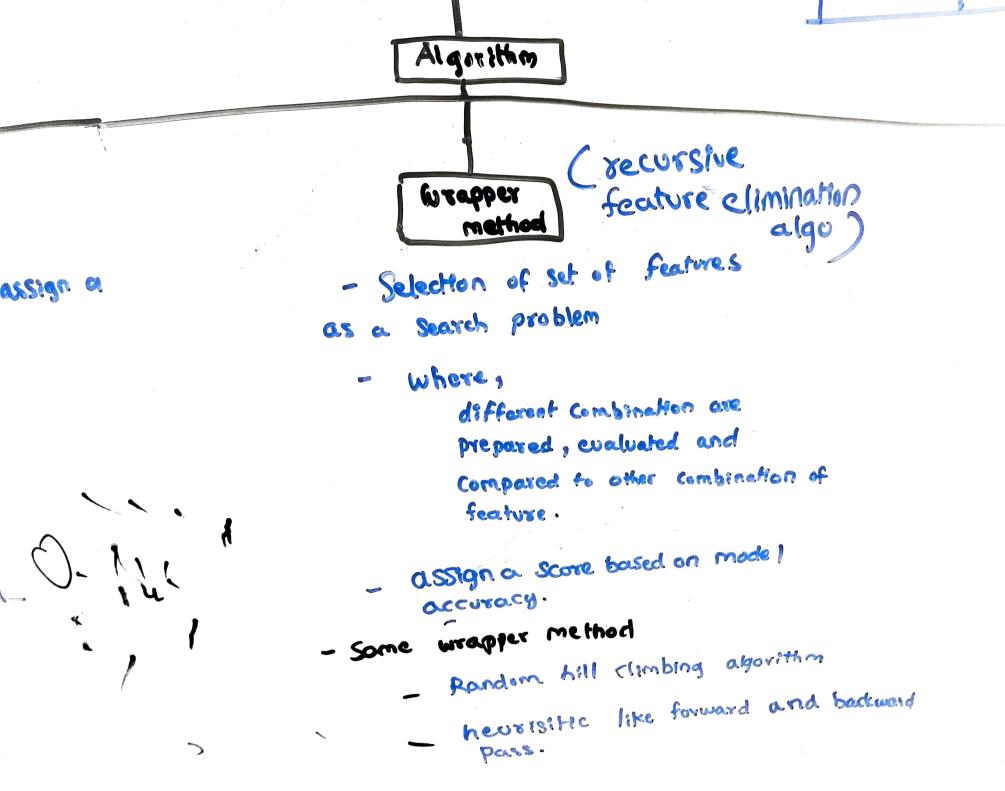
Filter method

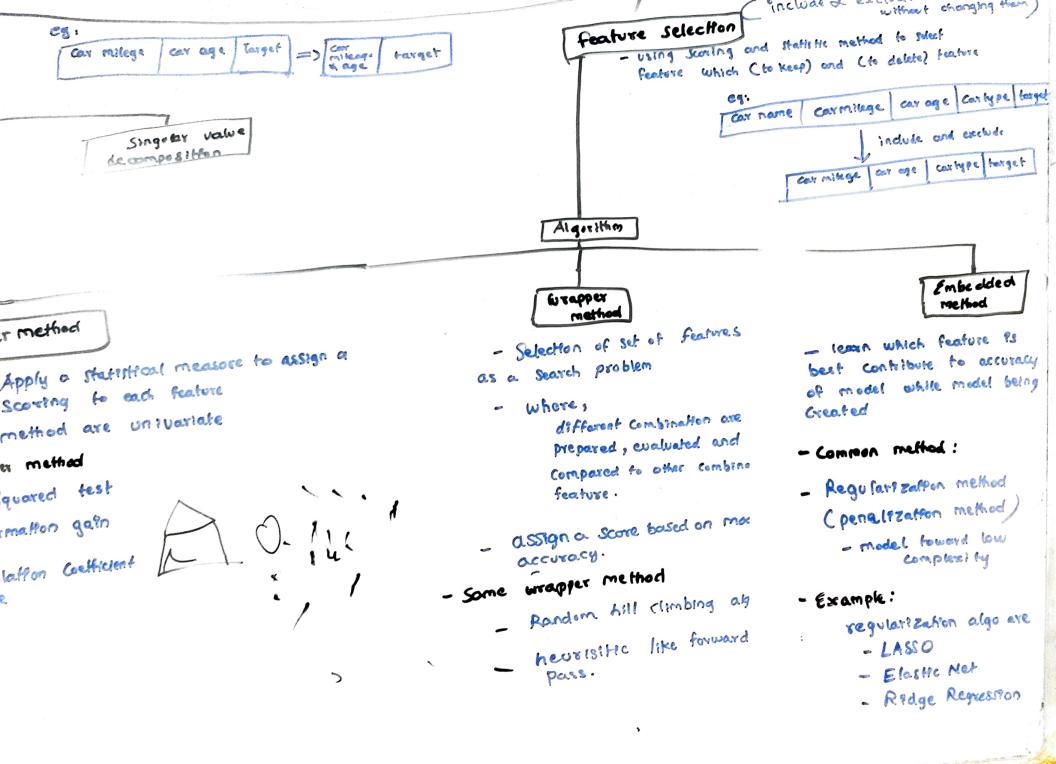
- Apply a statistical measure to assign a Scoring to each feature
- method are univariate

