# **Employee-Attrition**

This is a fictional data set created by IBM data scientists https://www.kaggle.com/pavansubhasht/ibm-hr-analytics-attrition-dataset

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#### Acerca del dataset

df = pd.read\_csv('WA\_Fn-UseC\_-HR-Employee-Attrition.csv')
df

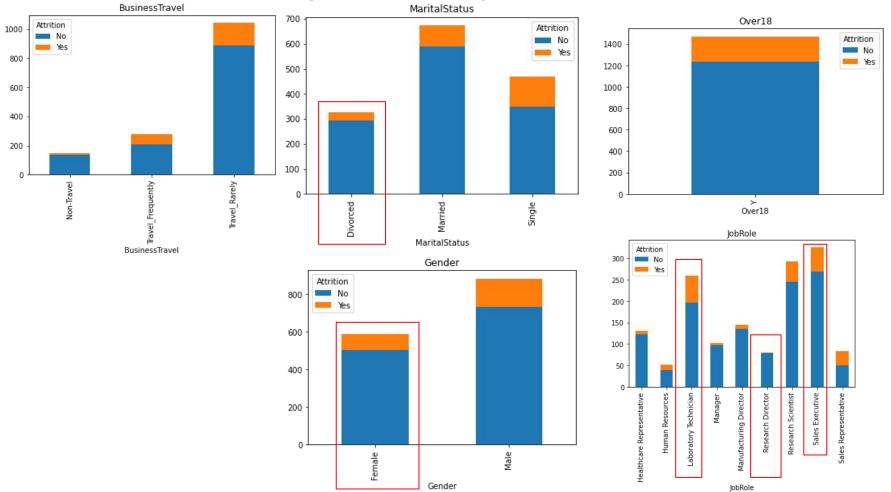
	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education
0	41	Yes	Travel_Rarely	1102	Sales	1	2
1	49	No	Travel_Frequently	279	Research & Development	8	1
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2
3	33	No	Travel_Frequently	1392	Research & Development	3	4
4	27	No	Travel_Rarely	591	Research & Development	2	1
				344			
1465	36	No	Travel_Frequently	884	Research & Development	23	2
1466	39	No	Travel_Rarely	613	Research & Development	6	1
1467	27	No	Travel_Rarely	155	Research & Development	4	3
1468	49	No	Travel_Frequently	1023	Sales	2	3
1469	34	No	Travel_Rarely	628	Research & Development	8	3

1470 rows × 35 columns

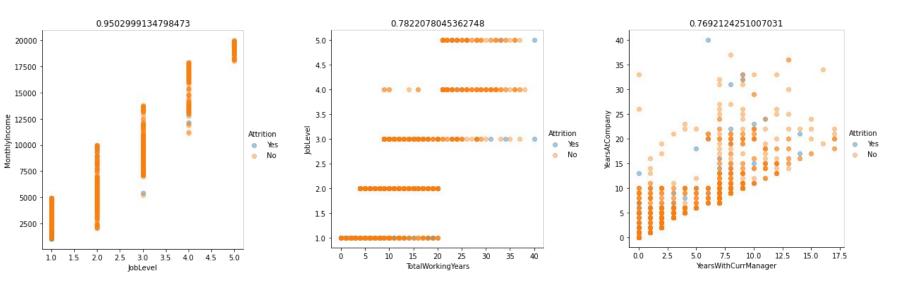
df.dtypes int64 Age Attrition object BusinessTravel object int64 DailyRate Department object DistanceFromHome int64 Education int64 EducationField object EmployeeCount int64 EmployeeNumber int64 EnvironmentSatisfaction int64 Gender object HourlyRate int64 JobInvolvement int64 JobLevel int64 JobRole object JobSatisfaction int64 MaritalStatus object int64 MonthlyIncome MonthlyRate int64 NumCompaniesWorked int64 Over18 object OverTime object PercentSalaryHike int64 PerformanceRating int64 RelationshipSatisfaction int64 StandardHours int64 StockOptionLevel int64 TotalWorkingYears int64 TrainingTimesLastYear int64 WorkLifeBalance int64 YearsAtCompany int64 YearsInCurrentRole int64 YearsSinceLastPromotion int64 YearsWithCurrManager int64 dtype: object



Distribución de registros categóricos por ...



