

PwC Employee Reviews Analysis

Source: glassdoor.com

Manel RAHALI

Soumaya ATOUI

Hadiqa Javaid

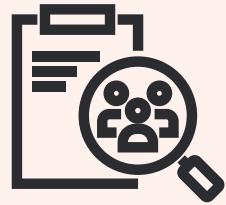
21/11/2025



A G E N D A

• Business use case	3
• Presenting DataFrame	4
• Data quality issues	5
• Final DataFrame:	6
• Univariate analysis : Rating distribution	7
• Distribution of risk vs. not at risk	8
• Correlation matrix	9
• ML models used	10
• Preprocessing (the country example)	11
• Results: CatBoost Model	12
• Recommandations	13

Business Use Case



**Help PwC HR teams
identify employees who
may be at risk of low
satisfaction**



**Understand which
countries, roles or
functions consistently
show higher risk levels**



**Build a model that
predicts the at_risk
label derived from
ratings ≤ 3**

Presenting the DataFrame

1. Content

Reviews from :
Current & Former
Employees of PwC

2. Size

12 Columns
43K+ Rows
(before cleaning)

3. How & When?

'GLASSDOOR'
2008 - 2024

Data quality issues

df

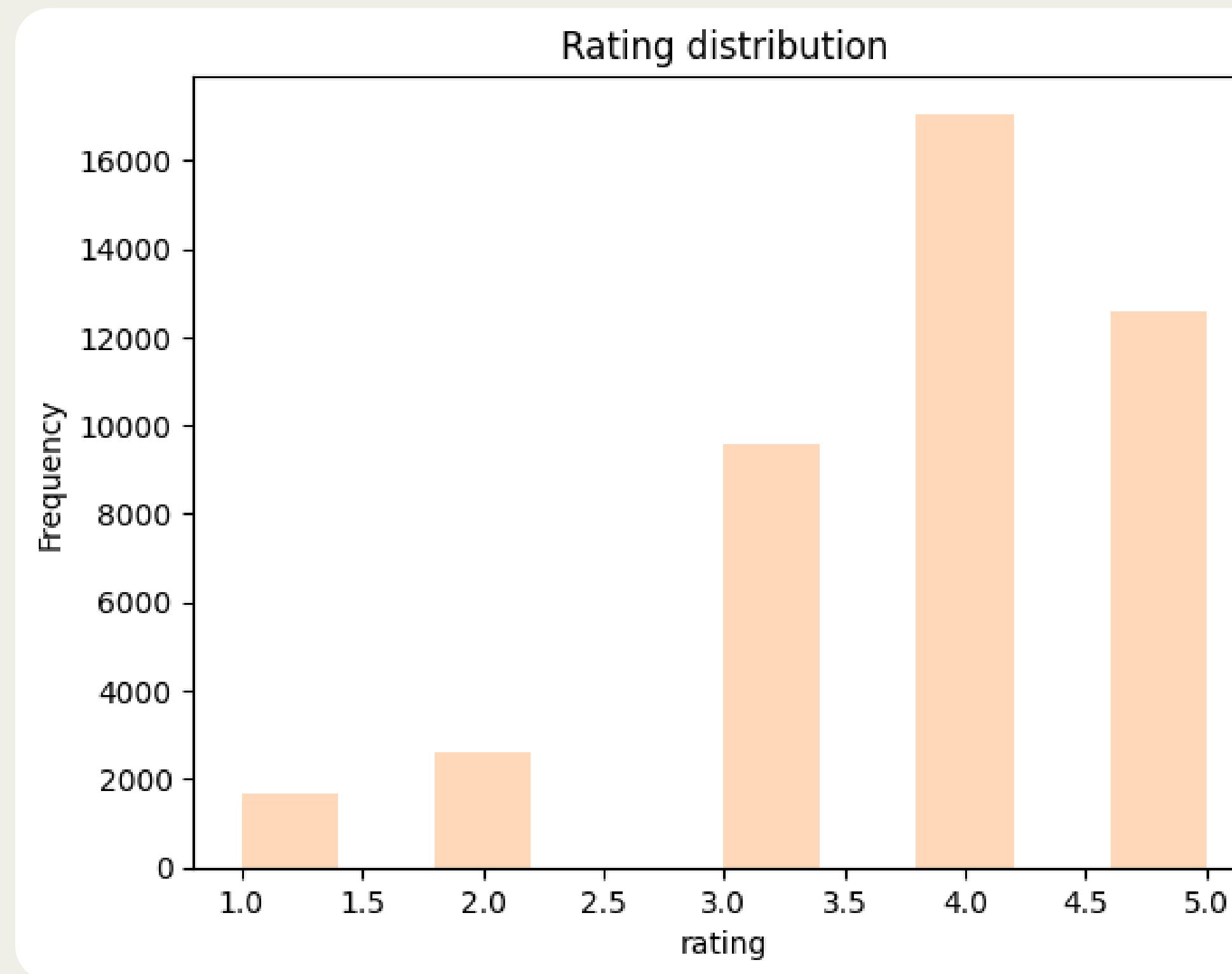
	Unnamed: 0	date	location	position	rating	employee_type	title	pros	cons	Recommander	Approbation du PDG	Perspective commerciale
0	0	19 avr. 2022	Sydney	Employé anonyme	2,0	Ancien employé	Toxic culture	Nice office and good benefits	Toxic culture, under paid, lack of recognition.	negativeStyles	noDataStyles	noDataStyles
1	1	29 déc. 2021	Islâmâbâd,	Associate Consultant	3,0	Ancien employé, plus d'un an	PwC Pakistan - Technology Advisory (Islamabad)	Good exposure, average compensation, good team...	Unprofessional work culture, severe lack of di...	positiveStyles	positiveStyles	negativeStyles
2	2	29 mars 2022	Singapour	Director	4,0	Employé actuel, plus de 3 ans	good balanced career	Good network; Average work life balance amongs...	Pay not competitive; Senior management only pa...	NaN	noDataStyles	noDataStyles
3	3	12 avr. 2022	Dublin, Dublin	Senior Manager IT	4,0	Employé actuel, plus de 3 ans	Good work life balance	good culture, supportive people & leadership	Lower salary compared to industry	noDataStyles	noDataStyles	noDataStyles
4	4	12 avr. 2022	Kuala Lumpur	Senior Manager	5,0	Employé actuel	Pros & Cons	Flexible working hours and workspace	Heavy workload and long working hours	noDataStyles	NaN	noDataStyles
...
43447	43447	20 févr. 2023	New York, NY	AML Analyst	4,0	Ancien intérimaire	Work	Laptop. Solid Team. Industry Standard working	You do what you need to do. That's it.	noDataStyles	noDataStyles	noDataStyles
43448	43448	21 févr. 2023	Jakarta	Assurance Associate	5,0	Ancien employé, plus d'un an	great experience	improve your accounting skills and teamwork	bad work-life balance especially in peak season	positiveStyles	positiveStyles	positiveStyles
43449	43449	21 févr. 2023	Singapour	Associate	4,0	Employé actuel	Fresh Graduate Experience	good coaching culture with coaching manager an...	no fixed workplace but flexible	NaN	noDataStyles	noDataStyles
43450	43450	28 févr. 2023	Chicago, IL	Senior Associate	3,0	Employé actuel	Decent	Great enviornment to be in	Too long of hours to work	noDataStyles	noDataStyles	noDataStyles
43451	43451	28 févr. 2023	Hong Kong	Management Trainee	4,0	Employé actuel	Nan	Standard work environment and pace	confused structure and program design	noDataStyles	noDataStyles	noDataStyles