Some Riller method

- Chi Squared test
- Information gain
- Correlation Coefficient Score







Embe dded method

best Contribute to accuracy

of model while model being Greated

- Common method:

- Regularization method)

- model toward low complexify

- Example:

regular? Zation algo are

- LASSO
- Elastic Net
- Rødge Regression

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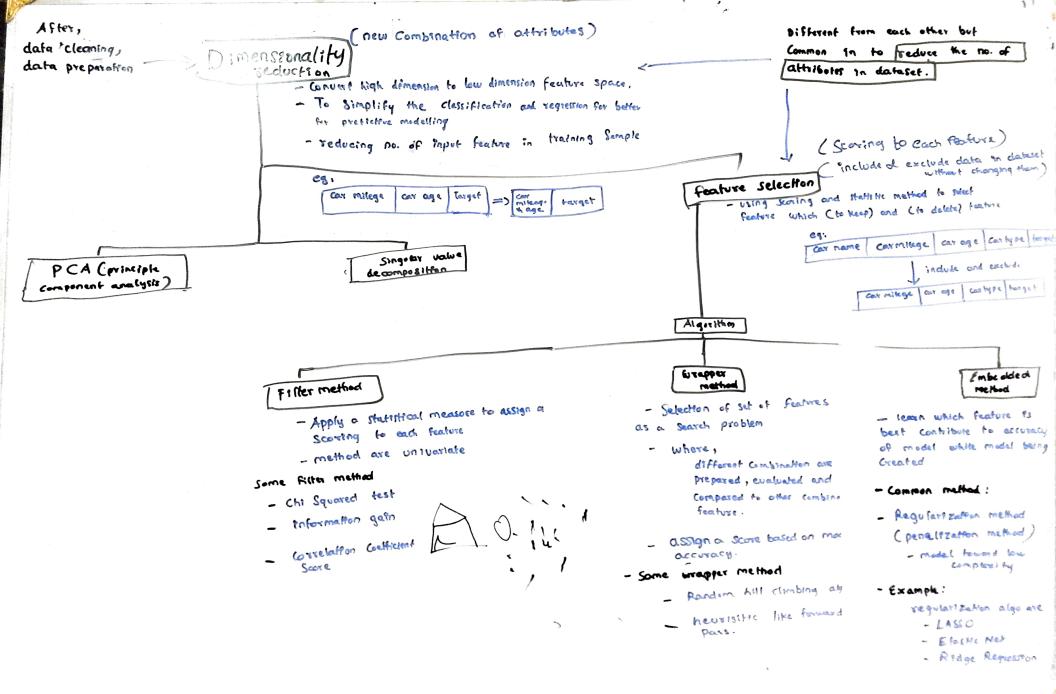
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Feature Selection

Feature selection is a process where you automatically select those features in your data that contribute most to the prediction variable or output in which you are interested.

Having irrelevant features in your data can decrease the accuracy of many models, especially linear algorithms like linear and logistic regression.

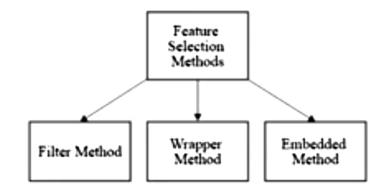
Three benefits of performing feature selection before modeling your data are:

- Reduces Overfitting: Less redundant data means less opportunity to make decisions based on noise.
- Improves Accuracy: Less misleading data means modeling accuracy improves.
- Reduces Training Time: Less data means that algorithms train faster.

You can learn more about feature selection with scikit-learn in the article Feature selection.