# Scatter de registros numéricos por ...



# Datos agregados y desagregados

X_agregated	
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	Age	BusinessTravel	DailyRate	Departmen
0	41	Travel_Rarely	1102	Sales
1	49	Travel_Frequently	279	Research & Developmen
2	37	Travel_Rarely	1373	Research & Developmen
3	33	Travel_Frequently	1392	Research & Developmen
4	27	Travel_Rarely	591	Research & Developmen
		,,		
1465	36	Travel_Frequently	884	Research & Developmen
1466	39	Travel_Rarely	613	Research & Developmen
1467	27	Travel_Rarely	155	Research & Developmen
1468	49	Travel_Frequently	1023	Sales
1469	34	Travel_Rarely	628	Research & Developmen

X\_disagregated

	Age	DailyRate	DistanceFromHome	Education	EmployeeNumber	EnvironmentSatisfaction	Н
0	41	1102	1	2	1	2	
1	49	279	8	1	2	3	
2	37	1373	2	2	4	4	
3	33	1392	3	4	5	4	
4	27	591	2	1	7	1	
	1	***					
1465	36	884	23	2	2061	3	
1466	39	613	6	1	2062	4	
1467	27	155	4	3	2064	2	
1468	49	1023	2	3	2065	4	
1469	34	628	8	3	2068	2	

1470 rows × 52 columns

# RandomForest (datos agregados)

0.9514091350826045

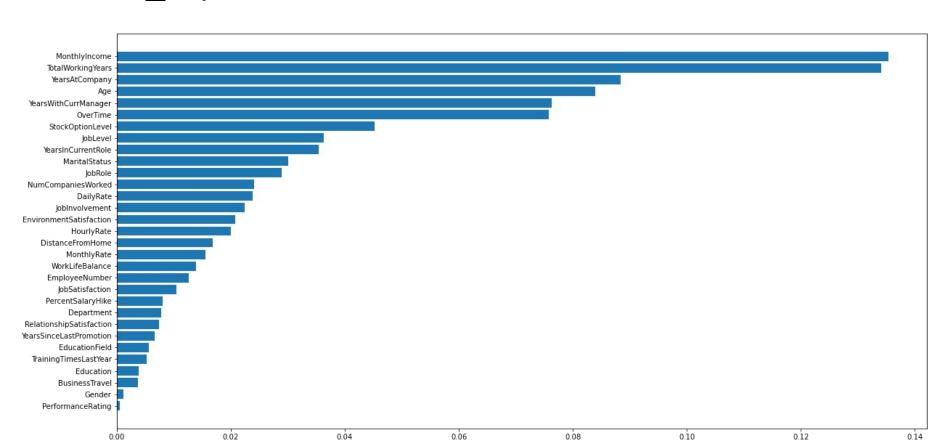
0.8299319727891157

```
rfc = RandomForestClassifier(bootstrap= True,
                                                    rfc = RandomForestClassifier(bootstrap= True,
                                                                                 max depth= 3,
                             max depth= 10,
                                                                                 max features= 'sqrt',
                             max features= 'sqrt',
                                                                                 min samples leaf= 2,
                             min_samples_leaf= 2,
                                                                                 min samples split= 5,
                             min samples split= 5,
                                                                                 n estimators= 500)
                             n estimators= 500)
                                                    rfc.fit(X agregated train, y train)
rfc.fit(X agregated train, y train)
                                                    print(rfc.score(X agregated train, y train))
print(rfc.score(X agregated train, y train))
                                                    print(rfc.score(X agregated test, y test))
print(rfc.score(X agregated test, y test))
```