43452 rows × 12 columns

Final Data Frame

employee_status	recommender_bin	perspective_commercial_e_bin	approbation_pdg_bin	year	at_risk	position_level_encoded	function_area_Audit-Assurance	function_area_Consulting-Advisory	
0	-1	0	0	2022	1	8	0	0	0
0	1	-1	1	2021	1	2	0	1	
1	1	0	0	2022	0	6	0	0	
1	1	0	0	2022	0	5	0	0	
1	1	1	1	2022	0	5	0	0	
...
0	1	0	0	2023	0	2	0	0	0
0	1	1	1	2023	0	2	1	0	
1	1	0	0	2023	0	2	0	0	
1	1	0	0	2023	1	3	0	0	
1	1	0	0	2023	0	0	0	0	
function_area_Deals-Transactions	function_area_Other	function_area_Risk	function_area_Support (HR/Ops/Finance)		function_area_Tax	function_area_Tech-IT-Data	country_encoded		
0	1	0		0	0	0	0	1	1
0	0	0		0	0	0	0	42	42
0	1	0		0	0	0	0	49	49
0	0	0		0	0	0	1	24	24
0	1	0		0	0	0	0	34	34
...
0	0	0		0	0	0	1	62	62
0	0	0		0	0	0	0	23	23
0	1	0		0	0	0	0	49	49
0	1	0		0	0	0	0	62	62
0	1	0		0	0	0	0	20	20

