Homework Solution - Stage 3 (Modelling)

Team:

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Submission:

- 1. Report: (https://github.com/zerobase-one/Pengabdi-FinPro.git)
- 2. Notebook:

(https://colab.research.google.com/drive/1PFYTh519RwDpR6SQikEsxsPCH0LNmR8g? usp=sharing)

Pre-Modelling

- Split Data Train & Data Test
- Scale Data Train & Data Test

Menggunakan:

```
-X_train_scaled = scaler.fit_transform(X_train)
-X test scaled = scaler.transform(X test)
```

- Handle Class Imbalance dengan RandomOverSampler
- Feature Selection (corr threshold: >0.07)

Age	-0.188209
DailyRate	-0.102016
DistanceFromHome	0.094112
JobLevel	-0.244937
MonthlyIncome	-0.277302
OverTime	0.295762
StockOptionLevel	-0.142361
TotalWorkingYears	-0.292519
TrainingTimesLastYear	-0.081083
YearsAtCompany	-0.242391
YearsInCurrentRole	-0.216552
YearsWithCurrManager	-0.231257
Accumulated_Satisfaction	-0.269434
YearsWorkingPerCompany	-0.268648
BusinessTravel_Non-Travel	-0.151371
BusinessTravel_Travel_Frequently	0.214879
BusinessTravel Travel Rarely	-0.114497

Department_Research & Development	-0.130844
Department_Sales	0.139799
MaritalStatus_Divorced	-0.128017
MaritalStatus_Married	-0.091326
MaritalStatus_Single	0.198954
JobRole_Healthcare Representative	-0.167948
JobRole_Laboratory Technician	0.154243
JobRole_Manager	-0.130960
JobRole_Manufacturing Director	-0.122460
JobRole_Research Director	-0.187044
JobRole_Sales Representative	0.184594

Modelling

• Train model dengan algoritma:

- Logistic Regression
- Linear SVC
- SVM
- KNN
- Gaussian Naive Bayes
- Perceptron
- Stochastic Gradient Descent
- Decision Tree
- Gradient Boosting Trees
- Random Forest

• Hasil Training Model

	•	
Mod	del Precision Score	
7	Decision Tree	100.00
9	Random Forest	100.00
8	Gradient Boosting Trees	94.67
3	KNN	90.64
1	SVM	87.12
5	Perceptron	84.42
0	Logistic Regression	77.32
2	Linear SVC	76.50
6	Stochastic Gradient Decent	68.76
4	Naive Bayes	58.06

Model Precision CV 10-Fold

9	Random Forest	96.20
8	Gradient Boosting Trees	89.16
7	Decision Tree	88.60

KNN	82.89
SVM	82.53
Linear SVC	76.01
Logistic Regression	75.96
Stochastic Gradient Decent	73.68
Perceptron	73.46
Naive Bayes	58.13
	SVM Linear SVC Logistic Regression Stochastic Gradient Decent Perceptron

Model Evaluasi 1

Model: Logistic Regression

Precision Scores: [0.78378378 0.69444444 0.7625 0.78082192

0.734375 0.703125

0.78571429 0.85483871 0.71014493 0.78666667]

Mean Precision Score: 0.760

Model: SVM

Precision Scores: [0.82608696 0.81428571 0.80246914 0.85333333

0.84285714 0.8028169

0.80519481 0.89393939 0.79710145 0.82278481]

Mean Precision Score: 0.826

Model: Linear SVC

Precision Scores: [0.80555556 0.70422535 0.7625 0.78082192

0.72131148 0.67692308

0.77464789 0.88333333 0.72727273 0.76623377]

Mean Precision Score: 0.760

Model: KNN

Precision Scores: [0.79487179 0.7654321 0.77011494 0.77272727 0.75

0.80246914

0.81176471 0.825 0.7875 0.78823529]

