

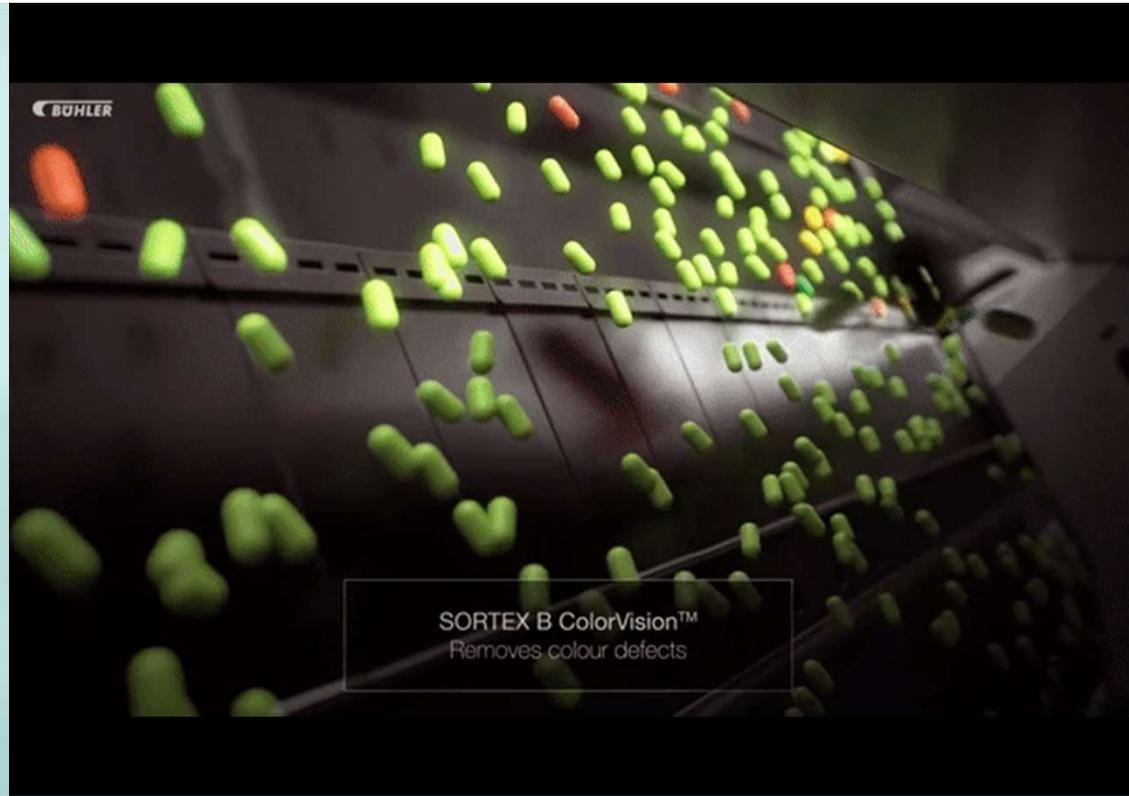
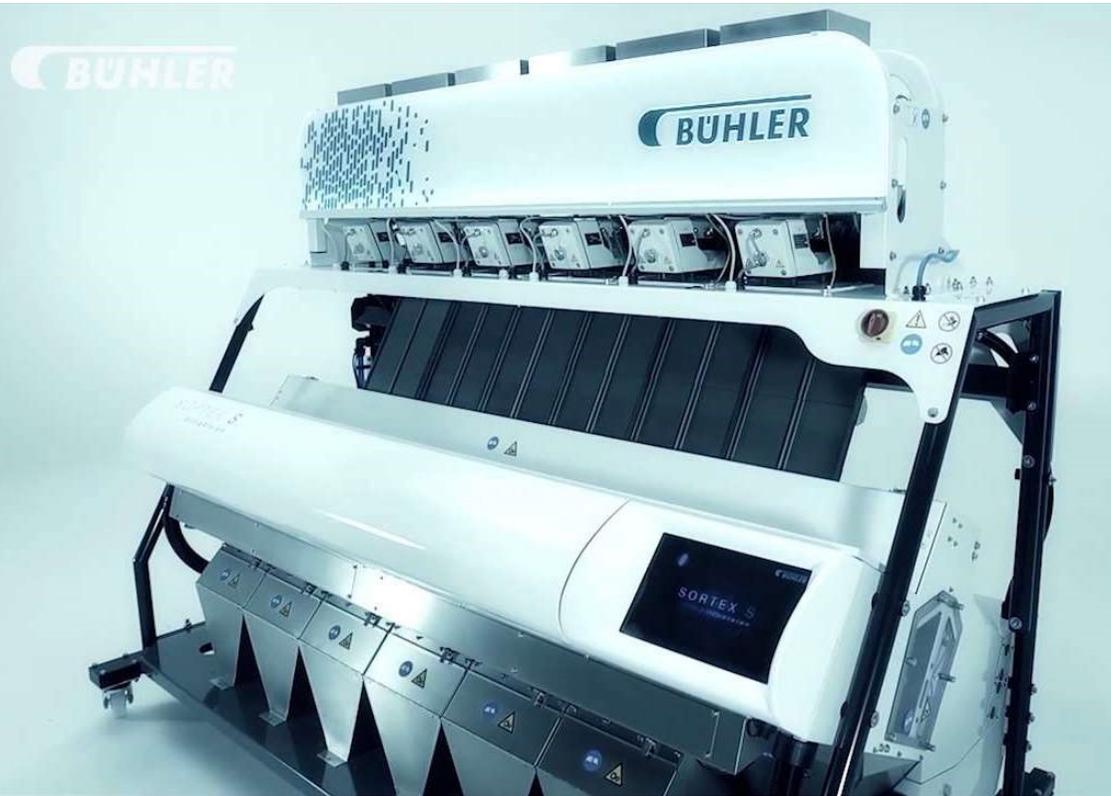
Industrialization of rice grain recognition

