**Fundamentals of Machine Learning**

**Report**

**Predicting the Price of a Football Player**

A Course Work Report Submitted in Partial Fulfillment of the Requirements

for the Degree of

**Bachelor of Technology**

**In**

**Computer Science and Engineering Department**

By

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**CSE I**



SCHOOL OF ENGINEERING AND TECHNOLOGY

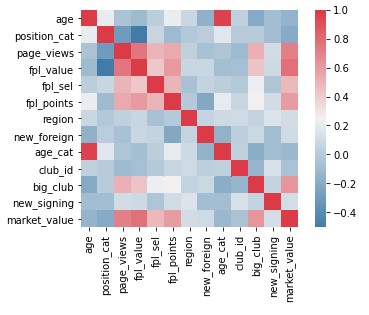
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1. **Pearson’s Correlation between each variable and output variable:**

|  |  |  |
| --- | --- | --- |
|  | position\_cat -0.202518 |  |
|  | age -0.144592 |  |
|  | age\_cat -0.116853 |  |
|  | club\_id -0.052287 |  |
|  | new\_foreign 0.097173 |  |
|  | region 0.110158 |  |
|  | new\_signing 0.115376 |  |
|  | fpl\_sel 0.484164 |  |
|  | fpl\_points 0.595919 |  |
|  | big\_club 0.624354 |  |
|  | page\_views 0.716096 |  |
|  | fpl\_value 0.771985 |  |
|  | market\_value 1.000000 |  |

The Correlation between each variable in data set with market value is found. The Correlation value indicates the dependence of the variables on each other. The correlation value lies between 0 and 1 (-1 to 1 if the variables are dependent on each other inversely). Higher the value of correlation, the higher dependency and the variables can be merged or removed according to the value of correlation. The Heat map below shows the interdependency of each variable on other. If the intensity of color is more, then it can be assumed that the variables are more correlated. Dark red represents higher dependency on each other positively and dark blue represents higher dependency on each other negatively**.**

**Heat Map showing correlation among all the variables:**



1. **Comparison of Different Regression Models:**

# Linear Regression

MSE score: 2253.1826765864375

R2 score: -11.382032075377651

# Lasso Regression

MSE score: 44.635748023994644

R2 score: -9.948664696519803

# Ridge Regression

MSE score: 47.42696824958404

R2 score: -11.360790565498204

**Kth Nearest Neighbour Regression**

RMSE value for k= 20 is: 8.774911812480529

# SVR Regression

MSE score: 10.47592874913882

R2 score: 0.39691213061159725

**Tree Regression:**

MSE score: 9.841880295889636

R2 score: 0.46770586839901285

# Random Forest

MSE score: 6.158698043109998

R2 score: 0.7915637053996277

# Gradient Boost Regression

MSE score: 6.180443705944269

R2 score: 0.790089177057778

# Tune the hyperparameters and build the most accurate model Linear Regression

Best Score: -4.319281419264125

Best Hyperparameters: {'copy\_X': True, 'fit\_intercept': True, 'normalize': True}

LinearRegression GridSearch Accuracy: -11.382032075377651

RMSE score: 54.74359131021873

R2 score: -15.468805129349892

# Lasso Regression

Best Score: -4.240566475479193

Best Hyperparameters: {'alpha': 0.01, 'copy\_X': False, 'fit\_intercept': True, 'max\_iter': 10,

'normalize': True, 'precompute': True, 'selection': 'random', 'warm\_start': True}

Lasso GridSearch Accuracy: -10.808741351789669

RMSE score: 44.991174613855385

R2 score: -10.123723485983499

# Ridge Regression

Best Score: -4.227049326675834

Best Hyperparameters: {'alpha': 0.1, 'copy\_X': True, 'fit\_intercept': True, 'max\_iter': 10,

'normalize': True, 'solver': 'sag'}

Ridge GridSearch Accuracy: -16.53391516942623

RMSE score: 7.3203682790345175 R2 score: 0.7055162179805238

# Kth Nearest Neighbour Regression

Best Score: 0.5200726289993008

Best Hyperparameters: {'leaf\_size': 1, 'n\_neighbors': 24, 'p': 1}

Knn GridSearch Score: 0.5680141582840241

RMSE value : 8.622021473671152

R2 score: 0.5914796253878369

# SVR Regression

Best Score: -6.099489733412793

Best Hyperparameters: {'gamma': 0.0001, 'kernel': 'rbf', 'max\_iter': 100}

SVR GridSearch Accuracy: 0.2185178234939857

MSE score: 10.47592874913882

R2 score: 0.39691213061159725

# Decision Trees

Best Score: -4.537185920582436

Best Hyperparameters: {'criterion': 'friedman\_mse', 'max\_depth': 5, 'max\_features': None,

'min\_samples\_leaf': 1, 'min\_samples\_split': 2, 'splitter': 'best'}

RMSE score: 7.11986429441369

R2 score: 0.7214270436370289

# Random Forest

Best Score: 0.7774042844927579

Best Hyperparameters: {'n\_estimators': 1600, 'min\_samples\_split': 2, 'min\_samples\_leaf': 1,

'max\_features': 'log2', 'max\_depth': 1000}

Random Forest GridSearch Accuracy: 0.7597747041311681

RMSE score: 6.679054568133203

R2 score: 0.75485360871305

**Gradient Boost Regression**

Best Score: -3.45074930135845

Best Hyperparameters: {'subsample': 1, 'n\_estimators': 500, 'max\_features': 'log2', 'max\_depth': 4, 'loss': 'lad', 'alpha': 0.1}

RMSE score: 6.906251422901736

R2 score: 0.7378919913863964

|  |  |  |  |
| --- | --- | --- | --- |
|  | **Before Tuning(R2)** | **After Tuning(R2)** | **Best Scores** |
| **Linear regression** | -11.382032075377651 | -15.468805129349892 | - |
| **Lasso Regression** | -9.948664696519803 | -10.123723485983499 |  |
| **Ridge Regression** | -11.360790565498204 | 0.7055162179805238 | - |
| **Near Neighbor**  **Regression** | 0.685247698439583 | 0.5914796253878369 | 0.5200726289993008 |
| **Support vector Machine** | 0.39691213061159725 | 0.39691213061159725 | - |
| **Decision Trees** | 0.46770586839901285 | 0.7214270436370289 |  |
| **Random Forest** | 0.7915637053996277 | 0.75485360871305 | 0.7774042844927579 |
| **Gradient Boost**  **Regression** | 0.790089177057778 | 0.7378919913863964 | - |

From all the above models we can observe that the **SUPPORT VECTOR REGRESSION,** reports us the best results as the R2 value of the SVR is lowest among all other regressions. The Best Score of SVR is -6.099489733412793, Best Hyperparameters: {'gamma': 0.0001, 'kernel': 'rbf', 'max\_iter': 100}, Accuracy: 0.2185178234939857, MSE score:

10.47592874913882 and R2 score: 0.39691213061159725

So, Support vector Machine is the best algorithm among all others in this scenario. So, we deployed Support Vector Machine Regression in Restful API.

# Model Deployment as RESTFUL API service



Attaching ipynb file containing all the tests and api model.