Mean Precision Score: 0.787

Model: Decision Tree

Precision Scores: [0.84337349 0.8625 0.8452381 0.87654321

0.91025641 0.8875

0.91025641 0.89873418 0.91025641 0.8452381]

Mean Precision Score: 0.879

Model: Gradient Boosting Trees

Precision Scores: [0.89189189 0.88 0.85365854 0.92

0.88311688 0.86419753

0.93243243 0.93150685 0.90789474 0.86075949]

Mean Precision Score: 0.893

Model: Random Forest

Precision Scores: [0.97222222 0.93333333 0.94666667 0.97260274

0.93421053 0.97260274

1. 0.97260274 0.98611111 0.95945946]

Mean Precision Score: 0.965

Model Evaluasi 2

Model: Logistic Regression

Precision Scores: [0.78378378 0.69444444 0.7625 0.78082192

0.734375 0.703125

0.78571429 0.85483871 0.71014493 0.78666667]

Mean Precision Score: 0.760

Standard Deviation of Precision Scores: 0.047

Model: SVM

Precision Scores: [0.82608696 0.81428571 0.80246914 0.85333333

0.84285714 0.8028169

0.80519481 0.89393939 0.79710145 0.82278481]

Mean Precision Score: 0.826

Standard Deviation of Precision Scores: 0.029

Model: Linear SVC

Precision Scores: [0.80555556 0.70422535 0.7625 0.78082192

0.72131148 0.67692308

0.77464789 0.88333333 0.72727273 0.76623377]

Mean Precision Score: 0.760

Standard Deviation of Precision Scores: 0.055

Model: KNN

Precision Scores: [0.79487179 0.7654321 0.77011494 0.77272727 0.75

0.80246914

0.81176471 0.825 0.7875 0.78823529]

Mean Precision Score: 0.787

Standard Deviation of Precision Scores: 0.022

Model: Decision Tree

Precision Scores: [0.86419753 0.85185185 0.86585366 0.87654321

0.91025641 0.91025641

0.89873418 0.91025641 0.93421053 0.8875

Mean Precision Score: 0.891

Standard Deviation of Precision Scores: 0.025

Model: Gradient Boosting Trees

Precision Scores: [0.89189189 0.88311688 0.85365854 0.92

0.88311688 0.86419753

0.93243243 0.93150685 0.90789474 0.86075949]

Mean Precision Score: 0.893

Standard Deviation of Precision Scores: 0.028

Model: Random Forest

Precision Scores: [0.98591549 0.93333333 0.95945946 0.98611111

0.93421053 0.97260274

0.98611111 0.95945946 0.98611111 0.97260274]

Mean Precision Score: 0.968

Standard Deviation of Precision Scores: 0.020

Based on the precision score, model evaluation 1 & model evaluation 2, we decided to choose the model with lowest standard deviation of precision score and highest Precision Score. So, we decided to choose Random Forest Classifier.

Hyperparameter Tuning RandomForestClassifier dengan RandomizedSearchCV

Best parameters: {'n_estimators': 100, 'min_samples_split': 5, 'min_samples_leaf': 1,

'max_depth': 50, 'criterion': 'gini'}

Best precision score: 0.9610206865120349

Check Model with Best Params

Random Forest Precision Default Before (Training): 1.0

Random Forest Precision CV 10-Fold Default Before (Training): 0.959349593495935

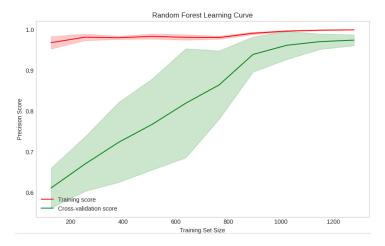
Random Forest Precision Best After (Training): 1.0

Random Forest Precision CV 10-Fold Best After (Training): 0.9528301886792453

Random Forest Precision Best After (Test): 0.666666666666666666

Random Forest Precision CV 10-Fold Best After (Test): 0.6428571428571429

Learning Curve



Predict Data Test using model with best params

[88] print("Precision: %s" % score)

