



# Rab Protein Classification

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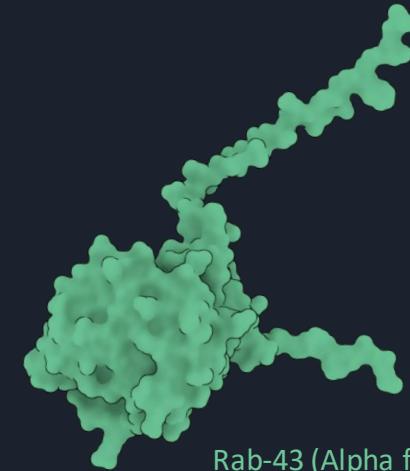
# Goal of our project



Classify the Rabs into their group



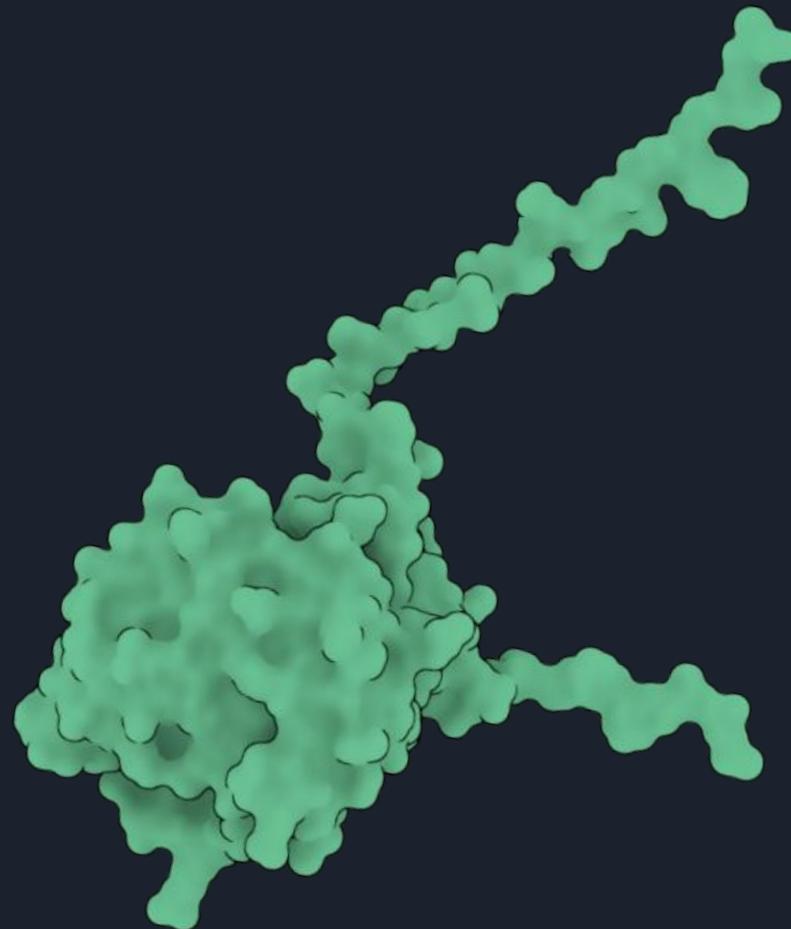
```
[ ] from sklearn.preprocessing import StandardScaler  
sc=StandardScaler()  
X_train=sc.fit_transform(x_train)  
X_test=sc.transform(x_test)  
  
[ ] from sklearn.neighbors import KNeighborsClassifier  
k = 10  
classifier = KNeighborsClassifier(n_neighbors = k, metric = "minkowski", p=2)  
classifier.fit(X_train, y_train)  
y_pred = classifier.predict(X_test)  
  
[ ] from sklearn.metrics import accuracy_score  
print(accuracy_score(y_test, y_pred))  
cm = confusion_matrix(y_test, y_pred)  
print(cm)  
print(classification_report(y_test, y_pred))  
  
[ ] from sklearn.model_selection import train_test_split  
from sklearn.ensemble import RandomForestClassifier  
from sklearn.metrics import confusion_matrix, accuracy_score  
  
rf = RandomForestClassifier(n_estimators=100, random_state=42)  
rf.fit(X_train, y_train)  
y_pred = rf.predict(X_test)  
  
# calculer et afficher la matrice de confusion et la précision  
conf_matrix = confusion_matrix(y_test, y_pred)  
accuracy = accuracy_score(y_test, y_pred)  
  
print("Matrice de confusion : \n", conf_matrix)  
print("Précision : (accuracy) ")
```



Rab-43 (Alpha fold)



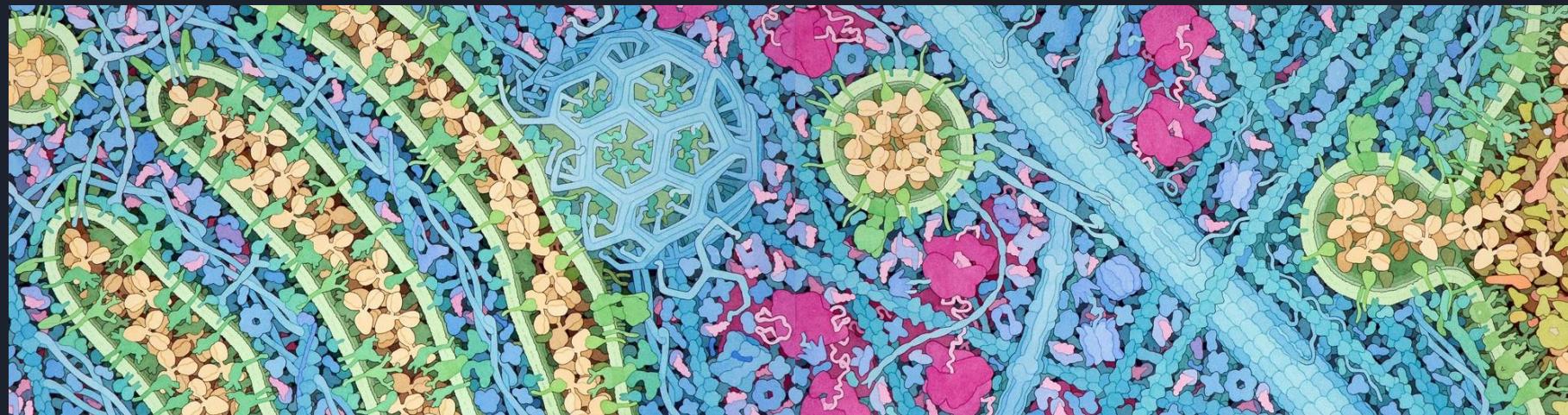
# What are Rab proteins?