KNIME Development and industrialization

Berlin, March 20th 2019

HES-SO Valais - Switzerland

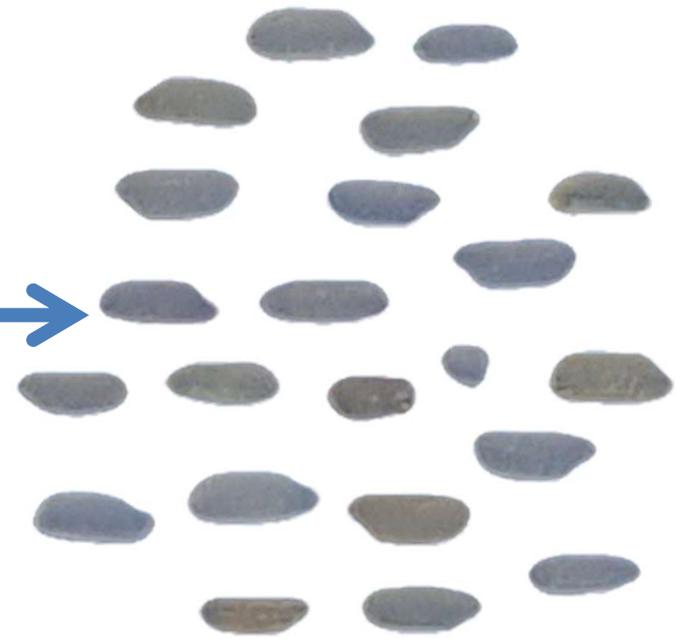
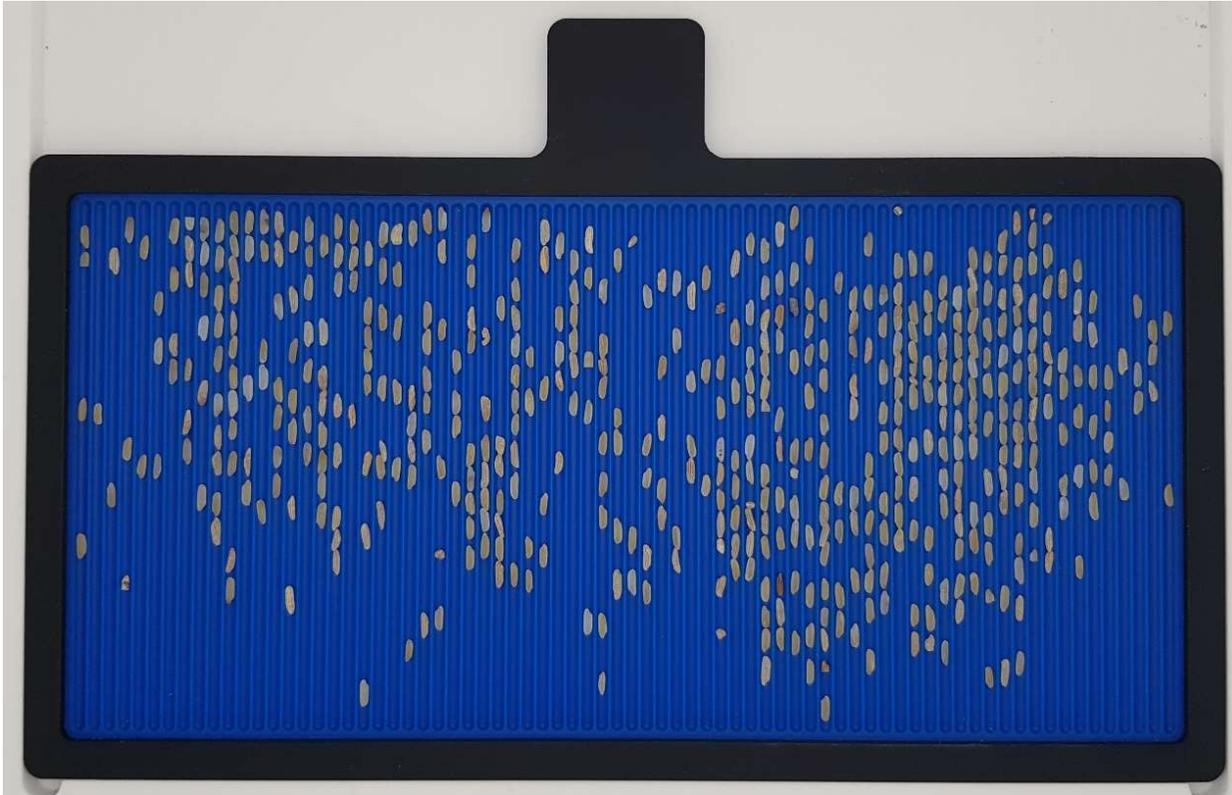
Dominique Genoud – dominique.genoud@hevs.ch



Sorts up to 20 tons per hour → up to 40% of good grains spared
Rice supplies half of the world needs for food

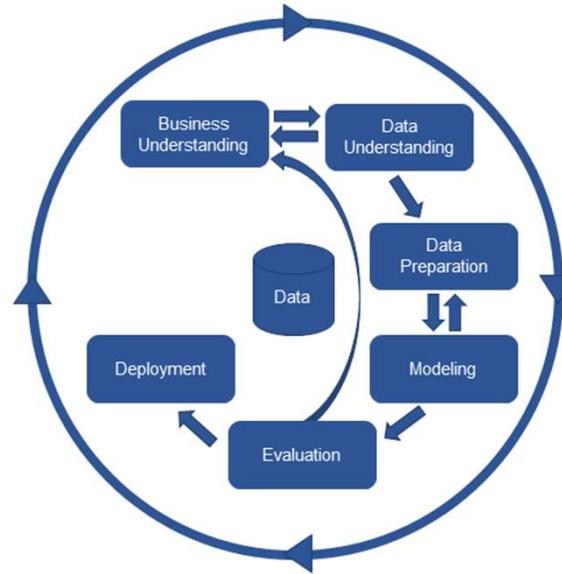


120'000 rice varieties

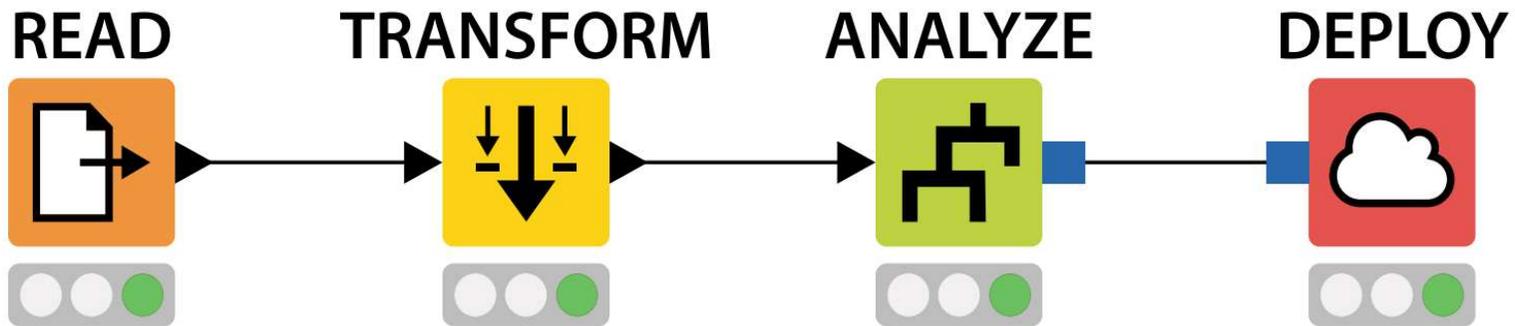


Conventional image processing: Color discrimination | Size of the grain | A lot of manual setup

Methodology



CRISP-DM

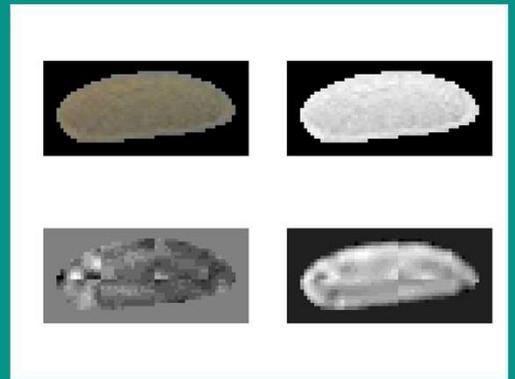
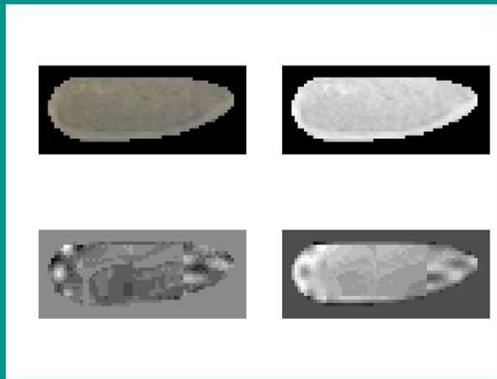