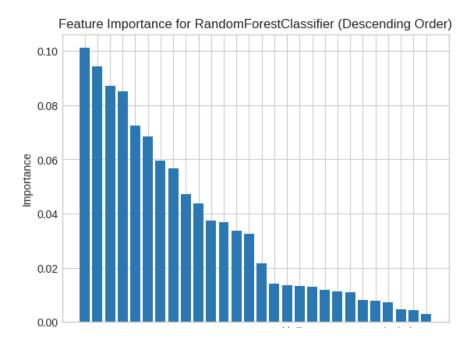
Precision: 80.0

[89] print(classification_report(y_test, predictions))

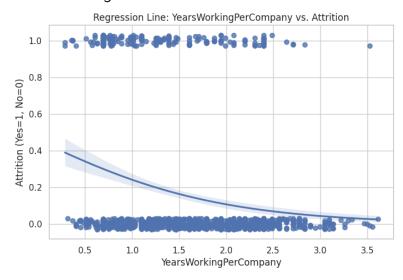
	precision	recall	f1-score	support
0 1	0.86 0.80	0.99 0.14	0.92 0.24	308 58
accuracy macro avg weighted avg	0.83 0.85	0.57 0.86	0.86 0.58 0.81	366 366 366

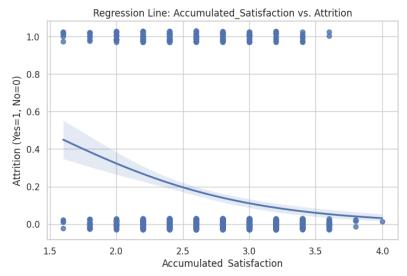
Feature Importance

- 1. Feature: MonthlyIncome, Score: 0.10109
- Feature: YearsWorkingPerCompany, Score: 0.09413
 Feature: Accumulated_Satisfaction, Score: 0.08709
- 4. Feature: Age, Score: 0.08498
- 5. Feature: DailyRate, Score: 0.07232
- 6. Feature: OverTime, Score: 0.06828
- 7. Feature: DistanceFromHome, Score: 0.05947
- 8. Feature: TotalWorkingYears, Score: 0.05674
- 9. Feature: YearsAtCompany, Score: 0.04714
- 10. Feature: YearsWithCurrManager, Score: 0.04364
- 11. Feature: YearsInCurrentRole, Score: 0.03737
- 12. Feature: JobLevel, Score: 0.03689
- 13. Feature: TrainingTimesLastYear, Score: 0.03364
- 14. Feature: StockOptionLevel, Score: 0.03251
- 15. Feature: BusinessTravel_Travel_Frequently, Score: 0.02153
- 16. Feature: MaritalStatus Single, Score: 0.01420
- 17. Feature: Department Research & Development, Score: 0.01361
- 18. Feature: JobRole_Laboratory Technician, Score: 0.01320
- 19. Feature: Department_Sales, Score: 0.01300
- 20. Feature: MaritalStatus Divorced, Score: 0.01189
- 21. Feature: MaritalStatus Married, Score: 0.01120
- 22. Feature: BusinessTravel Travel Rarely, Score: 0.01107
- 23. Feature: JobRole_Healthcare Representative, Score: 0.00798
- 24. Feature: BusinessTravel Non-Travel, Score: 0.00791
- 25. Feature: JobRole Manufacturing Director, Score: 0.00713
- 26. Feature: JobRole Sales Representative, Score: 0.00467
- 27. Feature: JobRole Research Director, Score: 0.00438
- 28. Feature: JobRole_Manager, Score: 0.00293



Menarik untuk dianalisis lebih lanjut, bahwa kedua fitur baru yang dihasilkan dari ekstraksi fitur-fitur yang sudah ada menempati posisi Top 3 feature importance. Selanjutnya kedua fitur ini menjadi objek menarik untuk ditelaah lebih lanjut untuk diberikan insight dan rekomendasi.