Feature Selection Methods

- Filter Methods
- Wrapper Methods
- Embedded Methods







Filter Methods

- Single Factor Analysis
- Uses Statistical algorithms like Chi-Square | Fisher Score | Anova |
 Mutual Information | Variance
- Uses Individual Feature Predictive Power
- Less computationally expensive
- Uses correlation power on feature to target space
- Useful in quick screening



Wrapper Methods

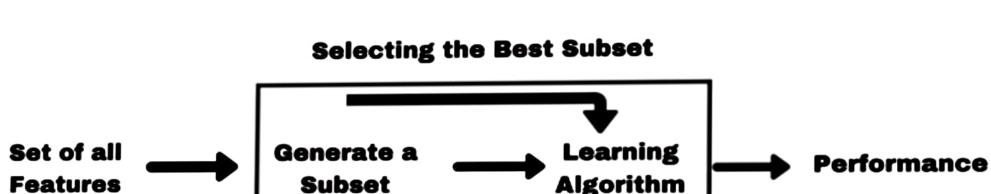
- Use combinations of variables to determine predictive power
- Find the best combination of variables
- Computationally expensive than filter method
- Perform better than filter method
- Not recommended on high number of features



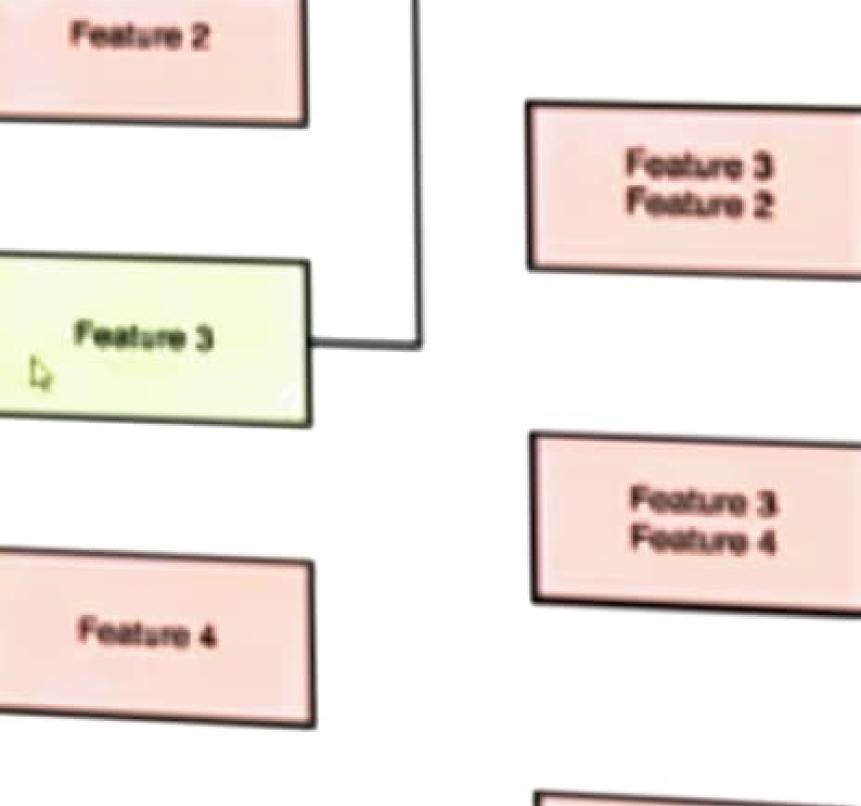
Forward Step Selection

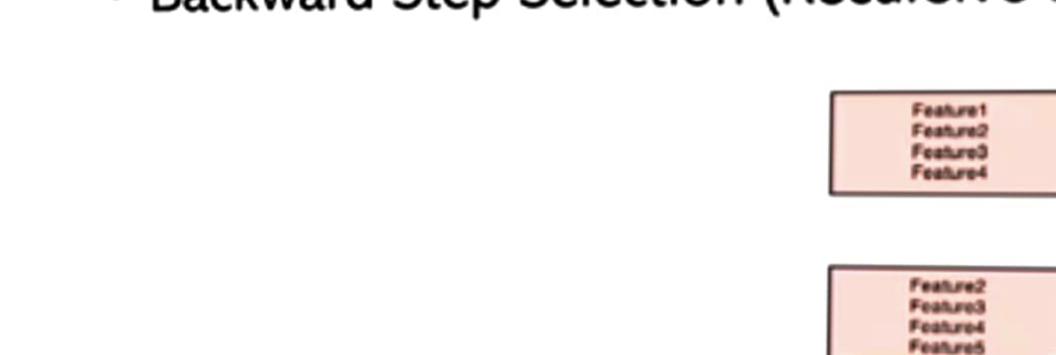
- Subset Selection (Exhaustive Feature)

Backward Step Selection (Recursive Feature Selection)



KGP Talkie





- fits the model with each possible combinations of N features.
- requires massive computational power
- Y = BO V = BO + Batty + Y = Co + Catty
 - (Y = BO, Y = BO + B1*X1, Y = CO + C1*X2, Y = DO + D1*X1 + D2*X2)
- Use test error to evaluate model newface.

Embedded Methods

- Faster than wrapper methods
- More accurate than filter method
- Perform feature selection during algorithm implementation thus embedded in it
- LASSO Regression
- RIDGE Regression
- Tree Importance

