



Predicting Post-Graduation Salaries using Decision Tree and Random Forest

Graduate Tracer Study

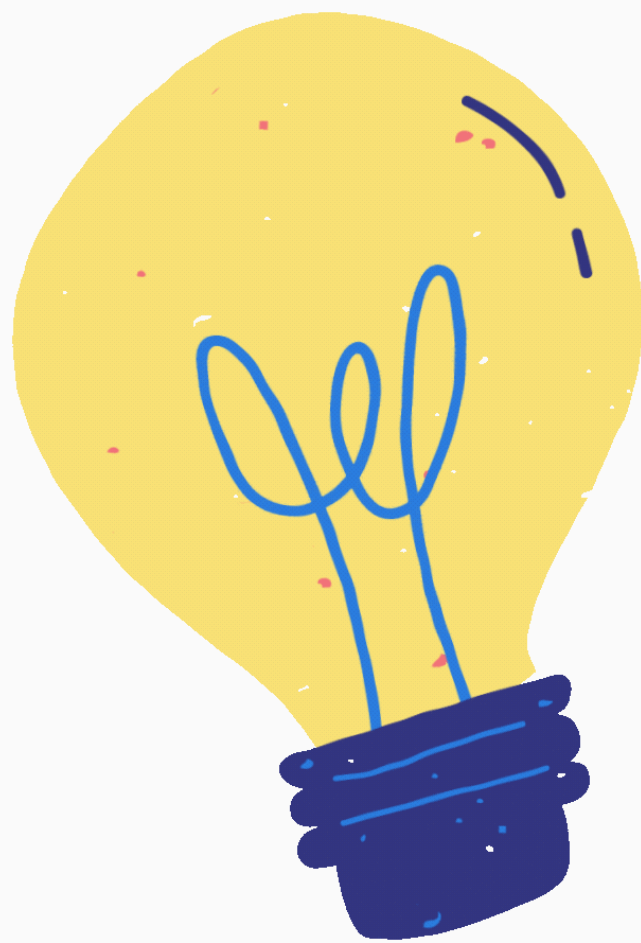
PREPARED BY

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CS2705B

PROBLEM STATEMENT

Limited exploration of the intricate relationship between program code, job occupation, company nature, years of working, and current salary in the existing dataset.





OBJECTIVES

- To study the correlation between program codes and post-graduation salaries.
- To build predictive models using classification technique to forecast post-graduation salaries based on the courses taken by graduates.
- To evaluate the accuracy of the predictive models.



| **Dataset Used**

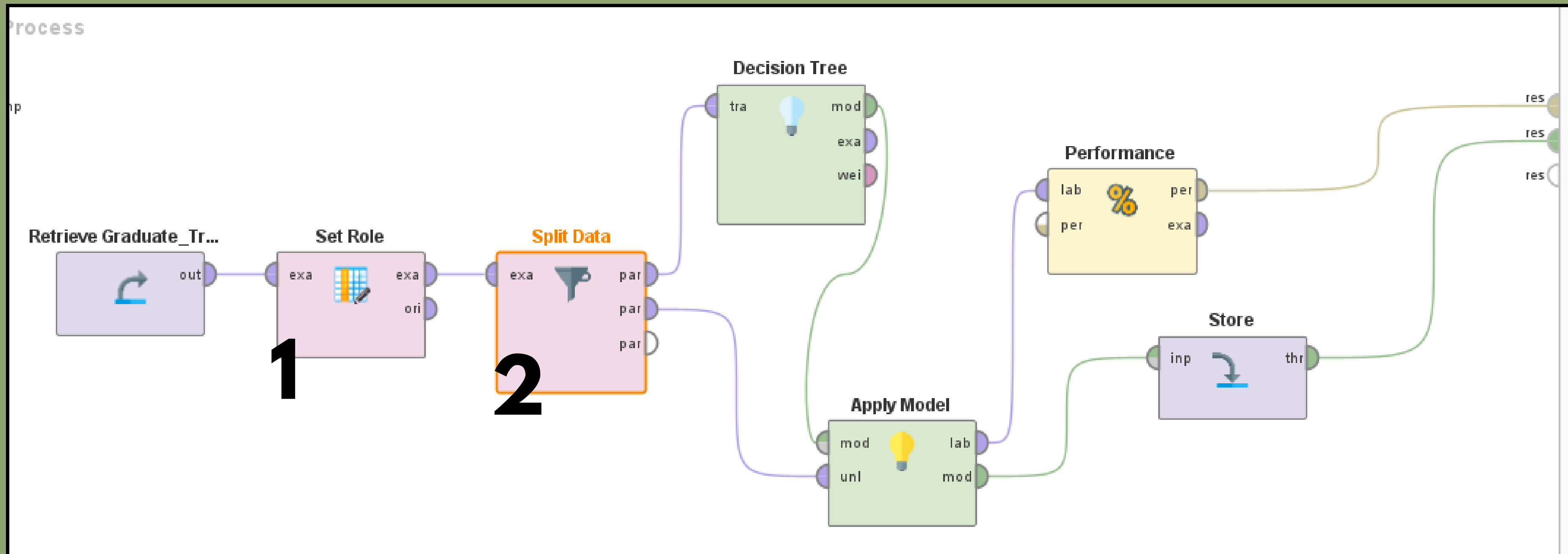
Data Source	Graduate Tracer Study
Total Data	902
Total Attributes	12
Target Class Attribute	Current Salary Midpoint
List of Attributes	Gender, Age, Graduation Year, Program Code, Current Occupation, Job category, Department, Company Name, Type of Company, Company's nature of business, Years of work experience (after graduation), Current Salary Midpoint