# What are Rab proteins?

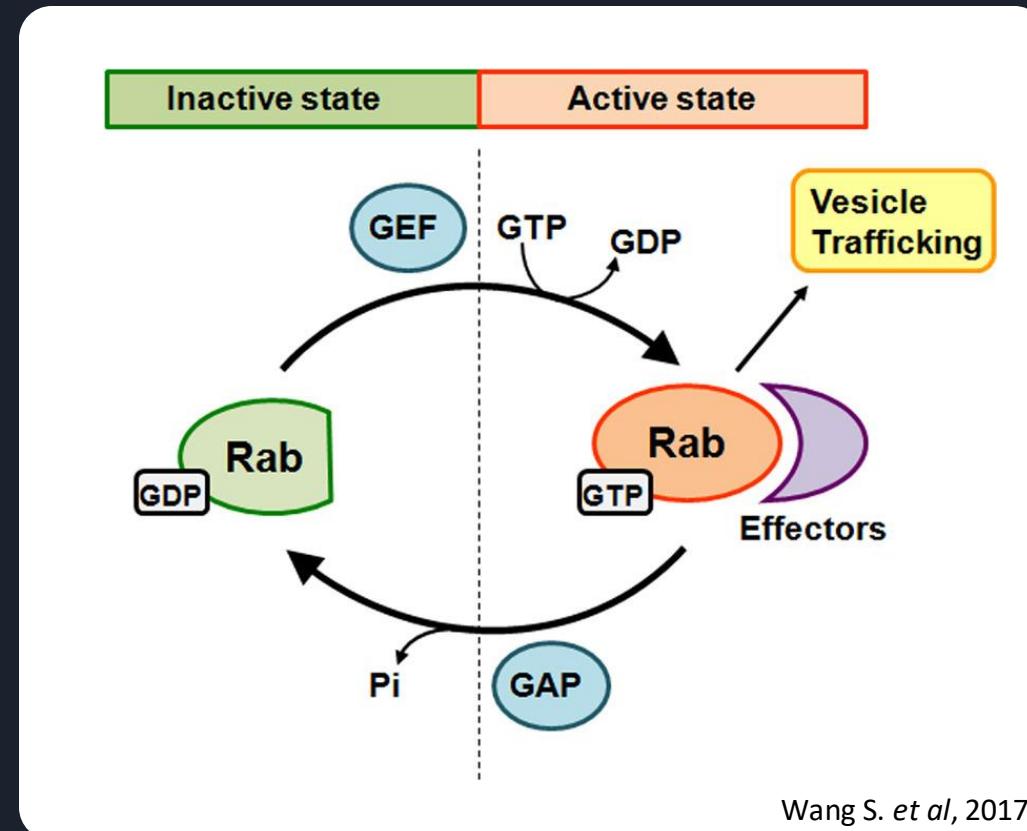
GTPase with functions in tethering in **vesicular trafficking**





