

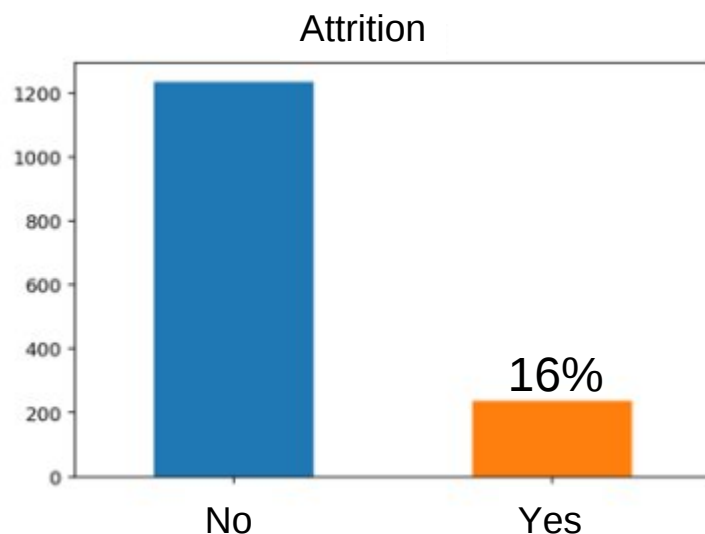


# EMPLOYEE ATTRITION AND HUMAN RESOURCE (HR) ANALYTICS

BY SHIVANSH TRIPATHI

# Dataset

There are 1,470 observations with 34 characteristics of employees (Features) and 1 target variable (attrition 0 for "no" or 1 for "yes")