Grafik di atas menunjukkan bahwa hubungan antara YearsWorkingPerCompany dengan Attrition menunjukkan bahwa semakin besar nilai YearsWorkingPerCompany maka kecenderungan untuk Attrition akan menurun. Hal ini menjelaskan bahwa semakin besar nilai, semakin loyal karyawan tersebut. Dalam upaya meningkatkan loyalitas karyawan, diperlukan aksi nyata oleh perusahaan. Salah satunya dengan melihat fitur Accumulated_Satisfaction yang juga menunjukkan semakin tinggi angka kepuasan karyawan, semakin rendah kecenderungan Attrition dari karyawan itu sendiri. Oleh karena itu, aksi nyata peningkatan kepuasan karyawan secara bersamaan juga akan meningkatkan nilai loyalitas karyawan. Maka, rekomendasi aksi yang dapat dilakukan oleh perusahaan adalah yang berorientasi pada peningkatan faktor kepuasan akumulasi yaitu, Environment Satisfaction, Job Satisfaction, Job Involvement, Relationship Satisfaction dan Worklife Balance karyawan.

RECOMMENDATIONS:

Rekomendasi aksi/program:

- 1. Environment Satisfaction:
 - To ensure that the compensation and benefits offered to employees are attractive in order to keep employee morale high. The role of compensation and benefits is critical in hiring and retaining qualified and talented employees. (Sinta, Azmieti K., Setiadi, B., Jumawan, Damayanti, Endah S. and Soehaditama, Josua Panatap. (2023). "Employee Retention Strategy: Analysis Path Career, Compensation Benefit, Organization Commitment & Reward System". Retrieved from: https://journal.formosapublisher.org/index.php/eajmr/article/view/3672/3404.)
 - Uses various rewards or incentives to attract, keep people and motivate them to achieve personal goals and goals organization. (Sinta, Azmieti

K., Setiadi, B., Jumawan, Damayanti, Endah S. and Soehaditama, Josua Panatap. (2023). "Employee Retention Strategy: Analysis Path Career, Compensation Benefit, Organization Commitment & Reward System". Retrieved from:

https://journal.formosapublisher.org/index.php/eajmr/article/view/3672/3404.)

2. Job Satisfaction:

- Focus on training and development. Identifying the training need and giving appropriate training to the employees make them to enhance their skills and motivate them to reach higher positions. (Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF LITERATURE ON EMPLOYEE RETENTION". Retrieved from:
 - https://www.researchgate.net/publication/335677274.
- It is essential to conduct a survey to measure the satisfaction level of employees in organizations. The reason behind this is that this would help in understanding the satisfaction level of the employees and steps can be taken to improve the satisfaction level. (Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF LITERATURE ON EMPLOYEE RETENTION". Retrieved from:
 - https://www.researchgate.net/publication/335677274.)
- A career path or career path is defined as a set of positions that each employee must go through to reach a certain level of position within a company. (Sinta, Azmieti K., Setiadi, B., Jumawan, Damayanti, Endah S. and Soehaditama, Josua Panatap. (2023). "Employee Retention Strategy: Analysis Path Career, Compensation Benefit, Organization Commitment & Reward System". Retrieved from:
 https://journal.formosapublisher.org/index.php/eajmr/article/view/3672/3404.)
- Security at work place especially in the case of women employees may help in retaining pool of women workforce. (Karumuri, Venkateswarlu and Singareddi, Sriprasanthi. (2014). "EMPLOYEE ATTRITION AND RETENTION: A THEORETICAL PERSPECTIVE". Retrieved from: https://citeseerx.ist.psu.edu/document?repid=rep1&type=pdf&doi=8a8ba9c73f3e624032e71c31a3cbf1a94f935905.)