# What are Rab proteins?

GTPase →  
exchange GDP against  
GTP to activate effectors

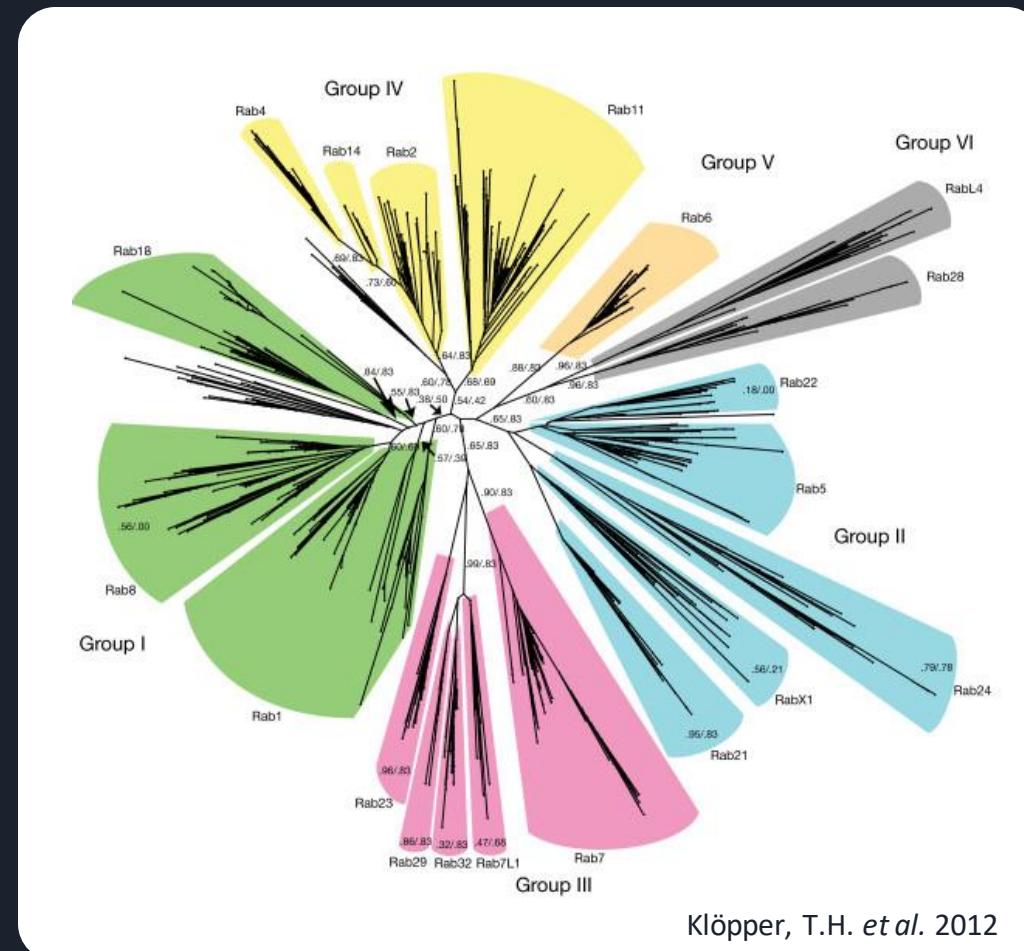


Wang S. et al, 2017



# What are Rab proteins?

Large **family** of proteins



Klöpper, T.H. et al. 2012



# ... and why study them?

## **Griscelli syndrome**

- partial albinism
- immunological defects
- primary neurological dysfunctions

## **Choroideremia**

- Blindness

## **Neurological disorders**



Rab-associated  
**diseases**



# Pipeline and methods

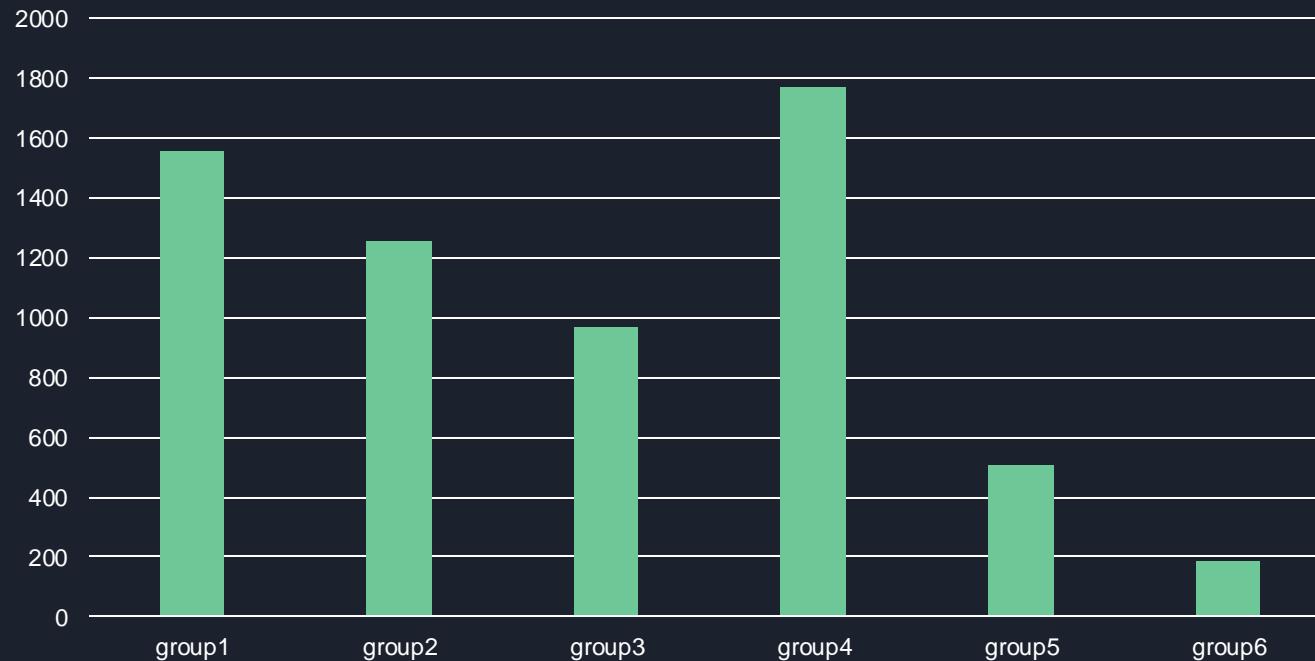




# What does our data look like?

```
>PoTr_Rab1C
MNPEYDYLFKLLLIGDSGVGKSCLLRFADDSYESTI
GVDFKIRTVEQDGKTIKLQLQIWTAGQERFRTITSSYYRAH
GIIIVVYDVTDQESFNNVKQWLNEIDRYASENVNKLLVGNC
DLTANKVVSYETAKAFADEIGIPFMETSAKNATNVEQAFMA
MAAEIKNRMASQPAMNNARPPTVQIRGQPVNQKSGCCSS
>PaTe_Rab1
MSLQQEYDYLFKILLIGNSAVGKSSLRLFADNVFNESFLP
TIGVDFKIRTFDLNGKTVKLQLQIWTAGQERFRTITNSYYKG
AHGIIILVYDVTDKQSFKDVENWLAEVEKYANENVVRVLVGN
KVDLESKREVTSEEGKELADSLNIRFIETSAKNSSNVEKAFT
ITLANEIKAKVAKSSEAIPVKTGPRITPDQQQNTVKDTGCC
>BrMa_Rab1
MVSINPEYDYLFKLLLIGDSGVGKSCLLRFADDTYTESIYI
STIGVDFKIRTIDLNGKTIKLQLQIWTAGQERFRTITSSYYR
GAHGIIVVYDITDQESFNNVKQWLQEIDRYACENVNKLLVG
NKCDLIIRRRAVEHSAKEYADQLGIPFLETSAKSSTNVEQA
FLTMASEIKNRMGPPIQQVGTGPSVRIGGSQPVNEKKSGGCC
>CaA1_Ypt1
MNNEYDYLFKLLLIGDSGVGKSCLLRFADDTYTPDYISTI
GVDFKIRTIELDGKTIKLQLQIWTAGQERFRTITSSYYRAH
GIIIVVYDVTDQESFNNVKQWLQEIDRYATGGVMKLLVGNC
DLSDKKIVEYTAKEFADALDIPFLETSALSSTNVEQAFYT
MARQIKAQMNTNNANAGNAANAKGKSNVNLRGESLTSNQSNS
CC
>PaTe_Rab1
MIKEYDYLFKLVIIGNSGVGKSSLRLFADDQFSESYLTTI
GVDFRFRTLPIDGKNVKLQLQIWTAGQERFRTITSAYYKGAD
GIVMVYDVTQGQSFDDIDKFWLHEVESYGEKVNQLLIIGNK
NDLDEQKQVETSKAEEYCKSHNMLFMECSAKTADHVNNAAFL
ELSRKLMACKDASQPPKTTNTTSNASQQSQSRGQTNTNTQQ
SKQLSAGNTNQKKQKDGGCC
```