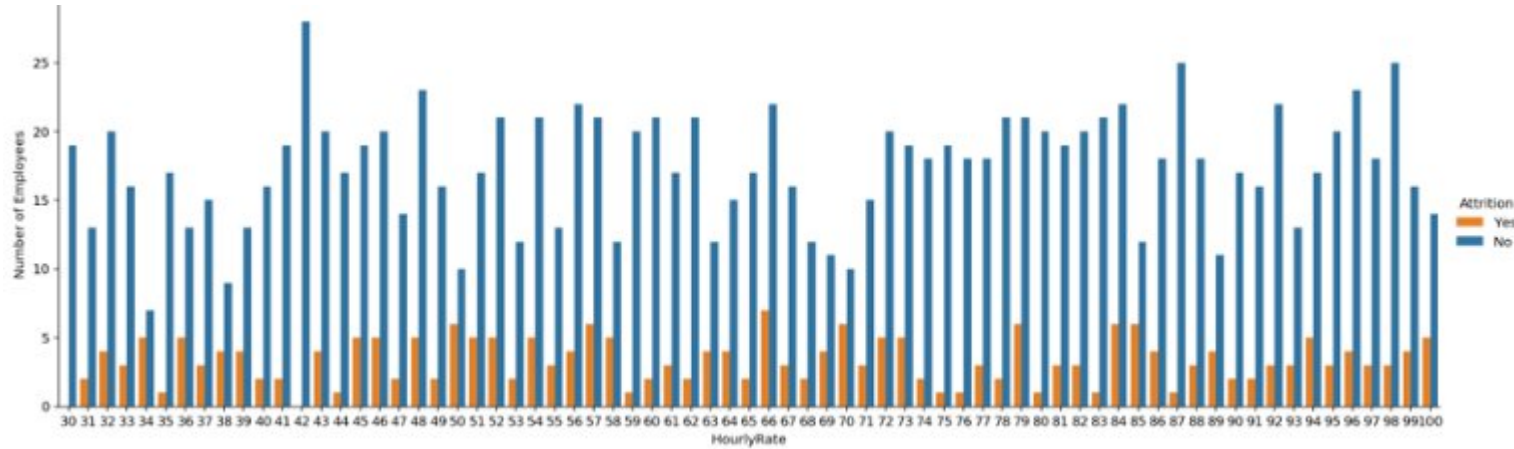


Features: ['Age', 'BusinessTravel', 'DailyRate', 'Department', 'DistanceFromHome', 'Education', 'EducationField', 'EmployeeCount', 'EmployeeNumber', 'EnvironmentSatisfaction', 'Gender', 'HourlyRate', 'JobInvolvement', 'JobLevel', 'JobRole', 'JobSatisfaction', 'MaritalStatus', 'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWorked', 'Over18', 'OverTime', 'PercentSalaryHike', 'PerformanceRating', 'RelationshipSatisfaction', 'StandardHours', 'StockOptionLevel', 'TotalWorkingYears', 'TrainingTimesLastYear', 'WorkLifeBalance', 'YearsAtCompany', 'YearsInCurrentRole', 'YearsSinceLastPromotion', 'YearsWithCurrManager']

# Exploratory Data Analysis (EDA)

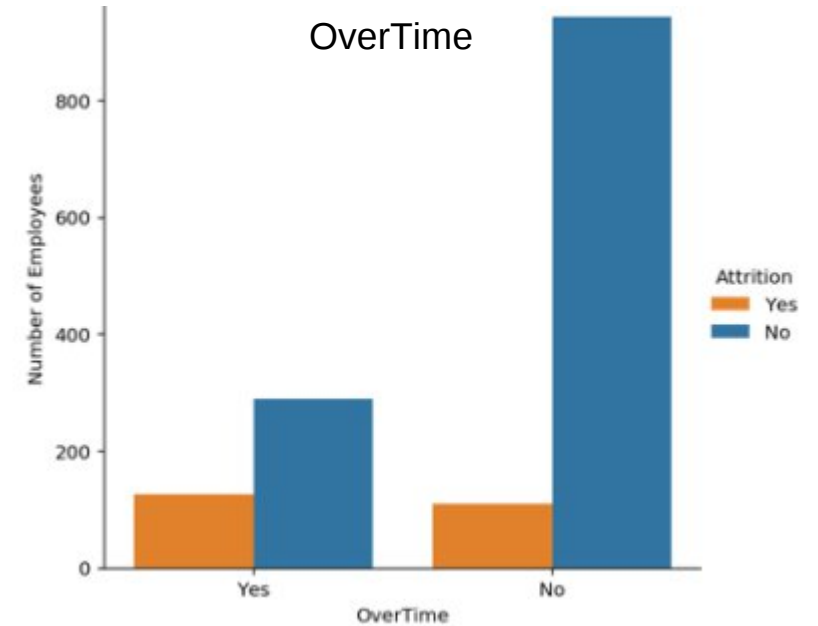
Attrition seems to happen at every level regardless of employee hourly rate

HourlyRate



Overtime could be a key factor leading to attrition

OverTime



# Iteration 1: Baseline Model Performance

Below 10 models have been evaluated:

- Gaussian Naive Bayes
- Bernoulli Naive Bayes
- Multinomial Naive Bayes
- Logistic Regression
- K Nearest Neighbour
- Decision Tree Classifier
- Random Forest Classifier
- Extreme Gradient Boost (XGBoost)
- Support Vector Classification (SVC)
- Linear SVC

The performance metrics used in the evaluation are:

- [Accuracy Score](#): proportion of correct predictions out of the whole dataset. Be careful when the target class is imbalance, for example, if a model predicts all flight passengers as non-terrorist, then the useless model would be 99.99% accurate.
- [Precision Score](#): proportion of correct predictions out of all predicted attrition cases.
- [Recall Score](#): proportion of correct predictions out of all actual attrition cases.
- [F1 Score](#): optimised balance between Precision and Recall for the selected relevant target.