DECISION TREE

Modelling



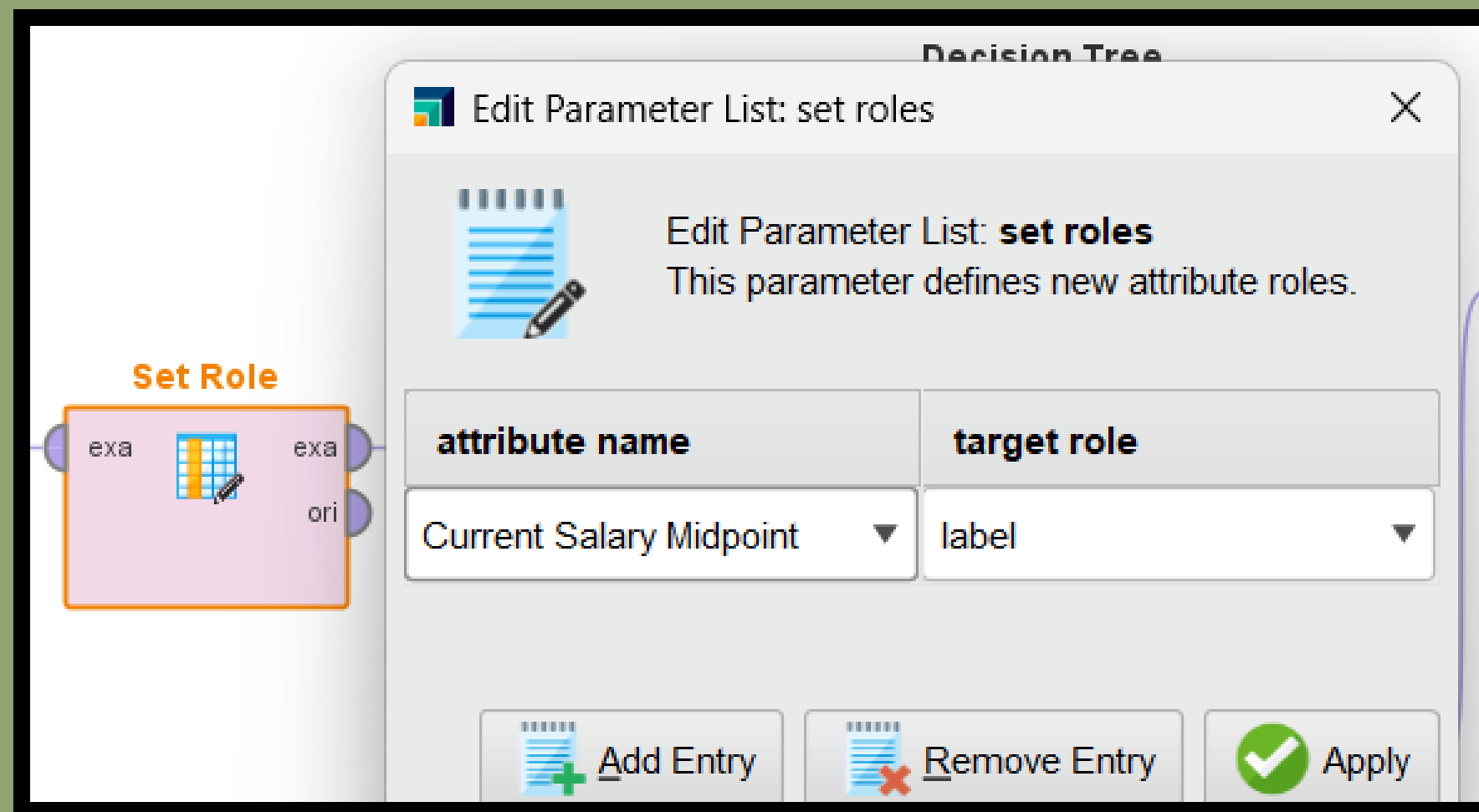
Process in RapidMiner



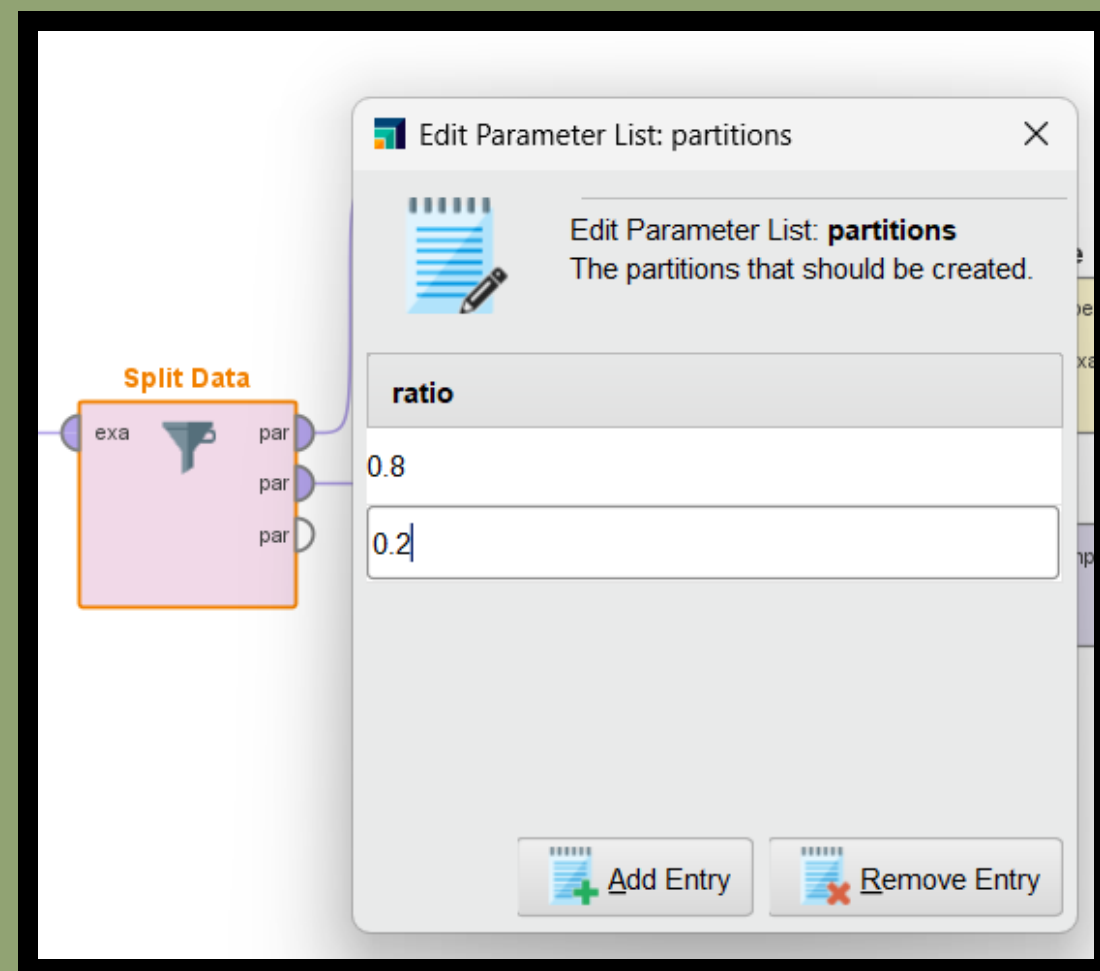
DECISION TREE

Setting Parameters

1 Set Role Class Label - Current Salary Midpoint



2 Split Data 0.8 - Training 0.2 - Testing



DECISION TREE

DECISION TREE

Evaluation

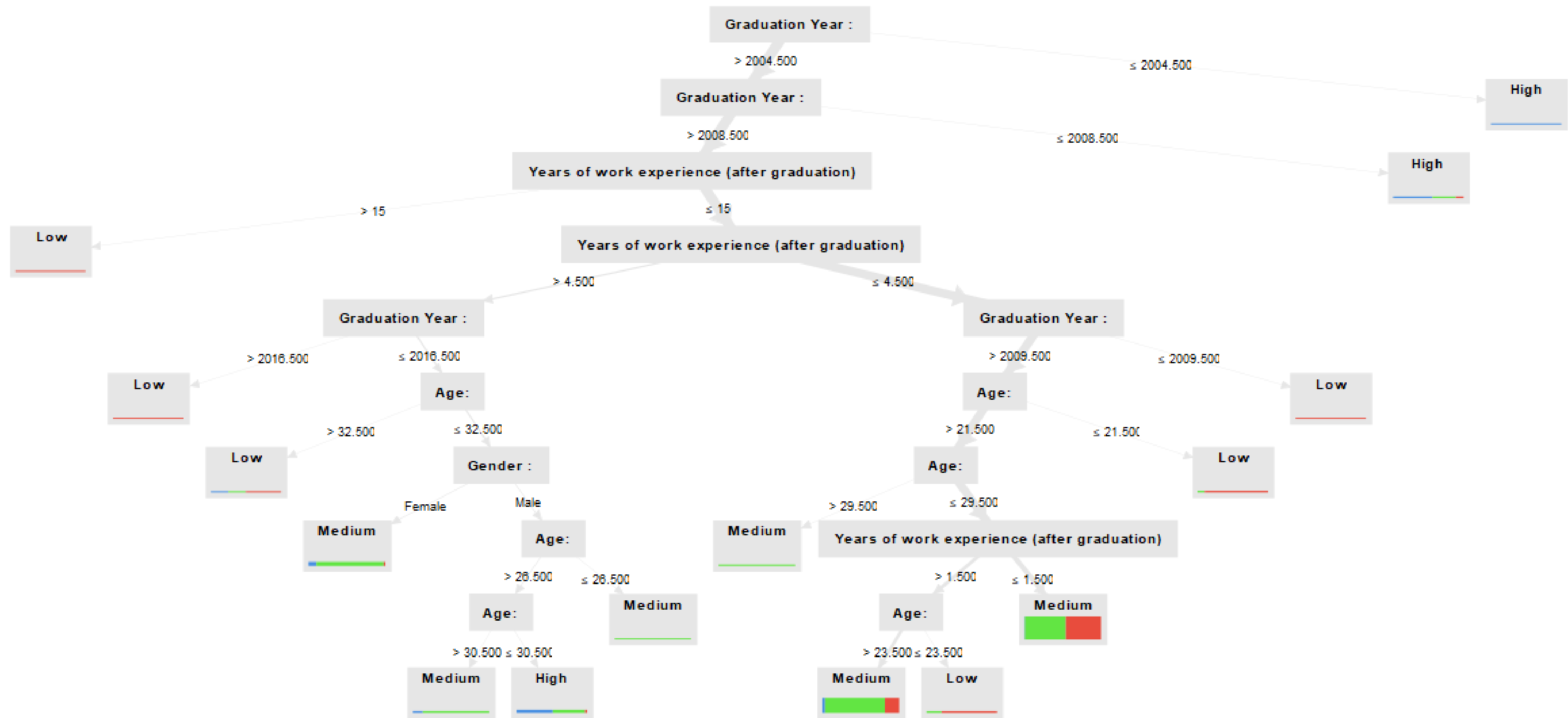


Confusion Matrix and Accuracy Percentage

accuracy: 62.15%

	true High	true Medium	true Low	class precision
pred. High	7	6	0	53.85%
pred. Medium	3	101	55	63.52%
pred. Low	1	2	2	40.00%
class recall	63.64%	92.66%	3.51%	

DECISION TREE



Tree

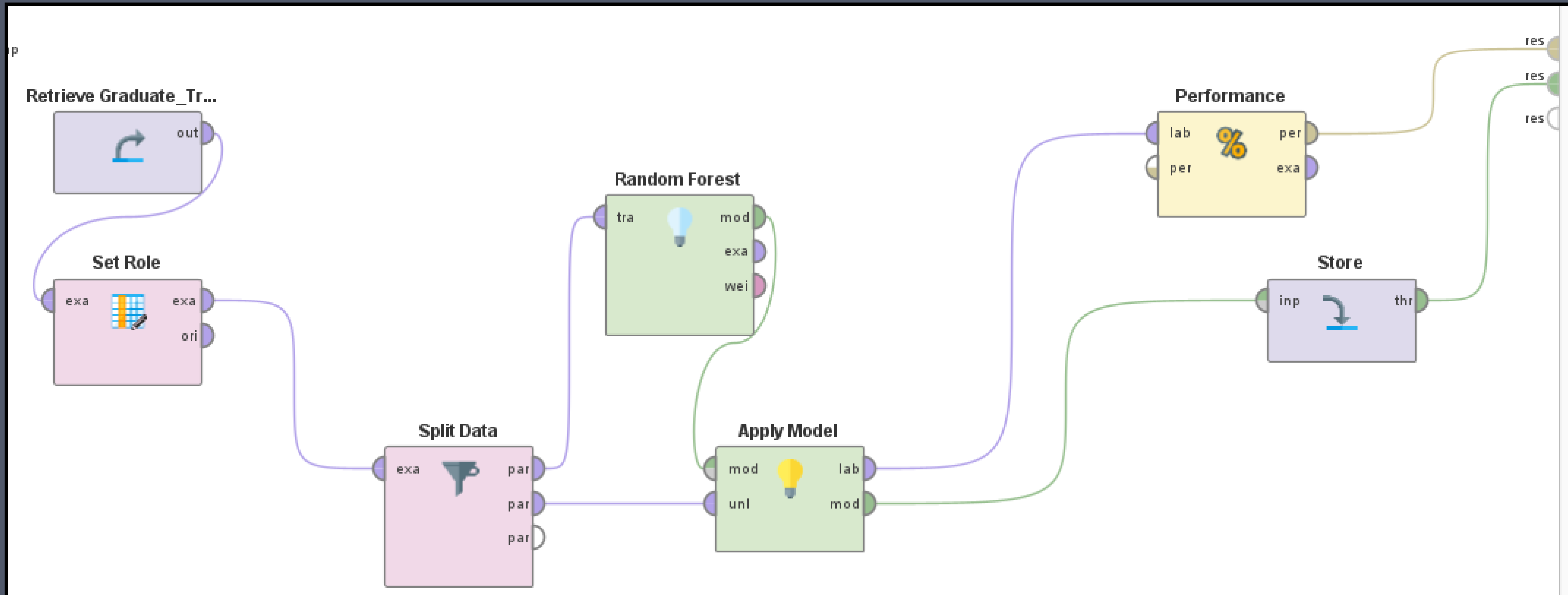
```
Graduation Year : > 2004.500
|  Graduation Year : > 2008.500
|  |  Years of work experience (after graduation): > 15: Low {High=0, Medium=0, Low=2}
|  |  Years of work experience (after graduation): ≤ 15