Number of Rab proteins per group



Unequal distribution



# What did we do?

Protein sequences

Features

Machine Learning

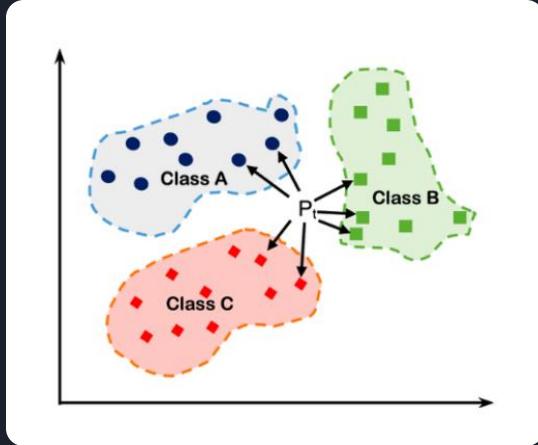
```
>RAB 1  
KVQKLSKTVIEVASLQNELATHL  
HVQTQINVLLD  
>RAB 2  
KAEALRRTAVDIASLQSELATHL  
QVQTQNNIN  
>RAB 3  
EVHQIEGRVVEISRLQEIFTKVL  
QQAEIIDSIHQ
```

Seq	Feat1	Feat2	Feat3
RAB 1	0.87	0.65	0.23
RAB 2	0.03	0.12	0.54
RAB 3	0.34	0.65	0.49

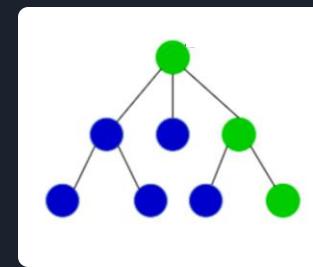
```
from sklearn.preprocessing import StandardScaler  
scaler=StandardScaler()  
X_train=scaler.fit_transform(X_train)  
X_test=scaler.transform(X_test)  
  
from sklearn.neighbors import KNeighborsClassifier  
k = 15  
knn = KNeighborsClassifier(n_neighbors = k, metric = "minkowski", p=2)  
knn.fit(X_train, y_train)  
y_pred = knn.predict(X_test)  
  
from sklearn.metrics import classification_report, accuracy_score  
print(classification_report(y_test, y_pred))  
print("accuracy : ", accuracy_score(y_test, y_pred))  
  
from sklearn.model_selection import train_test_split  
from sklearn.ensemble import RandomForestClassifier  
from sklearn.metrics import confusion_matrix, accuracy_score  
  
rf = RandomForestClassifier(n_estimators=100, random_state=42)  
rf.fit(X_train, y_train)  
y_pred = rf.predict(X_test)  
  
# calculer et afficher la matrice de confusion et la précision  
conf_matrix = confusion_matrix(y_test, y_pred)  
accuracy = accuracy_score(y_test, y_pred)  
  
print("Matrice de confusion :\n", conf_matrix)  
print("Précision : ", accuracy)
```