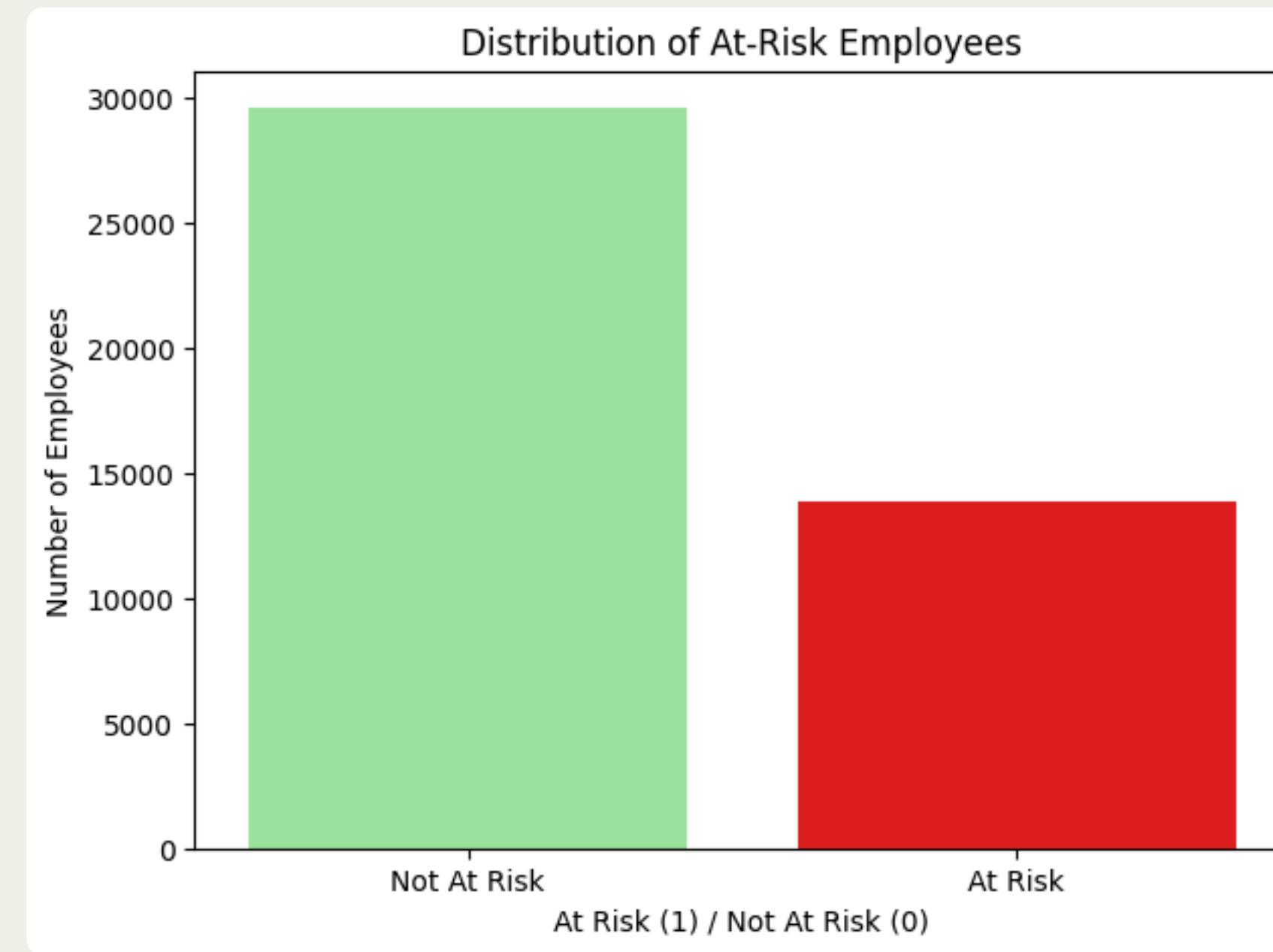
Univariate analysis : Rating distribution