|  |  |  Years of work experience (after graduation): > 4.500
|  |  |  |  Graduation Year : > 2016.500: Low {High=0, Medium=0, Low=2}
|  |  |  |  Graduation Year : ≤ 2016.500
|  |  |  |  |  Age: > 32.500: Low {High=1, Medium=1, Low=2}
|  |  |  |  |  Age: ≤ 32.500
|  |  |  |  |  |  Gender : = Female: Medium {High=5, Medium=42, Low=1}
|  |  |  |  |  |  Gender : = Male
|  |  |  |  |  |  |  Age: > 26.500
|  |  |  |  |  |  |  |  Age: > 30.500: Medium {High=2, Medium=13, Low=0}
|  |  |  |  |  |  |  |  Age: ≤ 30.500: High {High=18, Medium=16, Low=1}
|  |  |  |  |  |  |  |  Age: ≤ 26.500: Medium {High=0, Medium=2, Low=0}
|  |  |  |  |  |  Years of work experience (after graduation): ≤ 4.500
|  |  |  |  |  |  |  Graduation Year : > 2009.500
|  |  |  |  |  |  |  |  Age: > 21.500
|  |  |  |  |  |  |  |  |  Age: > 29.500: Medium {High=0, Medium=6, Low=0}
|  |  |  |  |  |  |  |  |  Age: ≤ 29.500
|  |  |  |  |  |  |  |  |  |  Years of work experience (after graduation): > 1.500
|  |  |  |  |  |  |  |  |  |  |  Age: > 23.500: Medium {High=7, Medium=168, Low=40}
|  |  |  |  |  |  |  |  |  |  |  Age: ≤ 23.500: Low {High=0, Medium=3, Low=11}
|  |  |  |  |  |  |  |  |  |  |  |  Years of work experience (after graduation): ≤ 1.500: Medium {High=3, Medium=180, Low=155}
|  |  |  |  |  |  |  |  |  |  |  |  |  Age: ≤ 21.500: Low {High=0, Medium=1, Low=9}
|  |  |  |  |  |  |  |  |  |  |  |  |  |  Graduation Year : ≤ 2009.500: Low {High=0, Medium=0, Low=2}
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  Graduation Year : ≤ 2008.500: High {High=5, Medium=3, Low=1}
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  Graduation Year : ≤ 2004.500: High {High=3, Medium=0, Low=0}
```