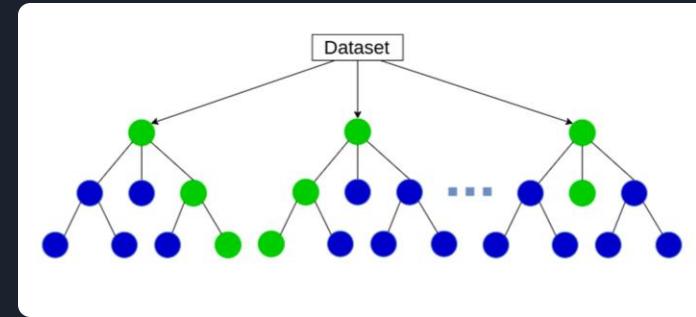
# Methods used



KNN



Decision tree



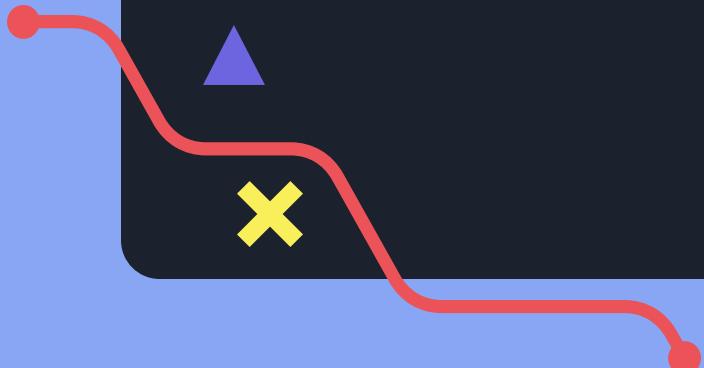
Random forest



# First feature: GAAC

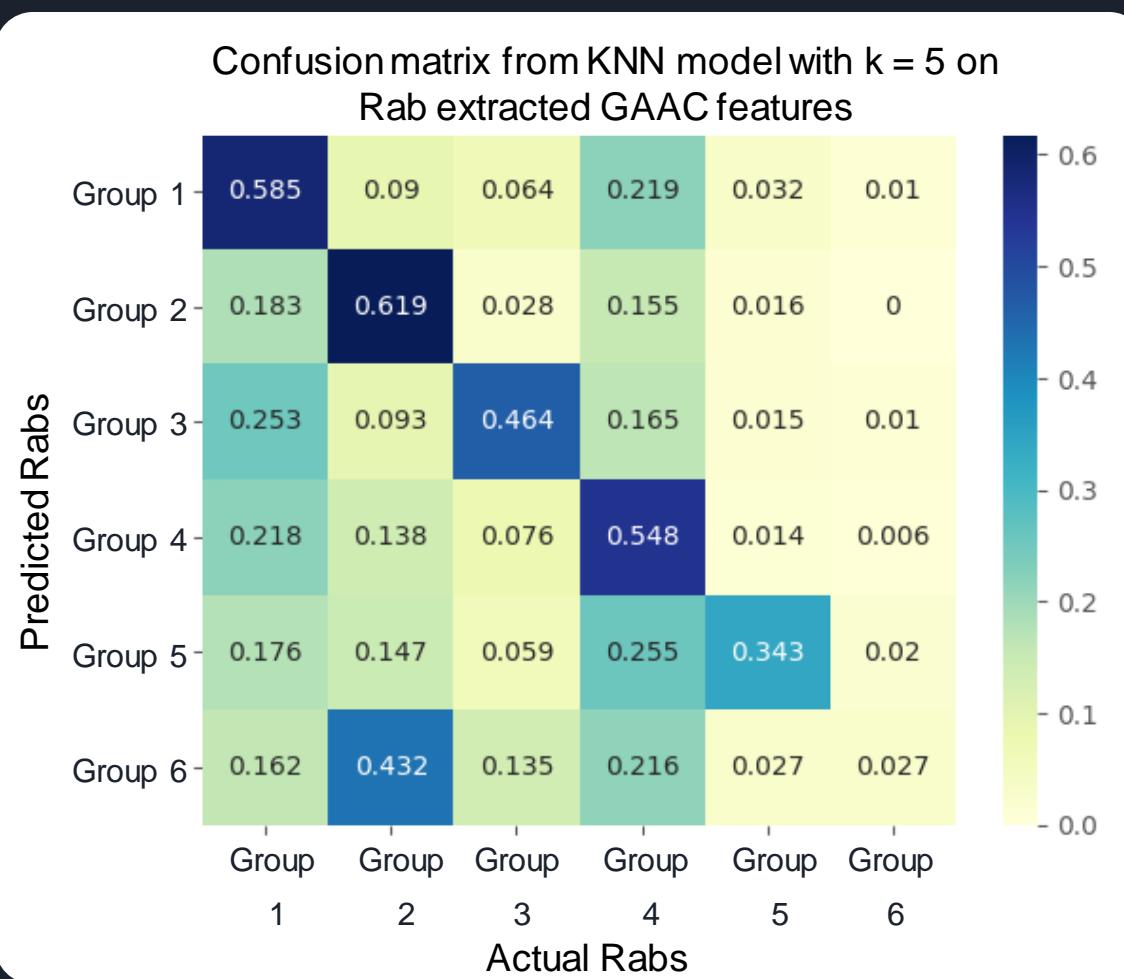
- GAAC = grouped amino acid composition
- **Physico-chemical** properties

	Aliphatic	Aromatic	Positively charged	Negatively charged	Uncharged
rab1	0,2	0,3	0,1	0,3	0,1