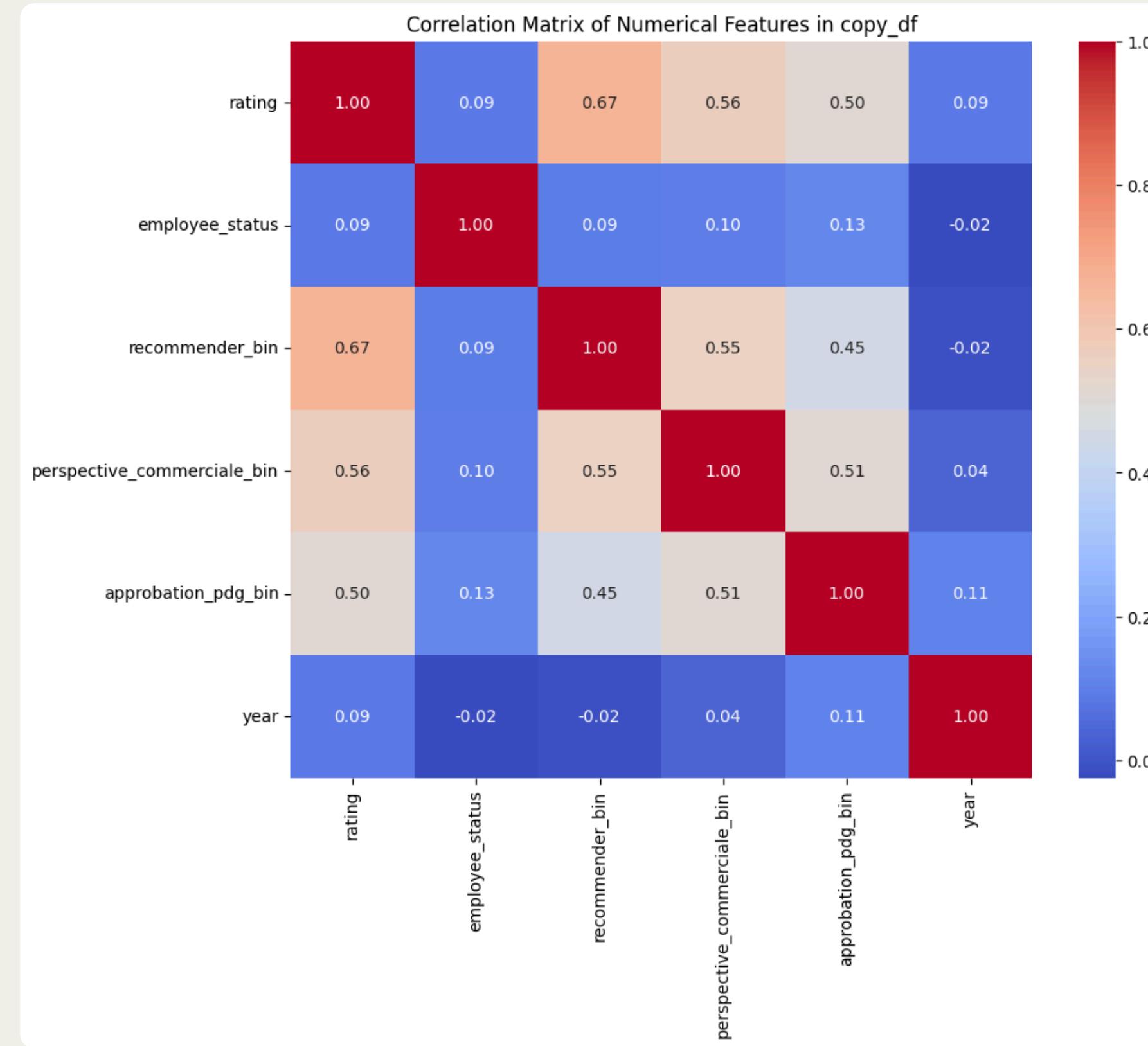
3.83

Average Rating

Distribution of risk vs. not at risk



Correlation Matrix



ML models used



Random Forest

test 1: 100% accuracy

test 2: 74 % accuracy



LightGBM

70% Accuracy



CatBoost

93% Accuracy

Preprocessing (the country example)