RANDOM FOREST

Modelling



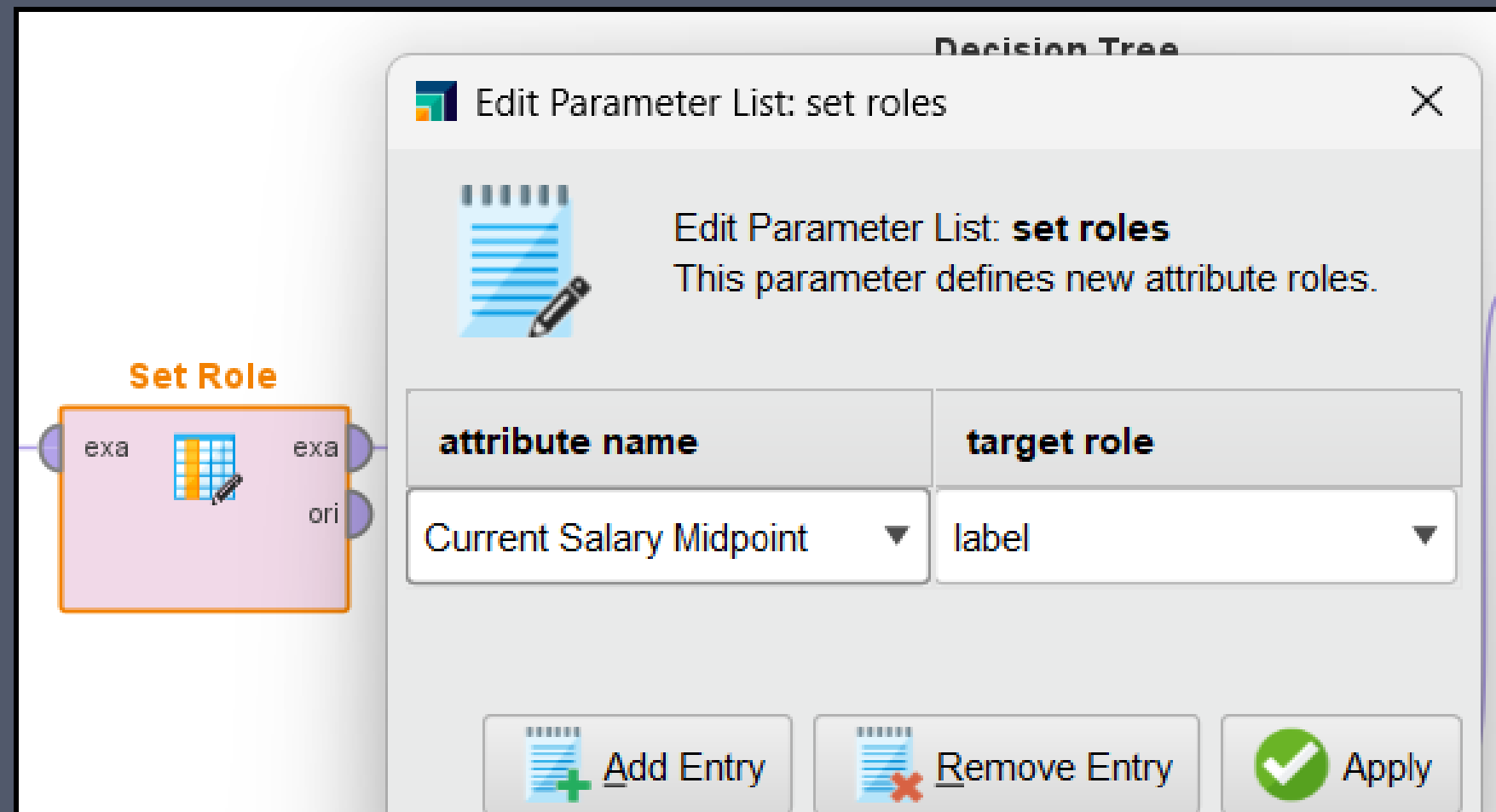
Process in RapidMiner



RANDOM FOREST

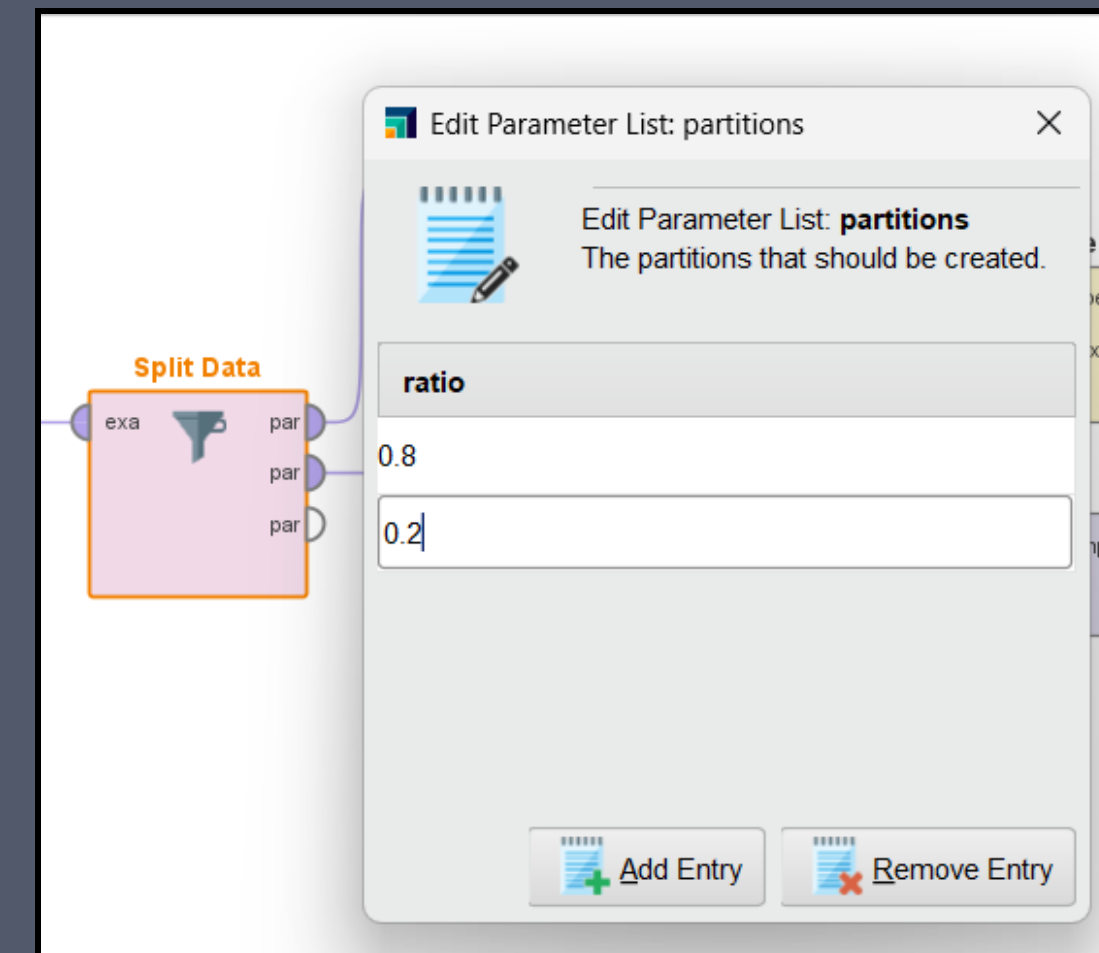
Set Role

Class Label - Current Salary Midpoint



Split Data

Training: Testing
0.8:0.2



RANDOM FOREST

RANDOM FOREST

Evaluation



Confusion Matrix and Accuracy Percentage

accuracy: 64.41%

	true High	true Medium	true Low	class precision
pred. High	2	1	1	50.00%
pred. Medium	9	93	37	66.91%
pred. Low	0	15	19	55.88%
class recall	18.18%	85.32%	33.33%	