rab2	0,3	0,1	0,3	0,1	0,2
rab3	0,1	0,2	0,3	0,3	0,1





# GAAC predictor using KNN



**Accuracy: 53%**

Precision: 0.523

Recall: 0.526

F1-Score: 0.519

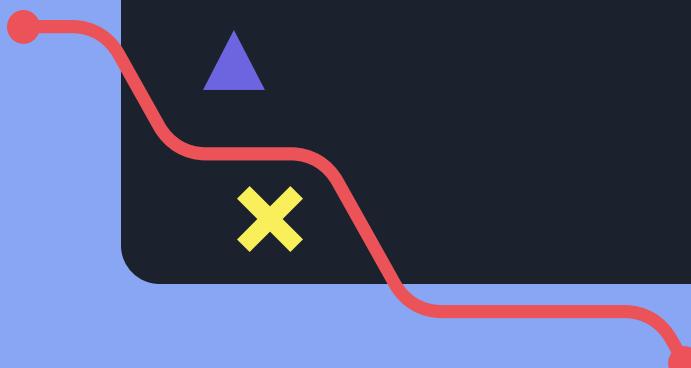


## Second feature: CKSAAP

- CKSAAP = Composition of **k-spaced** amino acid pairs

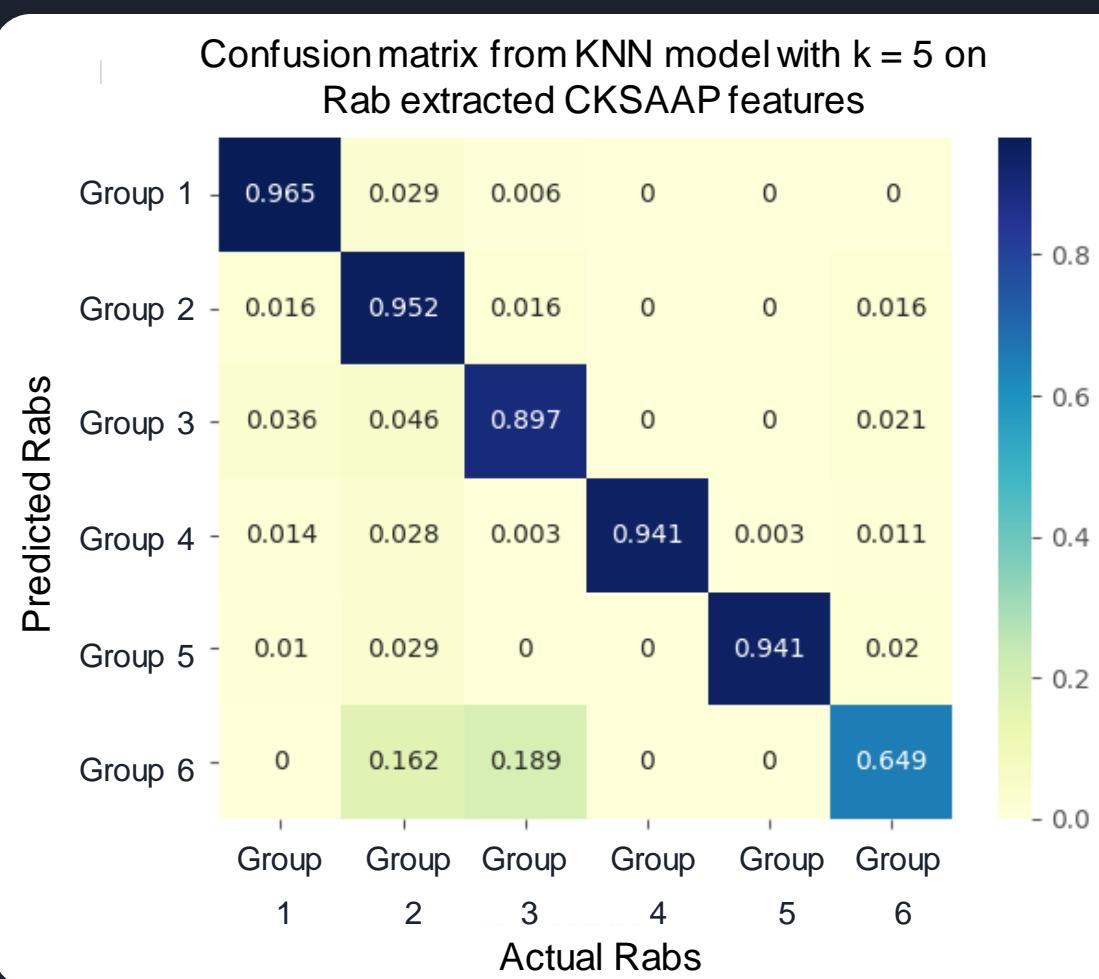
k	AA	LT
k=1	AxA	LxT
k=3	AxxxA	LxxxT
k=5	AxxxxxA	LxxxxxT