KNIME

Feature Extraction

S filename	D Column...	D Column...	I Column...	D Column...				
training_segmented\CA_23_1_GalaxyS8\CA_23_1_GalaxyS8_1.png	0.738	0.834	67	3.68	0.769	0.365	147.864	9.39
training_segmented\CA_23_1_GalaxyS8\CA_23_1_GalaxyS8_10....	0.896	0.884	40	3.222	0.783	0.448	136.311	9.417
training_segmented\CA_23_1_GalaxyS8\CA_23_1_GalaxyS8_2.png	1.071	0.866	45	3.25	0.732	0.414	135.466	9.346
training_segmented\CA_23_1_GalaxyS8\CA_23_1_GalaxyS8_3.png	1.067	0.917	34	2.7	0.766	0.574	122.255	10.948
training_segmented\CA_23_1_GalaxyS8\CA_23_1_GalaxyS8_4.png	0.834	0.928	28	2.273	0.82	0.82	145.127	9.317
training_segmented\CA_23_1_GalaxyS8\CA_23_1_GalaxyS8_5.png	0.781	0.863	65	4.25	0.762	0.299	146.002	8.49
training segmented\CA_23_1_GalaxyS8\CA_23_1_GalaxyS8_6.png	1.064	0.816	41	2.864	0.699	0.478	148.126	10.653

- External python script
- ~100 features
 - Color
 - Spatial
 - Geometric
 - 6 Different phone models



Pre-processing

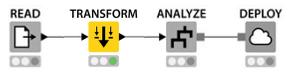


```

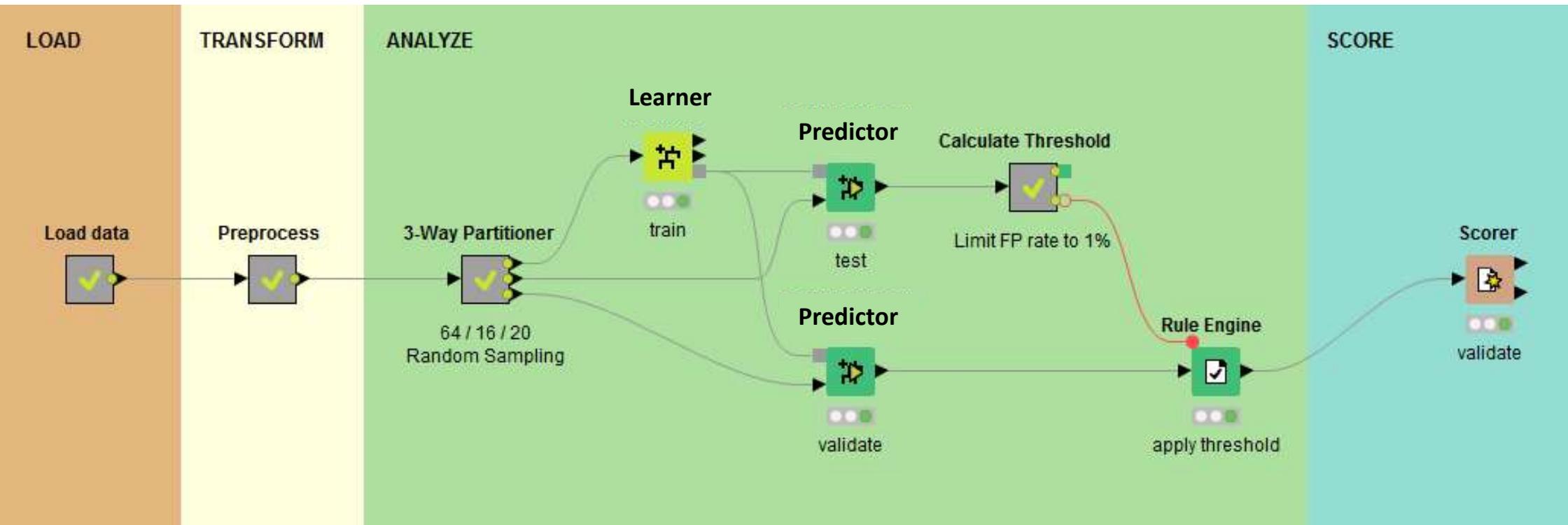
$Sample$ = 1 => "sample"
$Sample$ = 2 => "sortex rejects"
$Sample$ = 3 => "damage (other)"
$Sample$ = 4 => "red"
$Sample$ = 5 => "damage (pecky)"
$Sample$ = 6 => "damage (other)"
$Sample$ = 7 => "damage (other)"
$Sample$ = 8 => "heat damage"
$Sample$ = 9 => "damage (stain)"
$Sample$ >= 10 AND $Sample$ <= 12 => "good"
    
```

	S samplename	S class
\$8_147.png	good	accept
\$8_316.png	good	accept
\$8_15.png	damage (stain)	reject
\$8_183.png	heat damage	reject
\$8_3.png	damage (other)	reject
\$8_75.png	good	accept
_1.png	heat damage	reject
\$8_68.png	good	accept

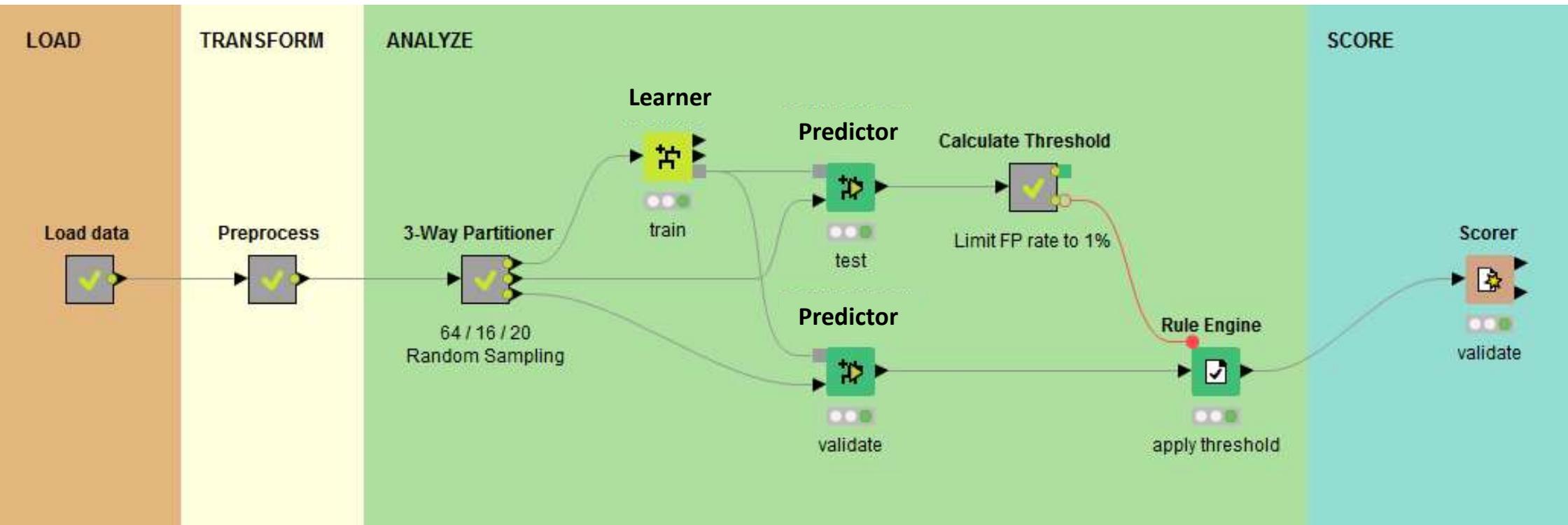
First loop → separate good grains from defect grains