RANDOM FOREST

Comparison between Decision Tree and Random Forest

Decision

tree

accuracy: 62.15%

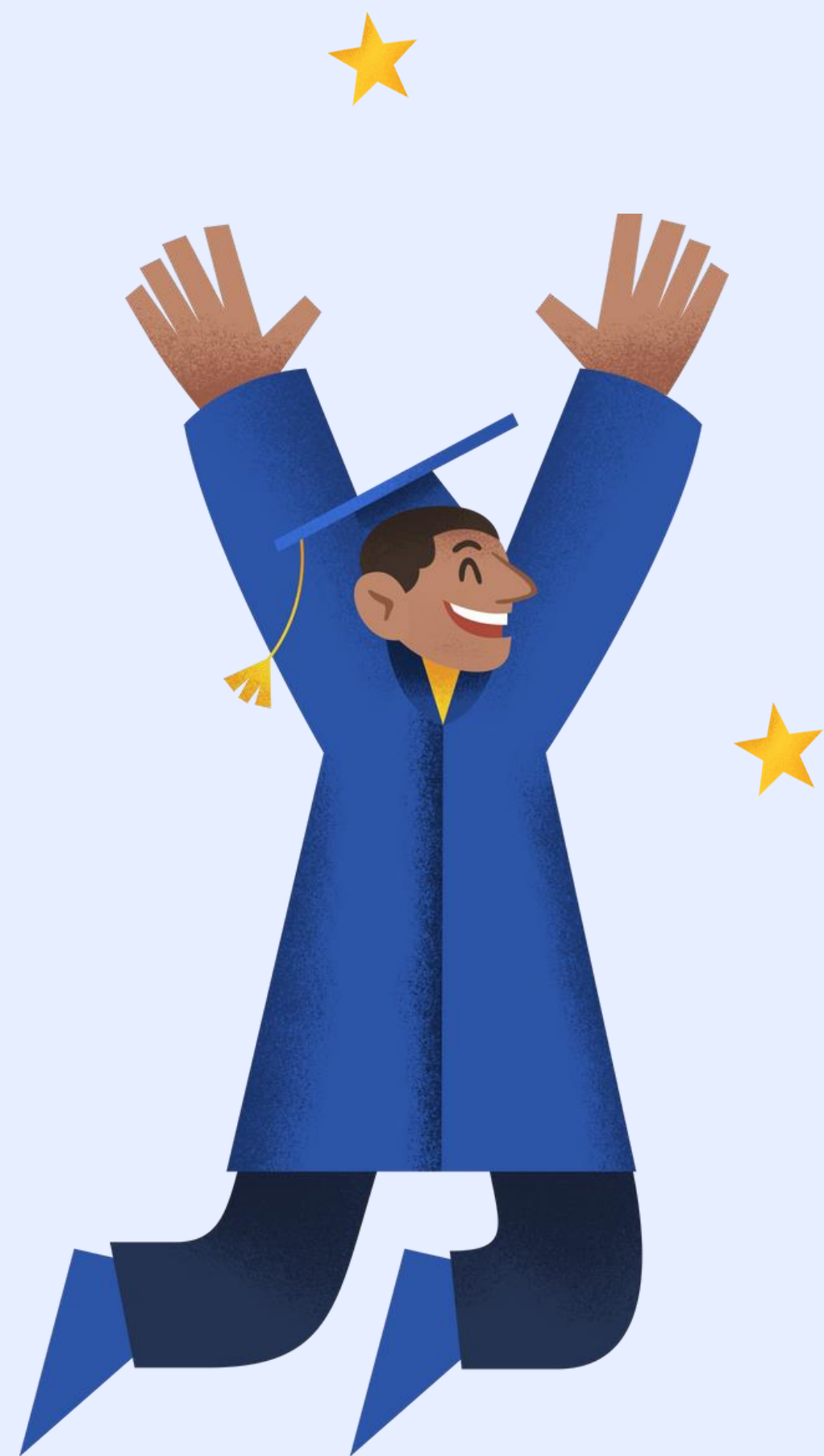
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Random

Forest

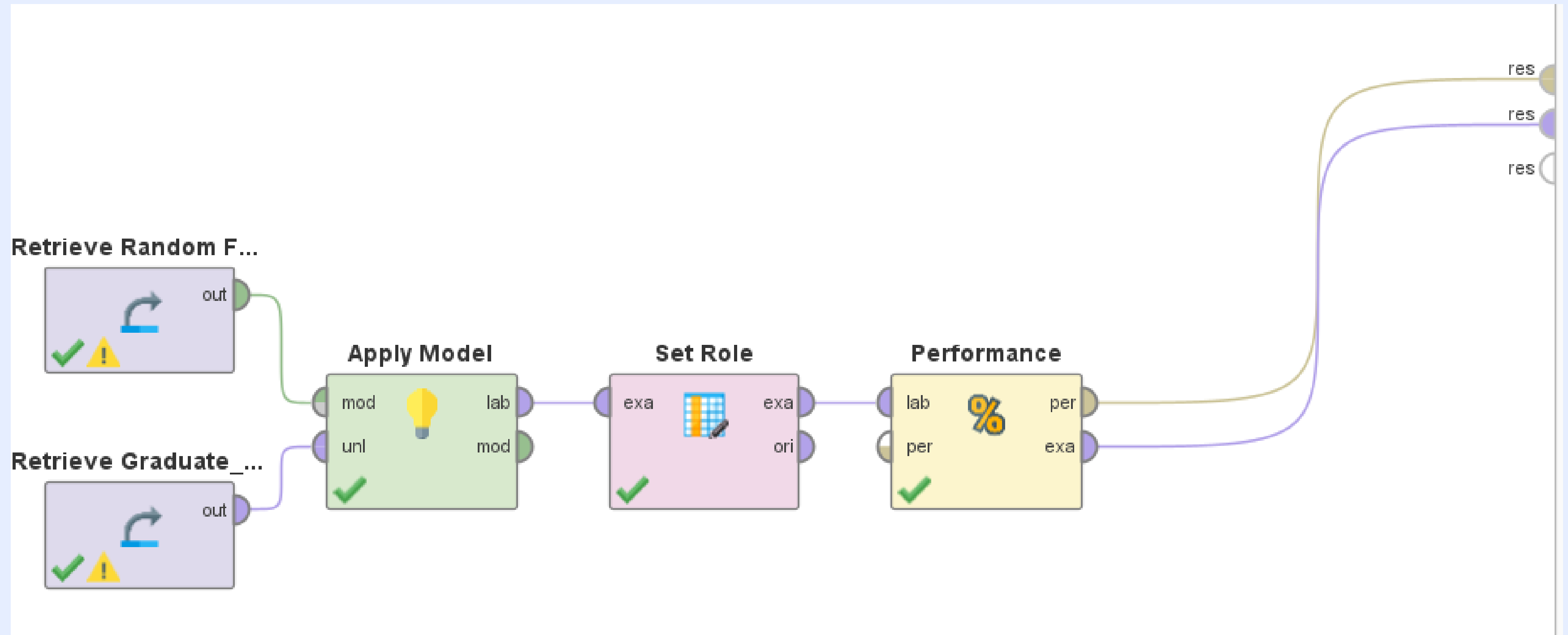
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| **Deployment**