Protein sequence : DSAWA**E**LSGCIKT





# CKSAAP predictor using KNN

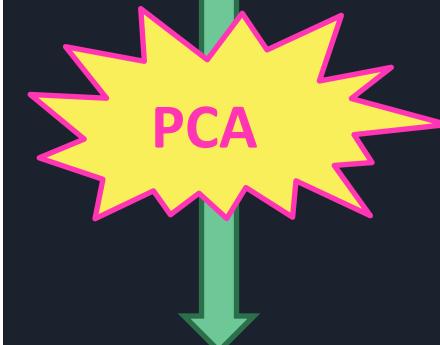


**Accuracy: 94.7%**

Precision: 0.950

Recall: 0.947

F1-Score: 0.945



**Accuracy: 93.4%**

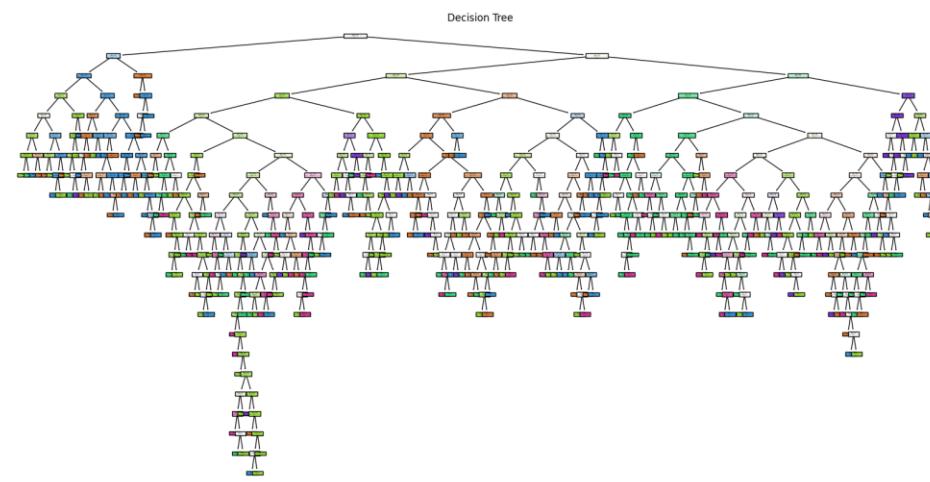
Precision: 0.936

Recall: 0.934

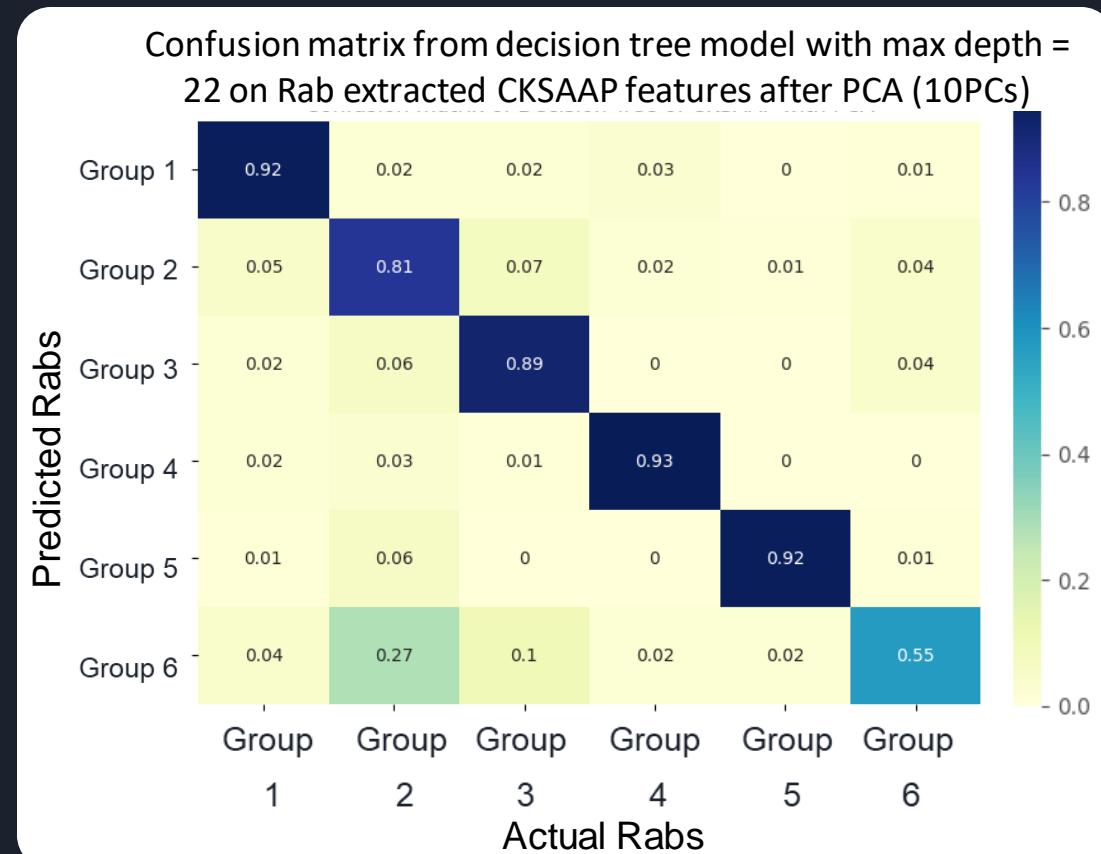
F1-Score: 0.934