The screenshot shows a Jupyter Notebook cell with the following code:

```
import category_encoders as ce

# Create a TargetEncoder instance
target_encoder = ce.TargetEncoder(cols=['country'])

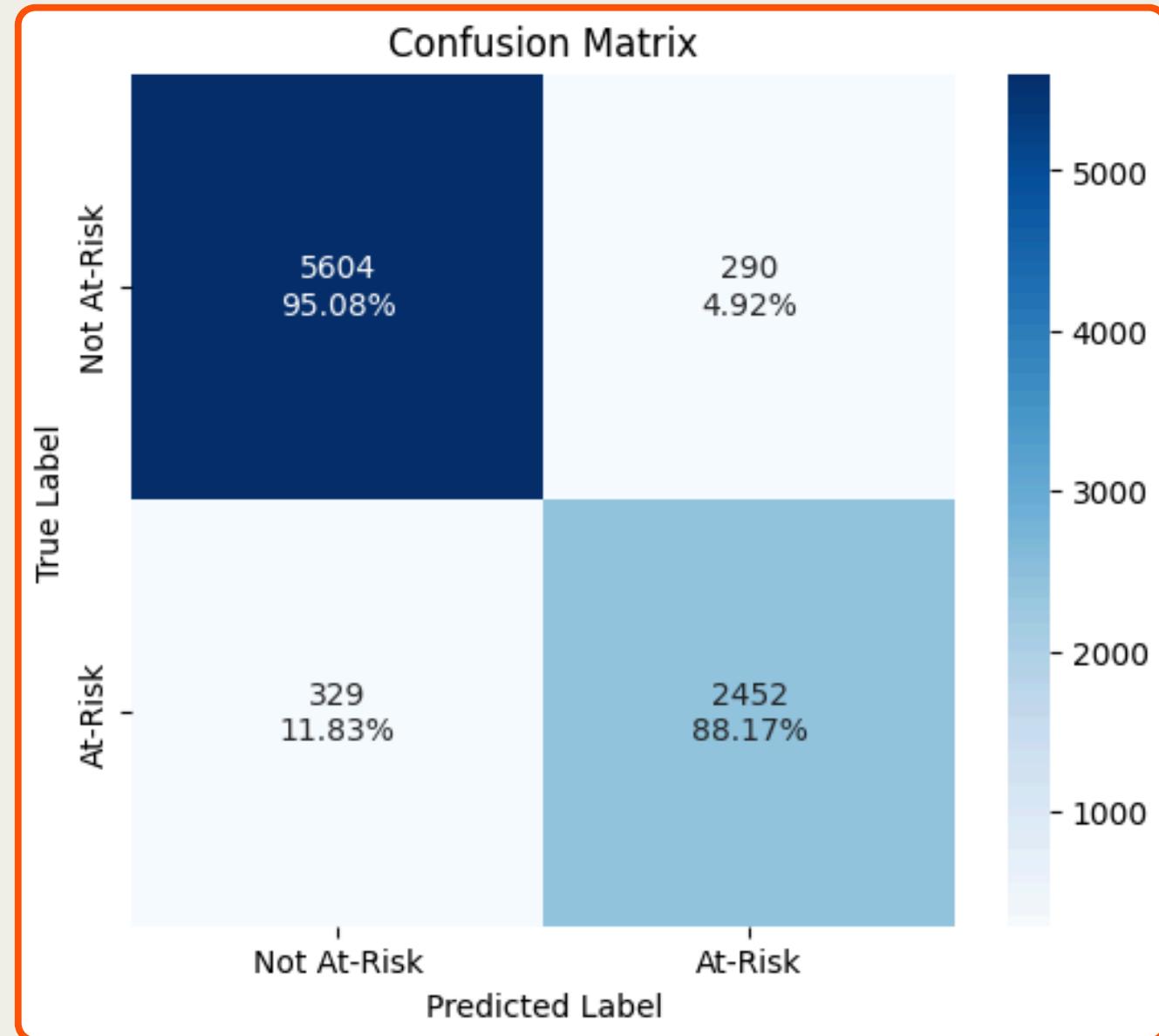
# Fit and transform the 'country' column
df_encoded['country_target_encoded'] = target_encoder.fit_transform(df_encoded['country'], df_encoded['at_risk'])

# Display the 'country' and 'country_target_encoded' columns
display(df_encoded[['country', 'country_target_encoded']].head())
```

The code imports the `category_encoders` module as `ce`. It creates a `TargetEncoder` instance for the `'country'` column. Then, it fits and transforms the `'country'` column using the `fit_transform` method, storing the result in the `'country_target_encoded'` column of the DataFrame `df_encoded`. Finally, it displays the first few rows of the DataFrame containing the `'country'` and `'country_target_encoded'` columns.

 CodeImage

Results: CatBoost Model



Confusion Matrix

Accuracy: 92.86 %
Precision: 89.42 %
Recall: 88.17 %
F1-Score: 88.79 %
ROC/AUC: 93.22 %

CatBoost Model Performance

Recommendations

What we Recommend?



Prioritize interventions per country/role/function

As we noticed that Hong Kong and Italy for example need more attention as they are more at risk, as well as Associates and the Audit assurance department



Collect Feedback

We recommend for PwC to install a system where the HR department collects feedback every Quarter



Use the model as an early-warning tool

Detect early warning signs based on engagement indicators through the new ML model

Thank you!

QUESTIONS?