3. Job Involvement:

 The management should consider the ideas and opinions of employees while any decisions are taken with related to work and other matter.
 Active participation of employees in the decision making process increases the morale of the employee. (Karumuri, Venkateswarlu and Singareddi, Sriprasanthi. (2014). "EMPLOYEE ATTRITION AND RETENTION: A THEORETICAL PERSPECTIVE". Retrieved from: https://citeseerx.ist.psu.edu/document?repid=rep1&type=pdf&doi=8a8ba9c73f3e624032e71c31a3cbf1a94f935905.)

 The employees can be empowered in their works by giving autonomy in performing their jobs without more involvement of superiors. (Karumuri, Venkateswarlu and Singareddi, Sriprasanthi. (2014). "EMPLOYEE ATTRITION AND RETENTION: A THEORETICAL PERSPECTIVE". Retrieved from:

 $\frac{https://citeseerx.ist.psu.edu/document?repid=rep1&type=pdf&doi=8a8ba}{9c73f3e624032e71c31a3cbf1a94f935905}.)$

To avoid monotony of work in some jobs, fun related activities, games, get together, stress management programs can be organized. (Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF LITERATURE ON EMPLOYEE RETENTION". Retrieved from:

https://www.researchgate.net/publication/335677274.)

4. Relationship Satisfaction:

- Equality and fairness treatment among the employees would create a sense that everyone is treated as one and the same i.e. there is no bias among the employees. (Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF LITERATURE ON EMPLOYEE RETENTION".
 Retrieved from: https://www.researchgate.net/publication/335677274.)
- It is highly important to conduct exit interview to identify the reasons behind leaving of employees in organizations. This would help in strengthening the strategies for employee retention. (Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF LITERATURE ON EMPLOYEE RETENTION". Retrieved from: https://www.researchgate.net/publication/335677274.)
- Mutual trust between the employer and employee must be created so as to bring a feel that employees are part of the organization and they are working not only for their benefit but also for the benefit of the organization and as well as for the benefit of the society. (Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF LITERATURE ON EMPLOYEE RETENTION". Retrieved from:

https://www.researchgate.net/publication/335677274.)

5. Worklife Balance:

 Work from home can be encouraged in organizations where there is no crèche facility which would boost the morale of the female employees. (Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF

LITERATURE ON EMPLOYEE RETENTION". Retrieved from: https://www.researchgate.net/publication/335677274.)

Fair and competitive salaries, performance related incentives may help in retaining the talented employees with the organization for a long period. (Karumuri, Venkateswarlu and Singareddi, Sriprasanthi. (2014). "EMPLOYEE ATTRITION AND RETENTION: A THEORETICAL PERSPECTIVE". Retrieved from: https://citeseerx.ist.psu.edu/document?repid=rep1&type=pdf&doi=8a8ba9c73f3e624032e71c31a3cbf1a94f935905.)

Reference:

- 1. Vetrivel, T., Ramesh, S. and M.S, Kamalaveni. (2019). "A REVIEW OF LITERATURE ON EMPLOYEE RETENTION". Retrieved from: https://www.researchgate.net/publication/335677274.
- Karumuri, Venkateswarlu and Singareddi, Sriprasanthi. (2014). "EMPLOYEE
 ATTRITION AND RETENTION: A THEORETICAL PERSPECTIVE". Retrieved from:
 https://citeseerx.ist.psu.edu/document?repid=rep1&type=pdf&doi=8a8ba9c73f
 3e624032e71c31a3cbf1a94f935905.
- 3. Sinta, Azmieti K., Setiadi, B., Jumawan, Damayanti, Endah S. and Soehaditama, Josua Panatap. (2023). "Employee Retention Strategy: Analysis Path Career, Compensation Benefit, Organization Commitment & Reward System". Retrieved from:

https://journal.formosapublisher.org/index.php/eajmr/article/view/3672/3404.