Model creation



Model creation → learn / predict / compute threshold

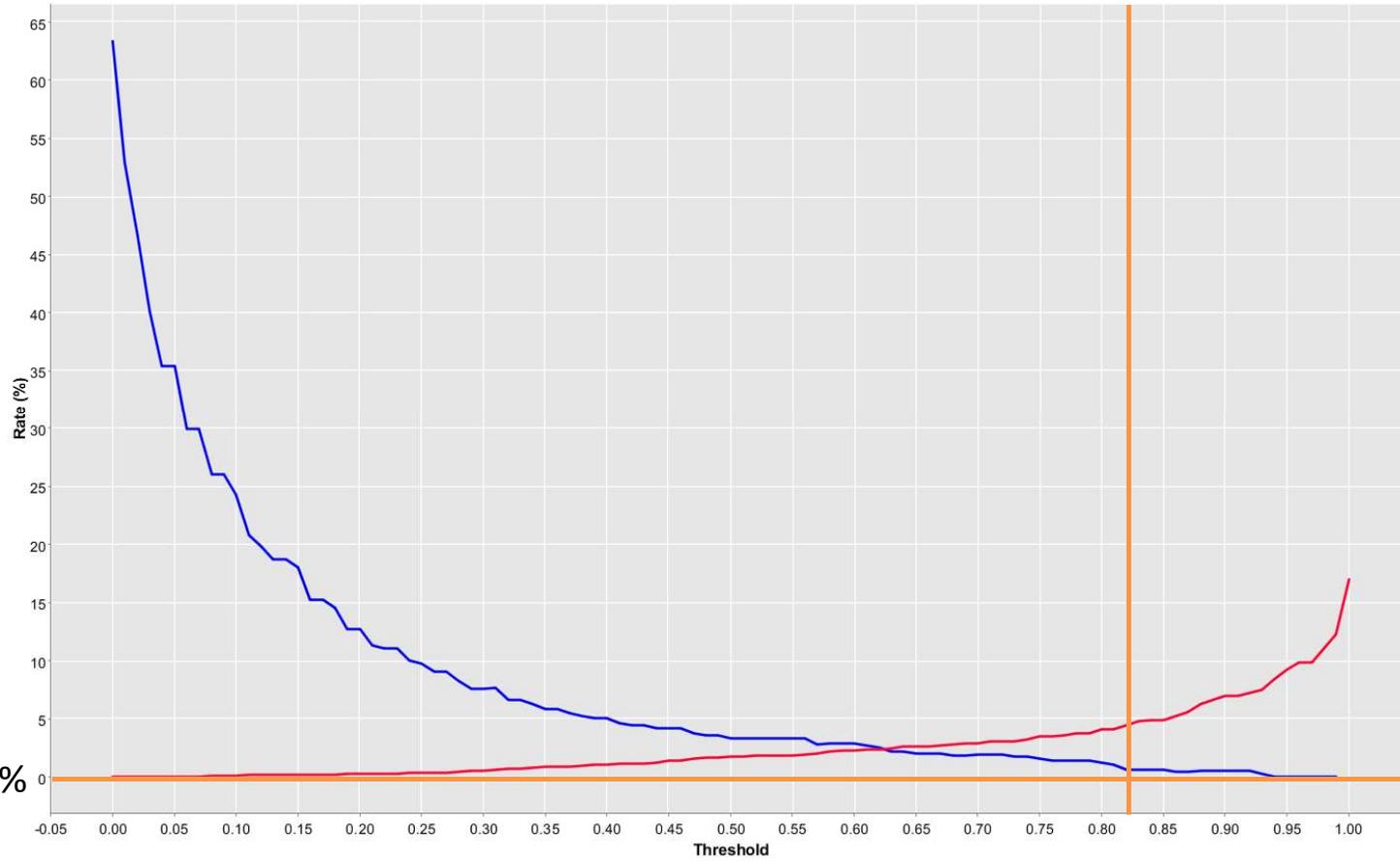


Training

FP and FN rates vs Threshold

Threshold at 0.82

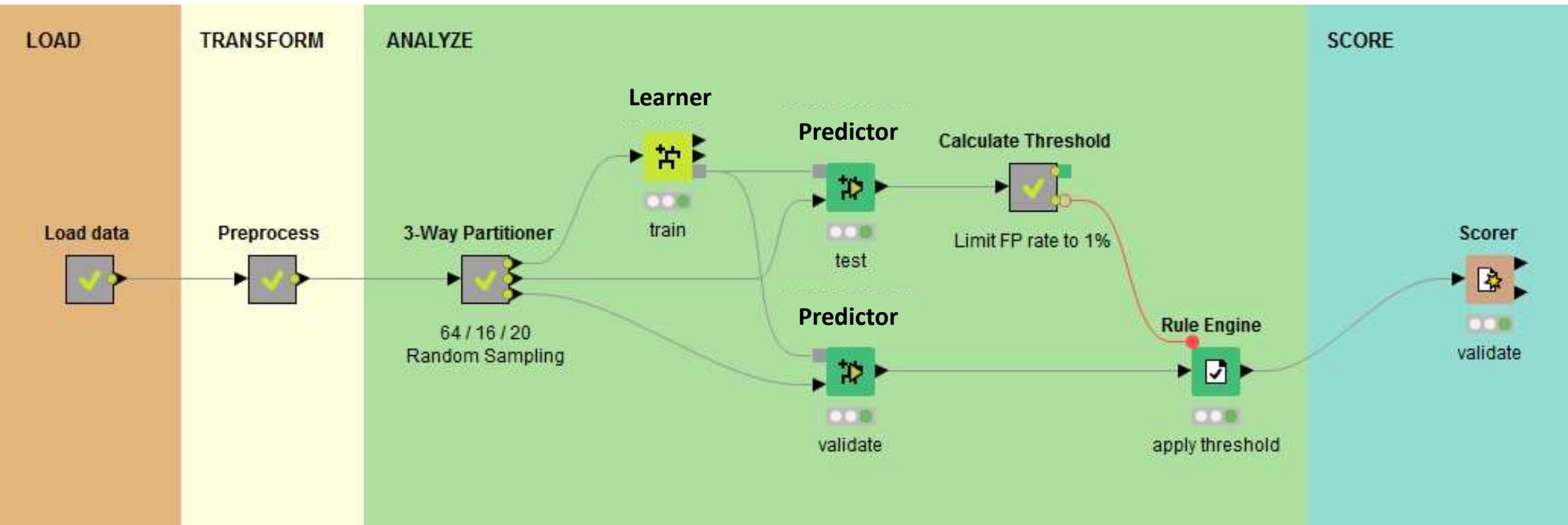
FP rate at 1%



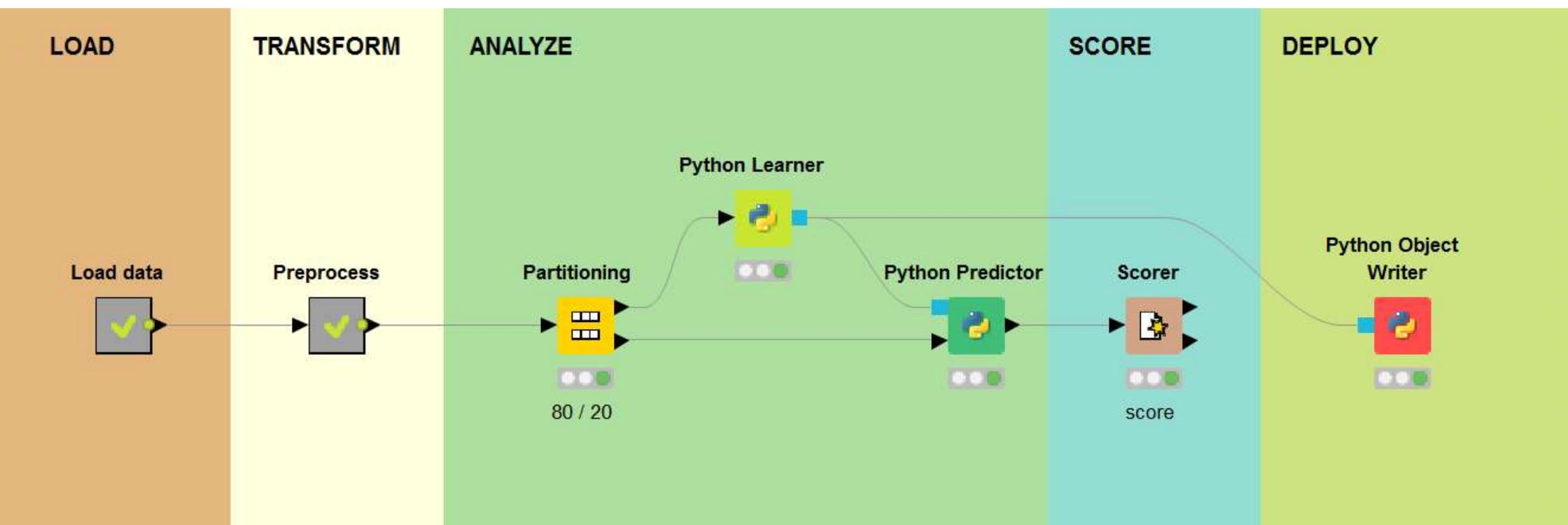
■ FN Rate
■ FP Rate



Model creation → validate performance with pre-computed threshold



Industrialisation → customer constraints (python and Azure server)



Industrialisation

```

from sklearn.ensemble import RandomForestClassifier
from sklearn import preprocessing

#set columns
columns = sorted(input_table.columns.drop(['filename', 'samplename', 'class']))

#label encoder

# Designate features to learn from
features = input_table[columns]

# Create a random forest classifier
output_model = {"columns" : columns, "labels" : sorted(input_table['class'].unique()), "threshold" : 0.82}

# Choose class labels column
label_encoder = preprocessing.LabelEncoder()
label_encoder.fit(output_model['labels'])
labels = label_encoder.transform(input_table['class'])

output_model['model'] = RandomForestClassifier(n_estimators=100, n_jobs=-1, criterion='entropy')

# Fit the model using the features and labels
output_model['model'].fit(features, labels)