# Iteration 1: Baseline Model Performance

Baseline model performance

	model	accuracy	acc(test)	precision	recall	f1score	rocauc	logloss
0	GaussianNB	0.778979	0.721088	0.399655	0.631579	0.484434	0.755946	9.633433
1	BernoulliNB	0.829917	0.792517	0.460915	0.263158	0.334298	0.742725	7.166274
2	MultinomialNB	0.548442	0.489796	0.200183	0.600000	0.300132	0.589342	17.622184
3	LogisticRegression	0.857144	0.877551	0.698690	0.210526	0.317851	0.791593	4.229238
4	KNearestNeighbour	0.823985	0.819728	0.342424	0.105263	0.160358	0.594289	6.226405
5	DecisionTree	0.784010	0.768707	0.306718	0.315789	0.318768	0.590567	7.988656
6	RandomForest	0.857144	0.846939	0.661111	0.168421	0.248970	0.733590	5.286566
7	XGBoost	0.866498	0.853741	0.711310	0.300000	0.420763	0.792229	5.051601
8	SVC	0.838435	0.840136	0.000000	0.000000	0.000000	0.500000	5.521505
9	LinearSVC	0.838435	0.840136	0.078247	0.278947	0.021818	0.554647	5.521505



After tuning hyperparameters and threshold

	model	accuracy	acc(test)	precision	recall	f1score	rocauc	logloss
0	GaussianNB	0.789966	0.721088	0.512195	0.446809	0.477273	0.732277	0.691588
1	BernoulliNB	0.836735	0.792517	0.279070	0.765957	0.409091	0.713670	0.471596
2	MultinomialNB	0.544218	0.489796	0.181250	0.617021	0.280193	0.556809	17.407283
3	LogisticRegression	0.861395	0.877551	0.522727	0.489362	0.505495	0.772676	0.364740
4	KNearestNeighbour	1.000000	0.744898	0.208333	0.212766	0.210526	0.529460	8.811016
5	DecisionTree	0.961735	0.802721	0.300971	0.659574	0.413333	0.657163	4.023447
6	RandomForest	1.000000	0.819728	0.402597	0.659574	0.500000	0.770308	0.376679
7	XGBoost	0.943027	0.853741	0.418919	0.659574	0.512397	0.806960	0.352849
8	SVC	0.954932	0.843537	0.159864	1.000000	0.275660	0.527522	0.435223
9	LinearSVC	0.198129	0.190476	0.157706	0.936170	0.269939	0.492377	27.960601

# Iteration 2: Feature Engineering, Feature Selection

## 'EducationField'

Life Sciences	606
Medical	464
Marketing	159
Technical Degree	132
Other	82
Human Resources	27



## 'JobRole'

Sales :	
Sales Executive	326
Sales Representative	83
Manager	37
Research & Development :	
Research Scientist	292
Laboratory Technician	259
Manufacturing Director	145
Healthcare Representative	131
Research Director	80
Manager	54
Human Resources :	
Human Resources	52
Manager	11