DEPLOYMENT OF RANDOM FOREST

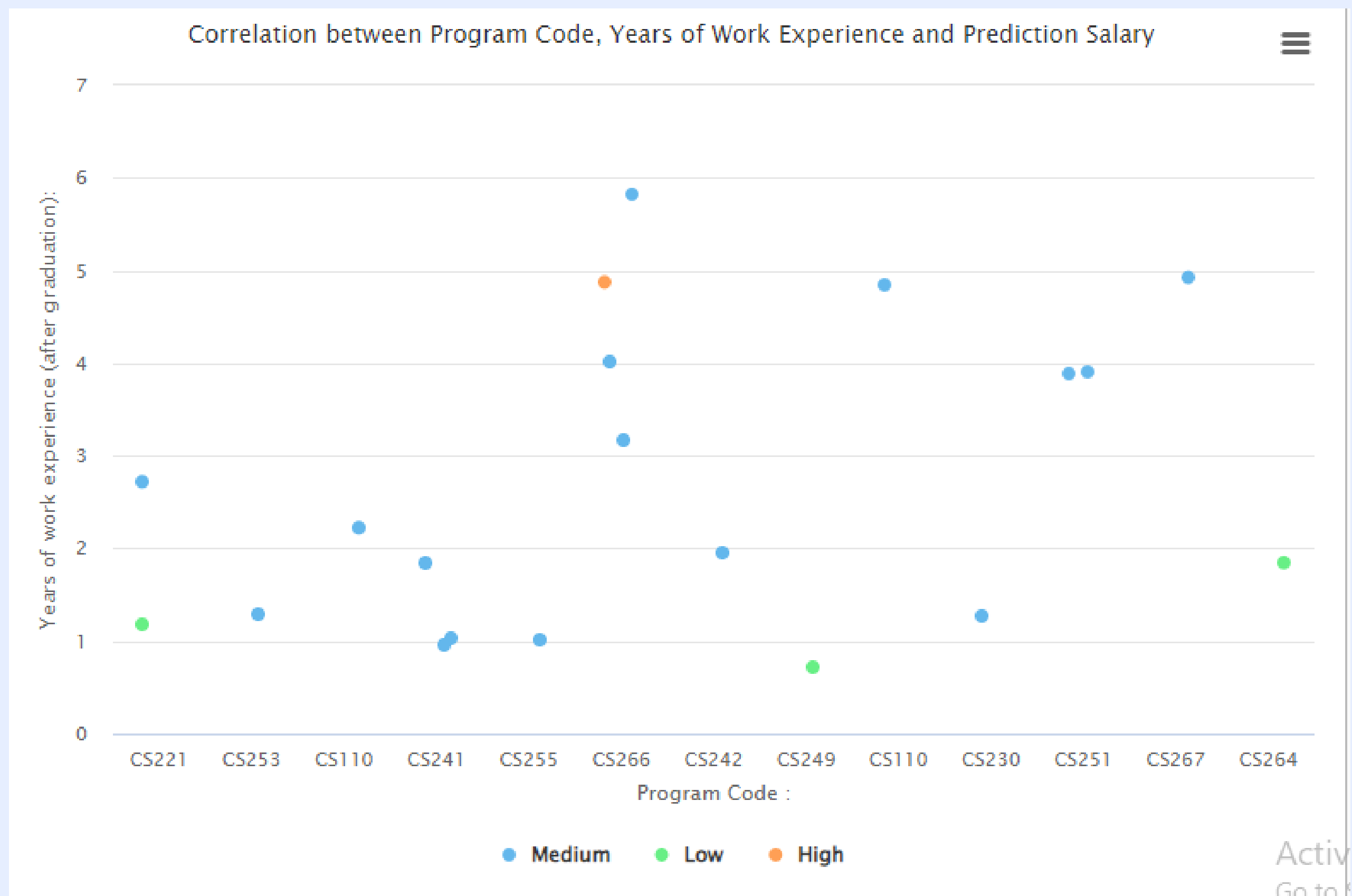


PERFORMANCE RESULT

accuracy: 75.00%

	true Medium	true Low	true High	class precision
pred. Medium	13	2	1	81.25%
pred. Low	2	1	0	33.33%
pred. High	0	0	1	100.00%
class recall	86.67%	33.33%	50.00%	

Precision:
71.53%
Recall: 56.67
F1-Score: 0.6782



LOW: < RM3000
MEDIUM: RM3000 - RM5000
HIGH: > RM5000 HIGH

RECOMMENDATION

Tailored Career Guidance for Graduates

Skill Enhancement Programs

Data Collection Enhancement

Conclusion

Random Forest model has higher accuracy of 62.15%

Students are able to make more informed decisions about their educational paths





THANK YOU
and have a nice day!