```



```

import preprocessing
import numpy as np

# Load data from the model
new_data = input_table[input_model['columns']]

# Prepare data to output
output_table = input_table.copy()

# Apply existing model + new data
predictions = input_model['model'].predict_proba(new_data)

# Calculate probabilities for classes
probabilities = input_model['model'].predict_proba(new_data)

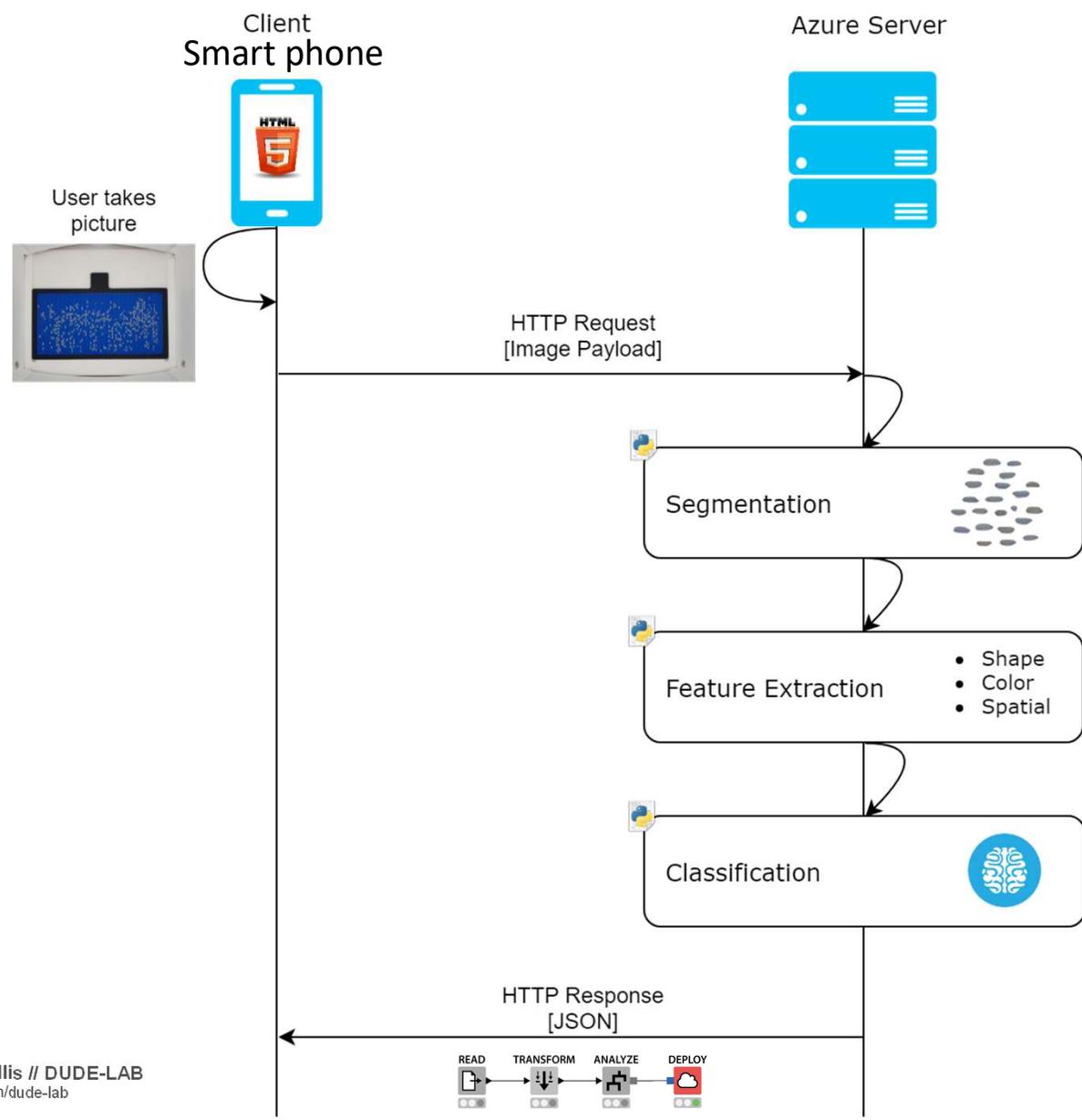
# Output probabilities for classes
output_table['probabilities'] = probabilities