## New feature 'EduField\_Dept'

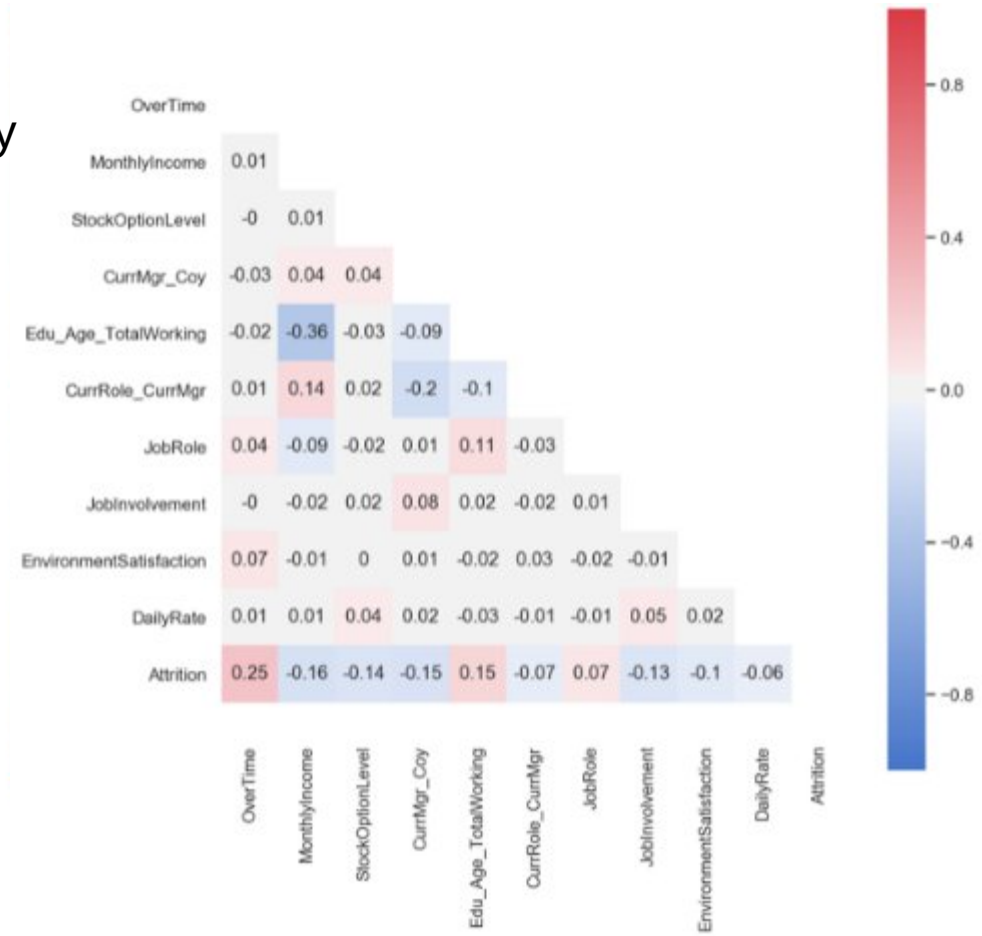
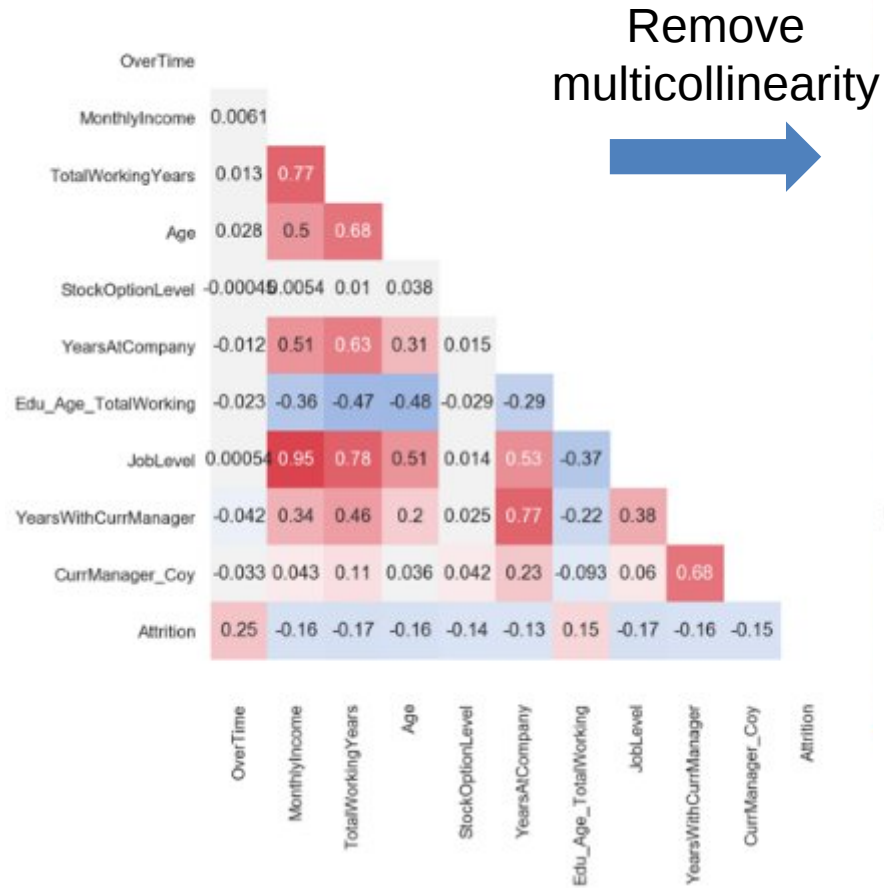
whether JobRole is related to EducationField:

- 0 = not related,
- 1 = related,
- 2 = somewhat related

Other new features created:

- **Job\_Coy** =  $\text{JobLevel} / (\text{YearsAtCompany} + 1)$
- **Edu\_Age\_TotalWorking** =  $\text{Education} / (\text{Age} + \text{TotalWorkingYears})$
- **CurrMgr\_Coy** =  $\text{YearsWithCurrManager} / (\text{YearsAtCompany} + 1)$
- **CurrRole\_CurrMgr** =  $\text{YearsInCurrentRole} / (\text{YearsWithCurrManager} + 1)$

# Iteration 2: Feature Engineering, Feature Selection



# Iteration 3: Over-sampling with SMOTE

(Original) Baseline X\_train.shape: (1176, 29)

0 986

1 190

Name: Attrition, dtype: int64

Model accuracy is 0.8333333333333334

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	precision	recall	f1-score	support
0	0.85	0.98	0.91	247
1	0.40	0.09	0.14	47
micro avg	0.83	0.83	0.83	294
macro avg	0.62	0.53	0.52	294
weighted avg	0.78	0.83	0.79	294

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[[241  6]
 [ 43  4]]
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After SMOTE over-sampling X\_train\_sm.shape: (1972, 29)

1 986

0 986

dtype: int64

Model accuracy is 0.8639455782312925

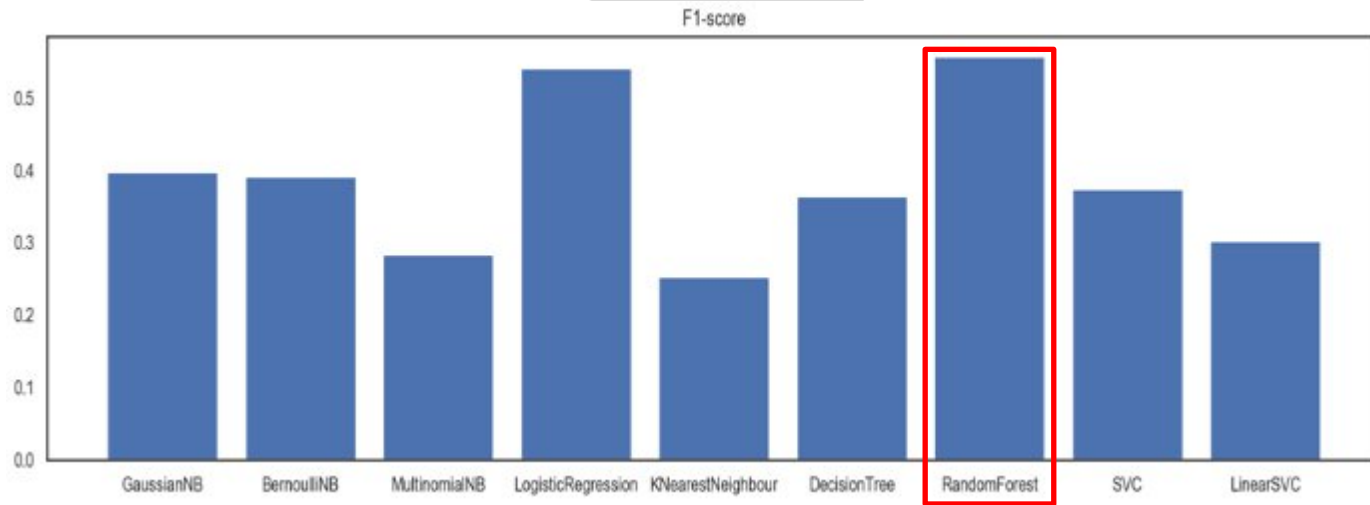
Model accuracy is 0.8639455782312925

	precision	recall	f1-score	support
0	0.87	0.98	0.92	247
1	0.73	0.23	0.35	47
micro avg	0.86	0.86	0.86	294
macro avg	0.80	0.61	0.64	294
weighted avg	0.85	0.86	0.83	294