0.8639455782312925

0.8299319727891157

# feature\_importances

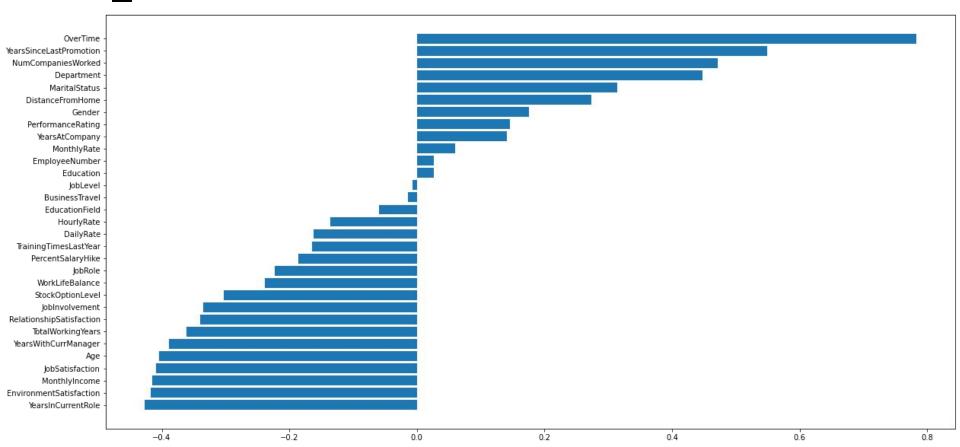


#### LogisticRegression (datoa agregados)

```
lr = LogisticRegression(random_state=42, multi_class='ovr').fit(X_agregated_train, y_train)
print(lr.score(X_agregated_train, y_train))
print(lr.score(X_agregated_test, y_test))
```

- 0.8794946550048591
- 0.8662131519274376

#### Coef\_

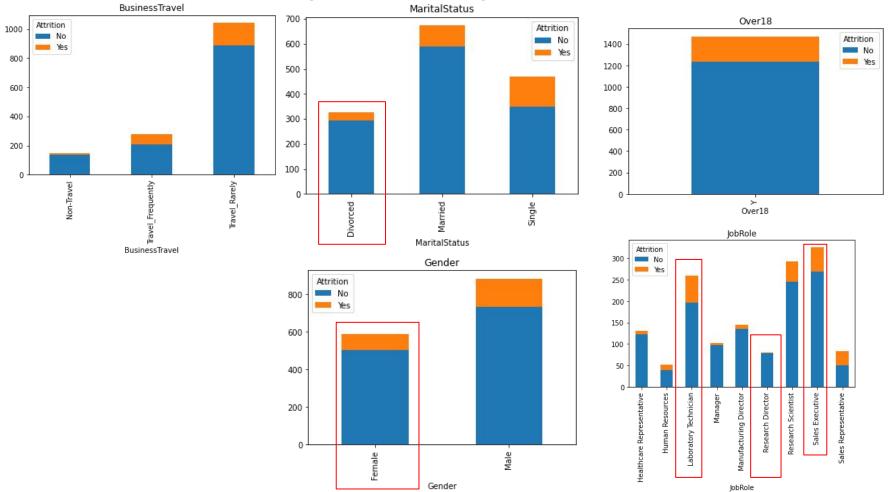


# RandomForest (datos desagregados)

0.8610301263362488 0.8390022675736961

MonthlyIncome -	
TotalWorkingYears -	
Age -	
YearsAtCompany ·	
YearsWithCurrManager -	
Yes_OverTime	
No_OverTime -	
StockOptionLevel -	
YearsInCurrentRole -	
JobLevel -	
Single -	
joblnvolvement -	
NumCompaniesWorked -	
EnvironmentSatisfaction -	
DailyRate -	
HourlyRate -	
Sales Representative -	
DistanceFromHome -	
WorkLifeBalance -	
MonthlyRate -	
Laboratory Technician	
JobSatisfaction -	
EmployeeNumber -	
Travel_Frequently	
Research & Development	
RelationshipSatisfaction -	
PercentSalaryHike -	
Sales -	
Sales Executive	
Research Scientist	
YearsSinceLastPromotion -	
Married -	
Education -	
TrainingTimesLastYear -	
Human Resources JobRole -	
Marketing -	
Divorced -	
Non-Travel -	
Travel_Rarely	
Healthcare Representative -	
Male -	
Manufacturing Director -	
Female -	
Life Sciences -	
Technical Degree -	
Medical -	
Human Resources	
Manager -	
Human Resources_Department	
Research Director -	
T.	
PerformanceRating -	
Other -	

Distribución de registros categóricos por ...



### Logistic regresion (datos desagregados)