# Output class as string
label_encoder = preprocessing.LabelEncoder()
label_encoder.fit(input_model['labels'])
output_table['prediction'] = label_encoder.inverse_transform(predictions)

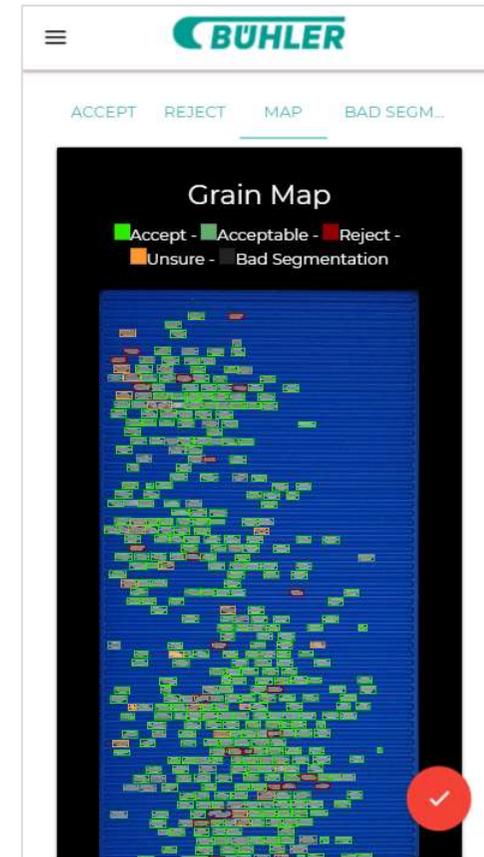
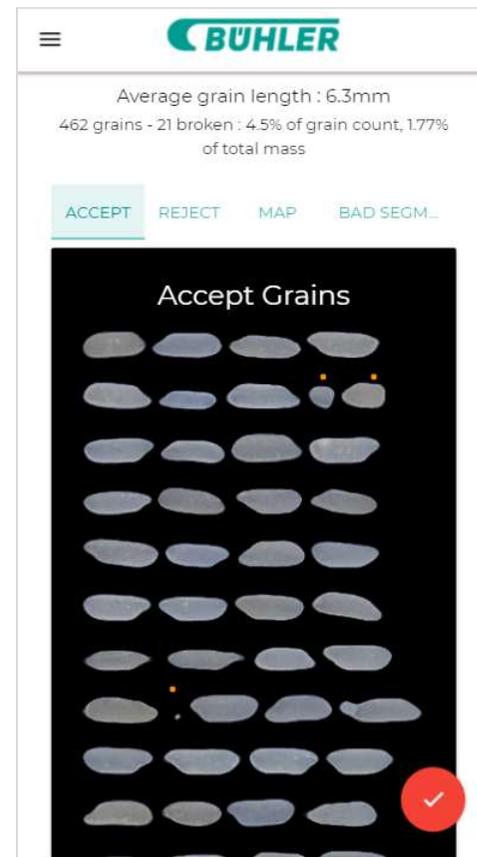
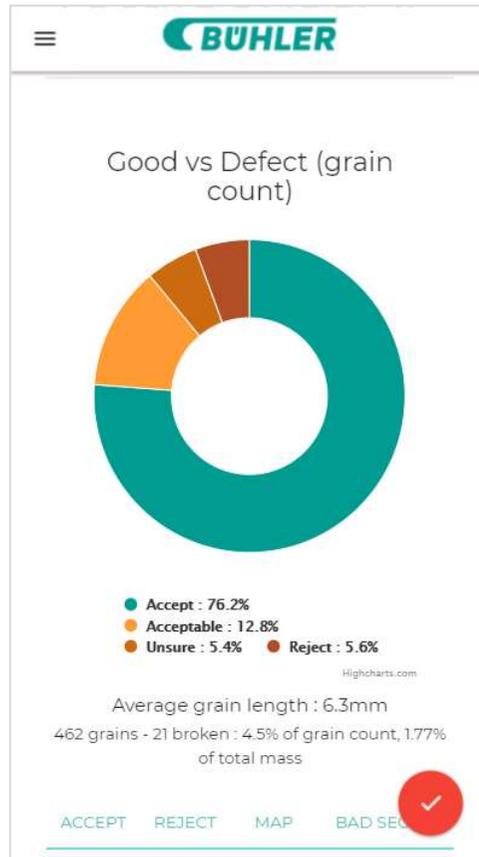
# (taking threshold into account)
output_table['prediction'] = [0 if k < input_model['threshold'] else 1 for k in probabilities[:,0]]

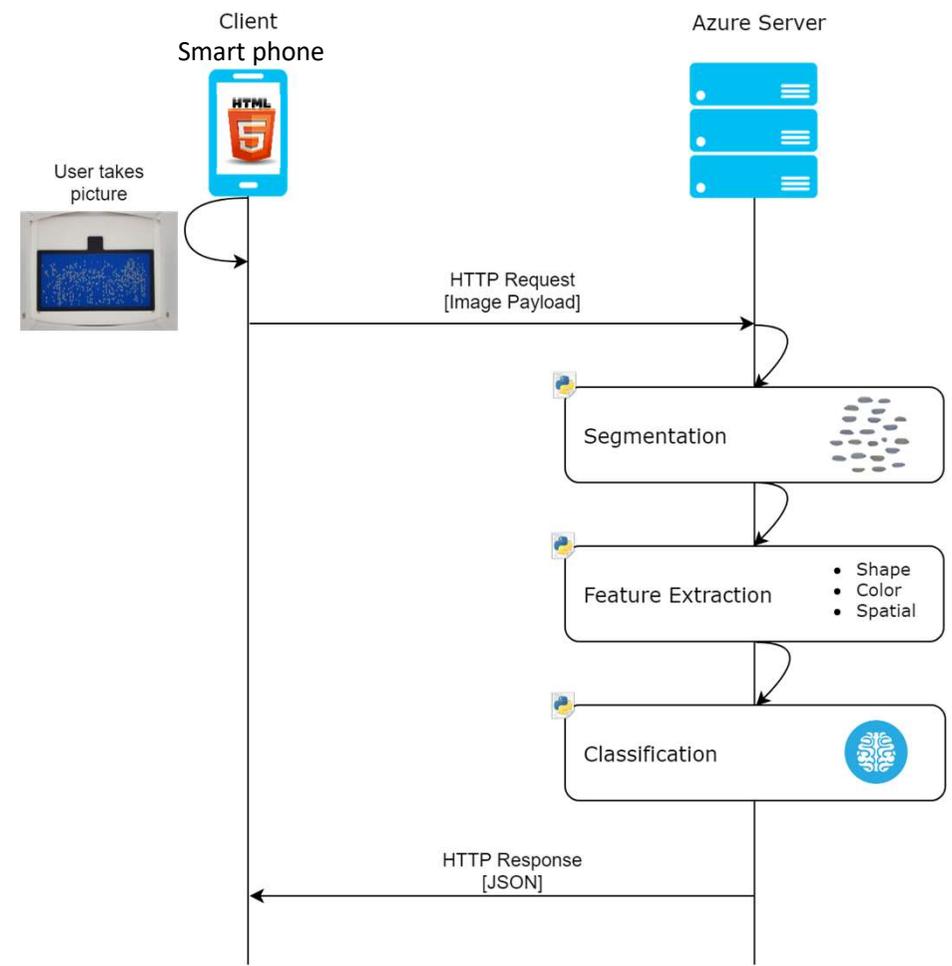
```