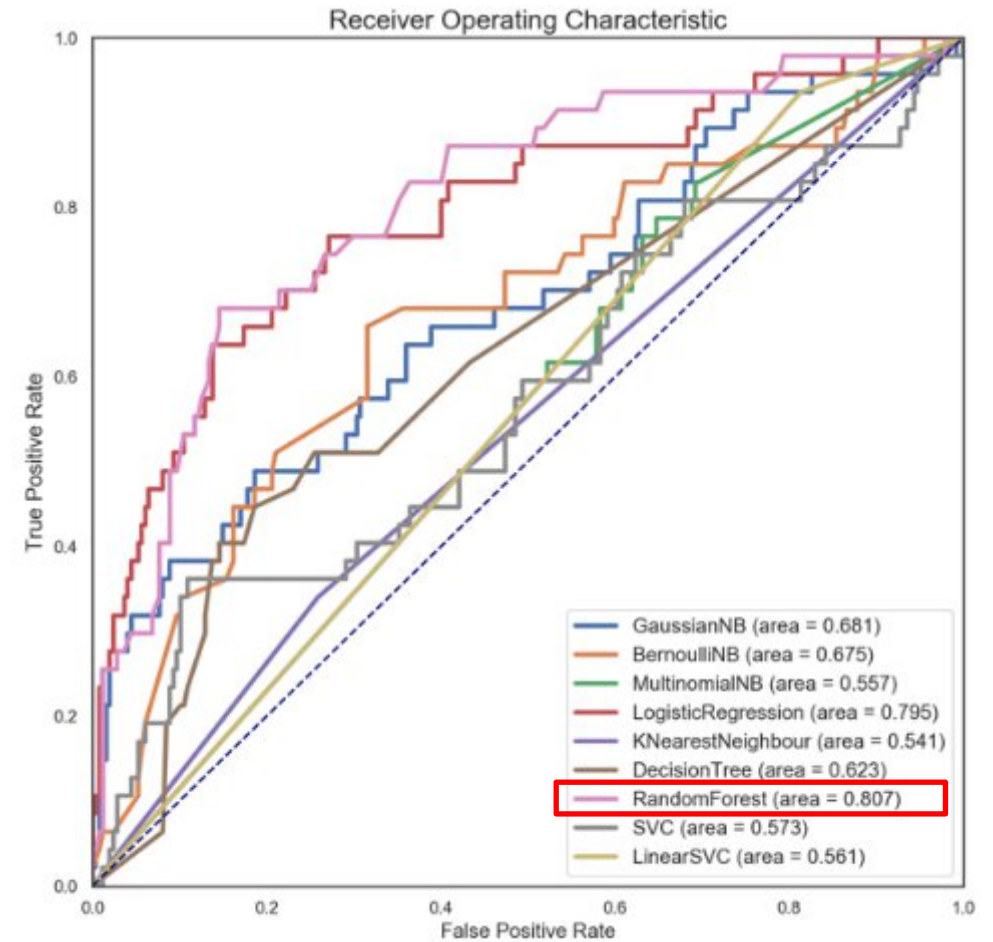
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# Final Model Performance (including tuning)



Best performing model:  
Random Forest Classifier



# Key Findings – Employee Attrition Factors

## Factor 1: **monetary**

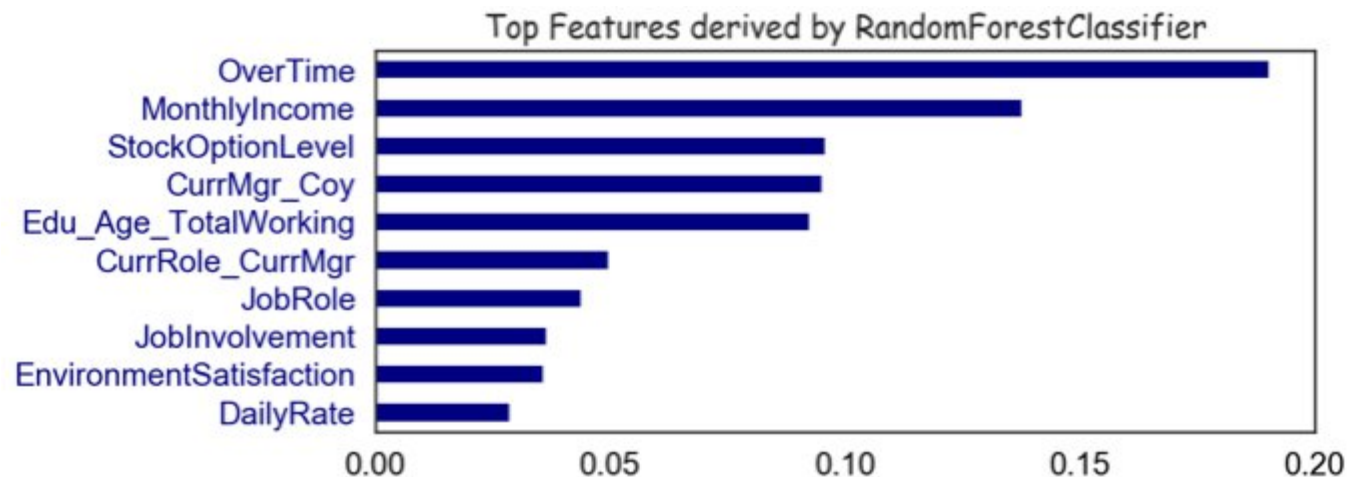
- 'OverTime'
- 'MonthlyIncome'
- 'StockOptionLevel'

## Factor 2: **personal relationships**

- 'CurrMgr\_Coy'
- 'CurrRole\_CurrMgr'
- 'JobRole'

## Factor 3: **employee engagement**

- 'JobInvolvement',
- 'EnvironmentSatisfaction'



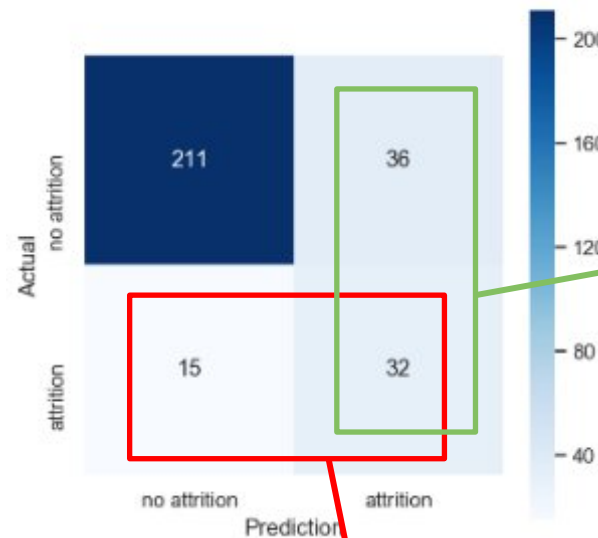
Note: Results are only based on the hypothetical dataset

# Conclusion

In this project, the **Random Forest Classifier** model has achieved prediction (Recall) score of **68.1%**

Out of all 47 attrition employees, 68.1% of them will be classified correctly using their background attributes and characteristics

Optimal threshold 0.288  
F1 Score = 55.7%



proportion of correct predictions out of all predicted attrition cases  
**Precision** =  $32/68$

**Accuracy** =  $(211+32)/294 = 82.7\%$

proportion of correct predictions out of all actual attrition cases  
**Recall** =  $32/47 = 68.1\%$



THANK YOU