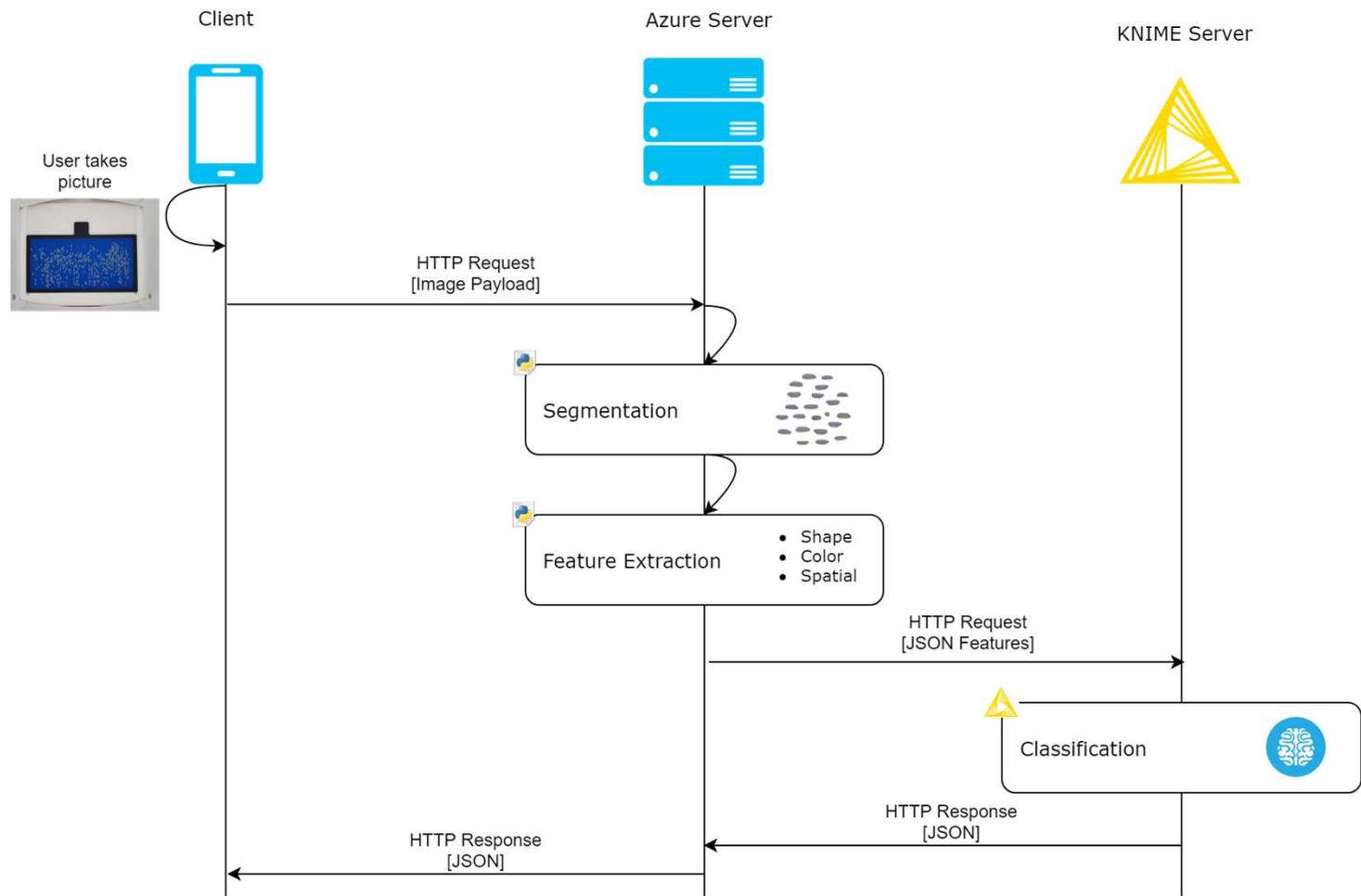


Recognition and statistics

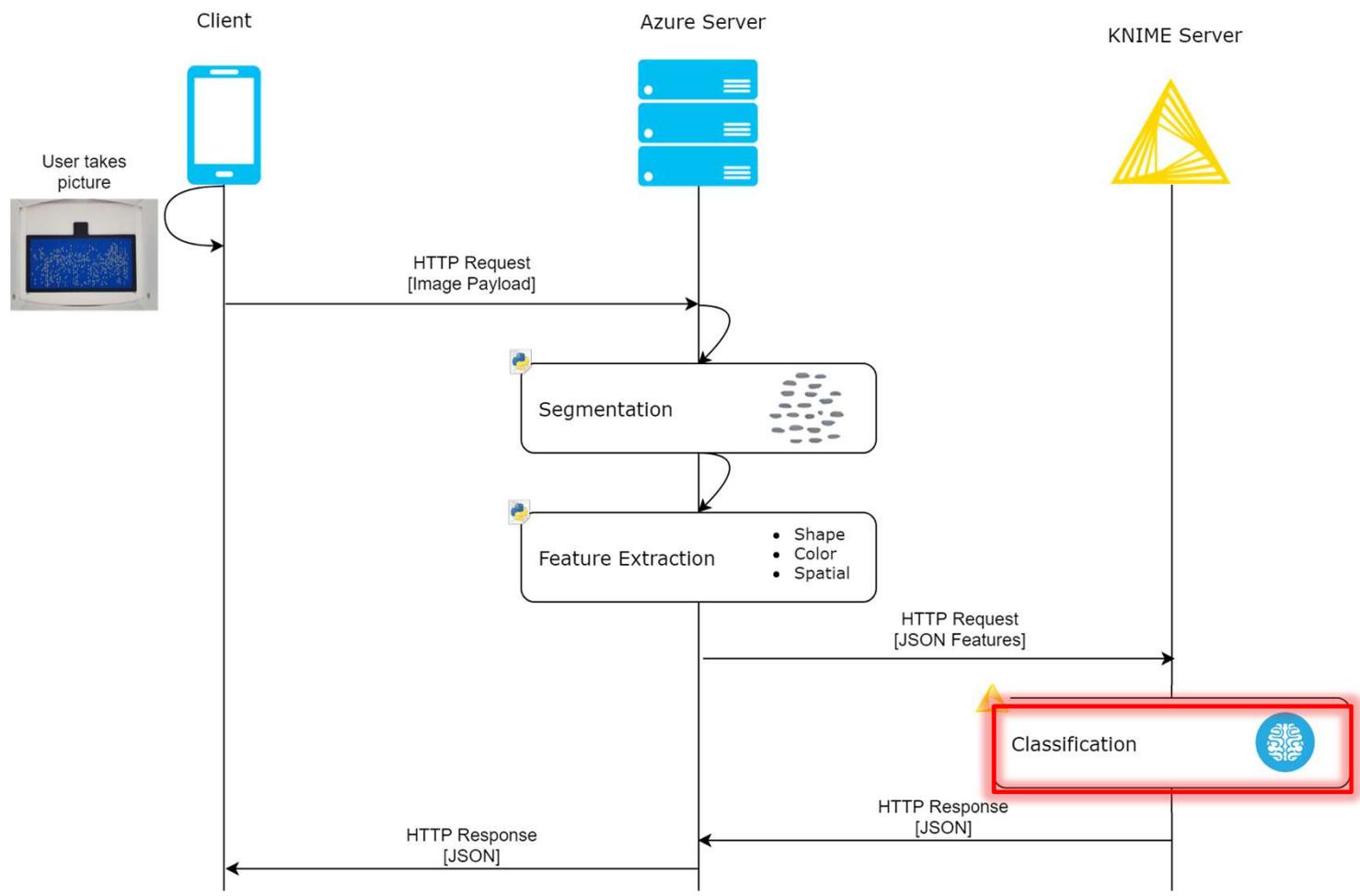




Current implementation, with customer constraints

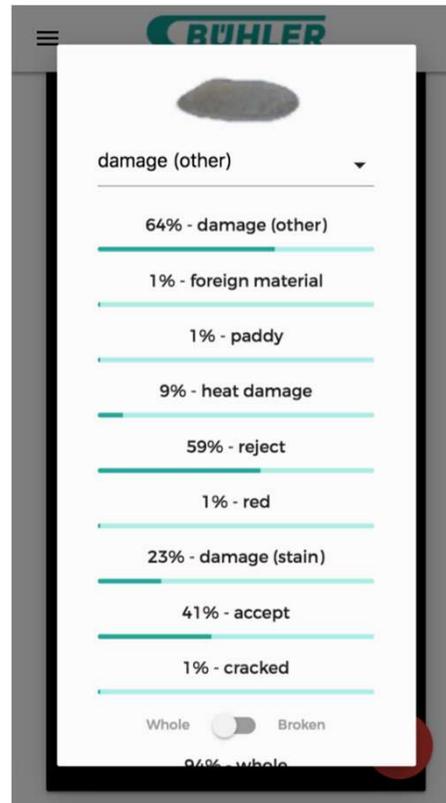
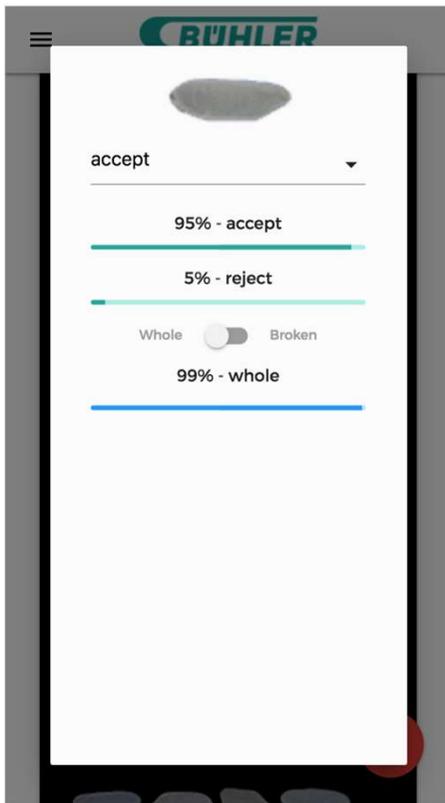


Further performance improvements using KNIME server



Further performance improvements using KNIME server

TotalSense.ai results



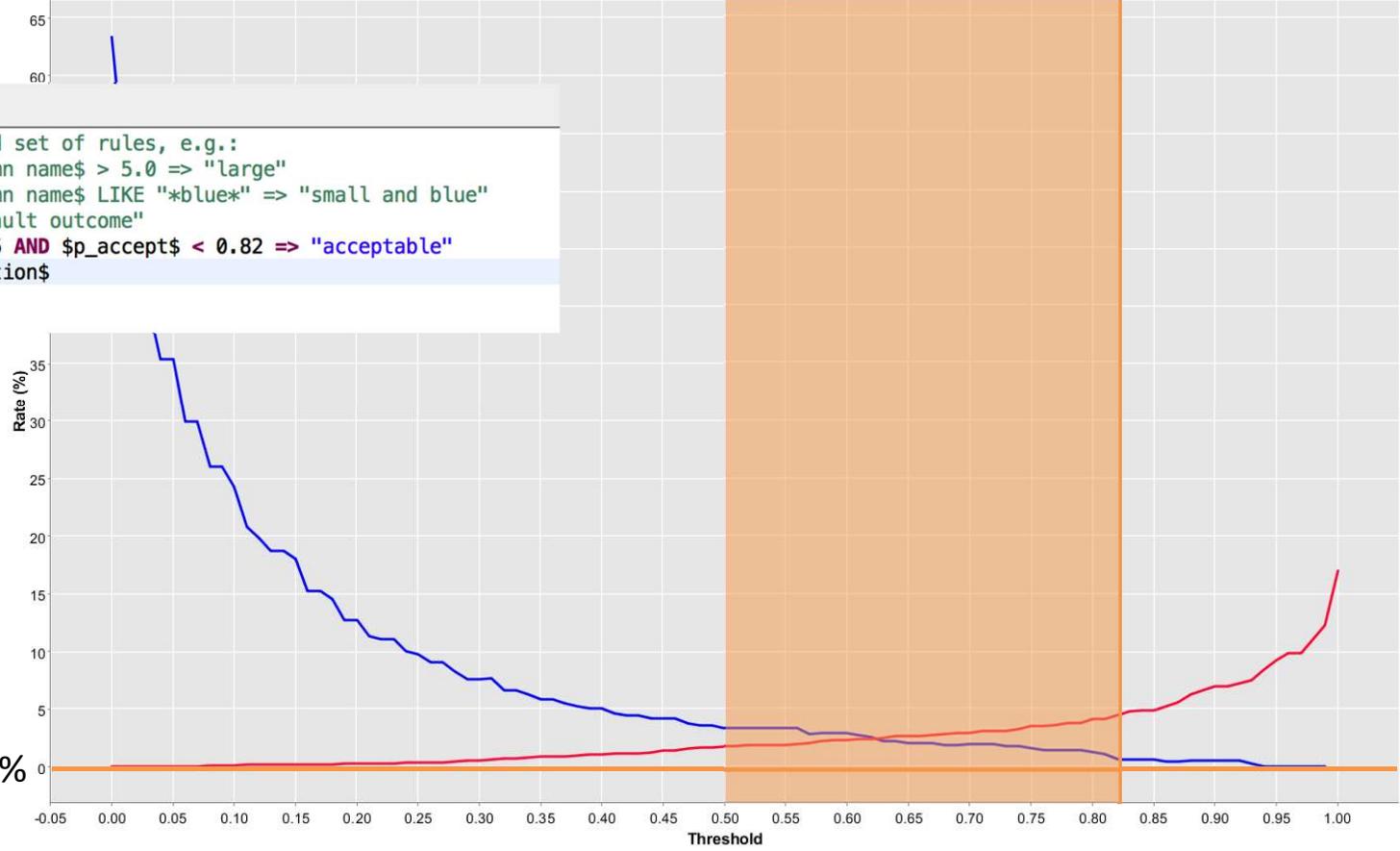
Training

FP and FN rates vs Threshold

Threshold at 0.82

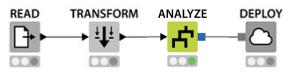
```

? 1 // enter ordered set of rules, e.g.:
? 2 // $double column name$ > 5.0 => "large"
? 3 // $string column name$ LIKE "*blue*" => "small and blue"
? 4 // TRUE => "default outcome"
S 5 $p_accept$ > 0.5 AND $p_accept$ < 0.82 => "acceptable"
S 6 TRUE => $prediction$
  
```



FP rate at 1%

■ FN Rate
■ FP Rate



Thank you for your attention.

Any questions?