# CKSAAP predictor using decision tree

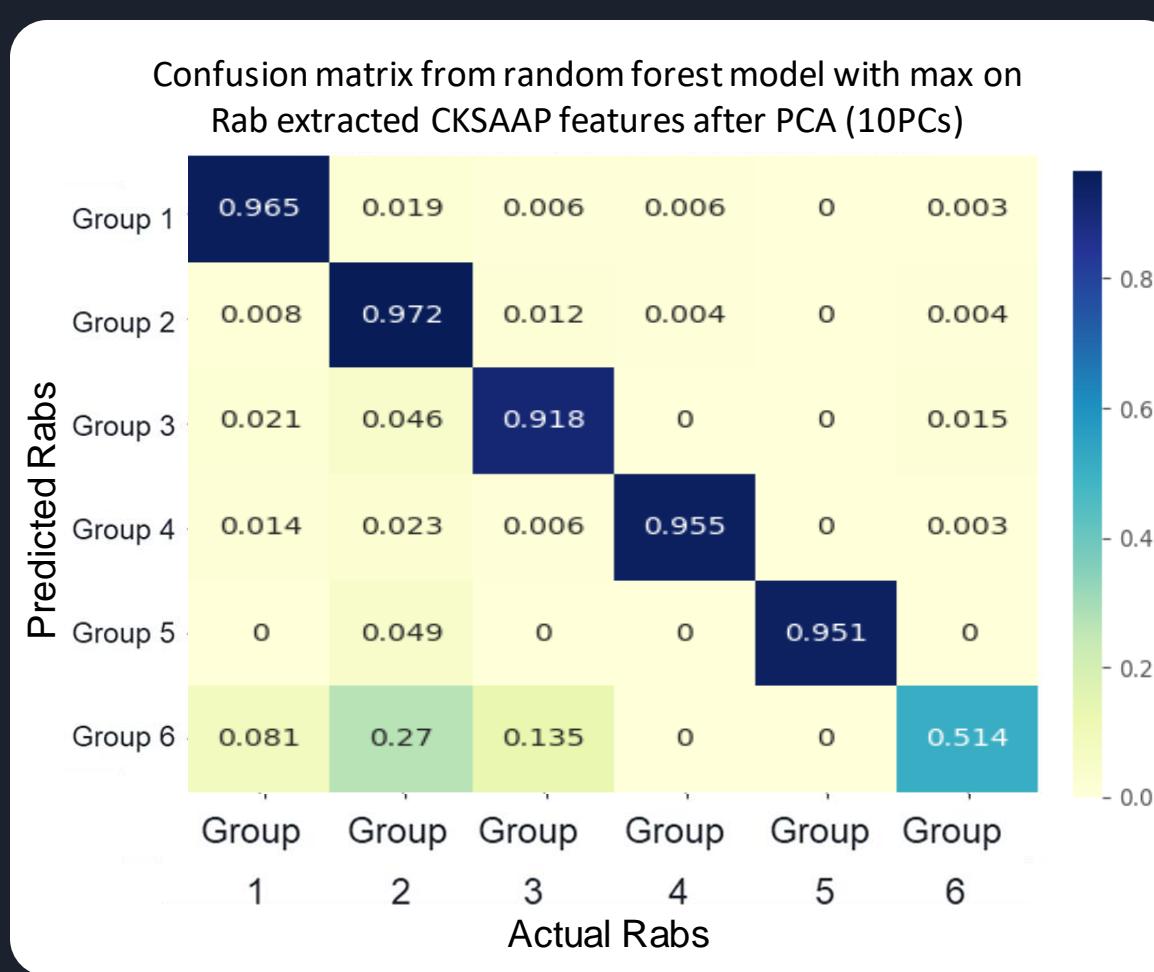


Accuracy: 89,28%  
Precision: 0.839  
Recall: 0.835  
F-score: 0.837



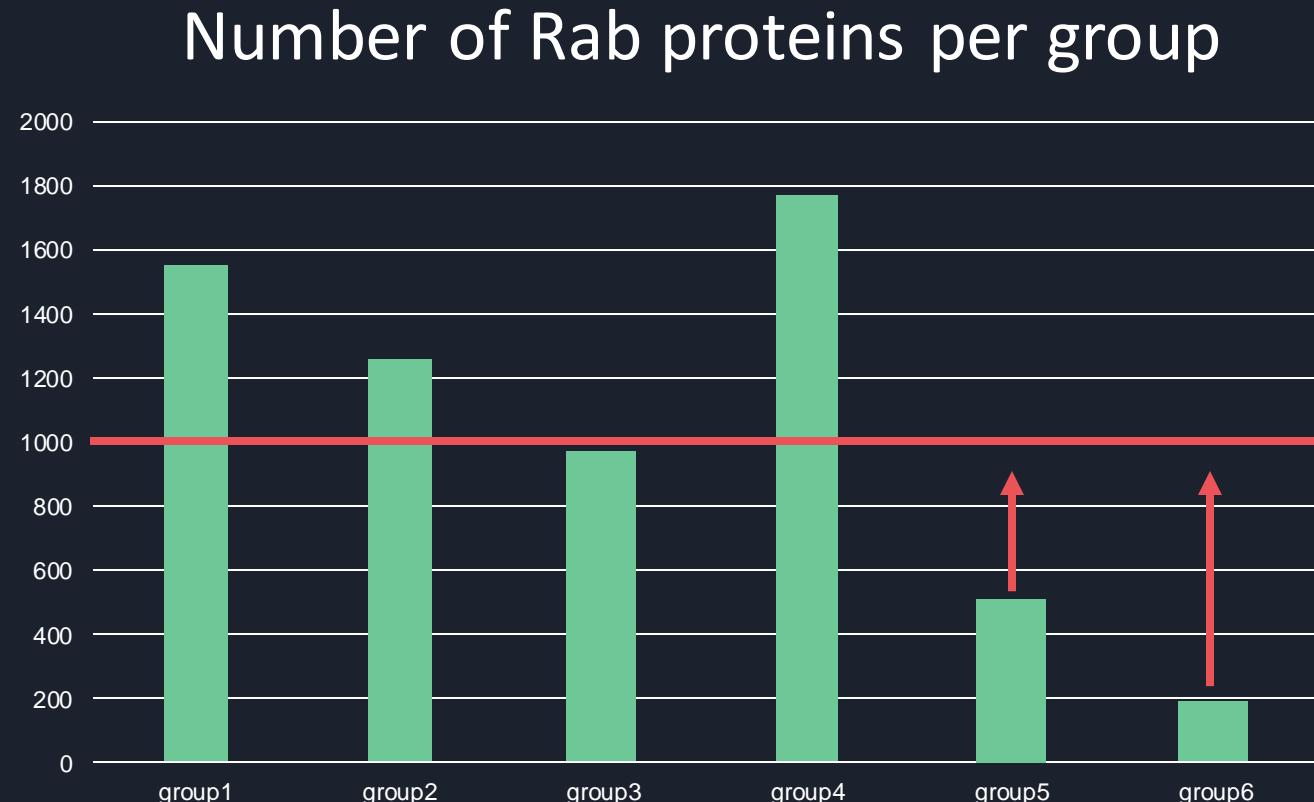


# CKSAAP predictor using random forest





# What is going on with group 6?

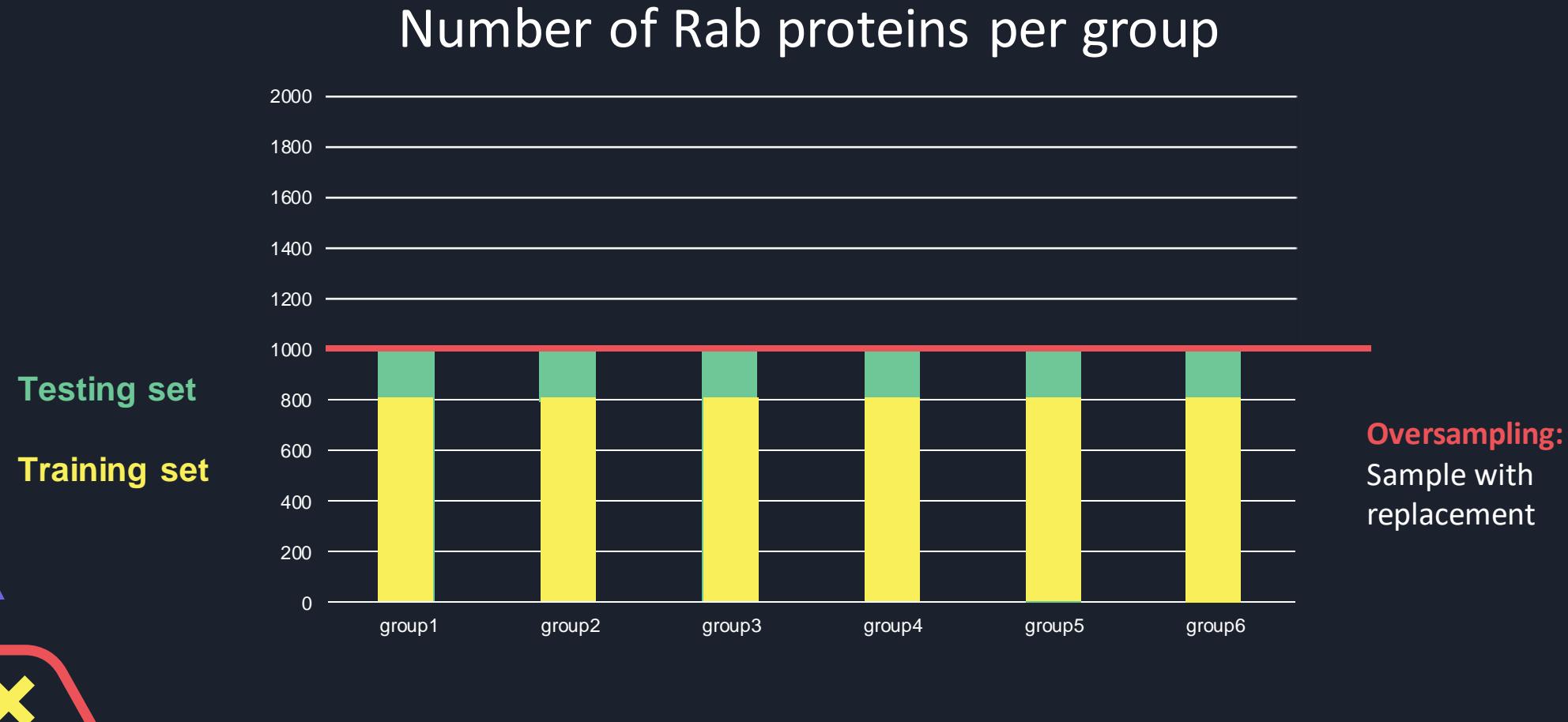


**Oversampling:**  
Sample with  
replacement





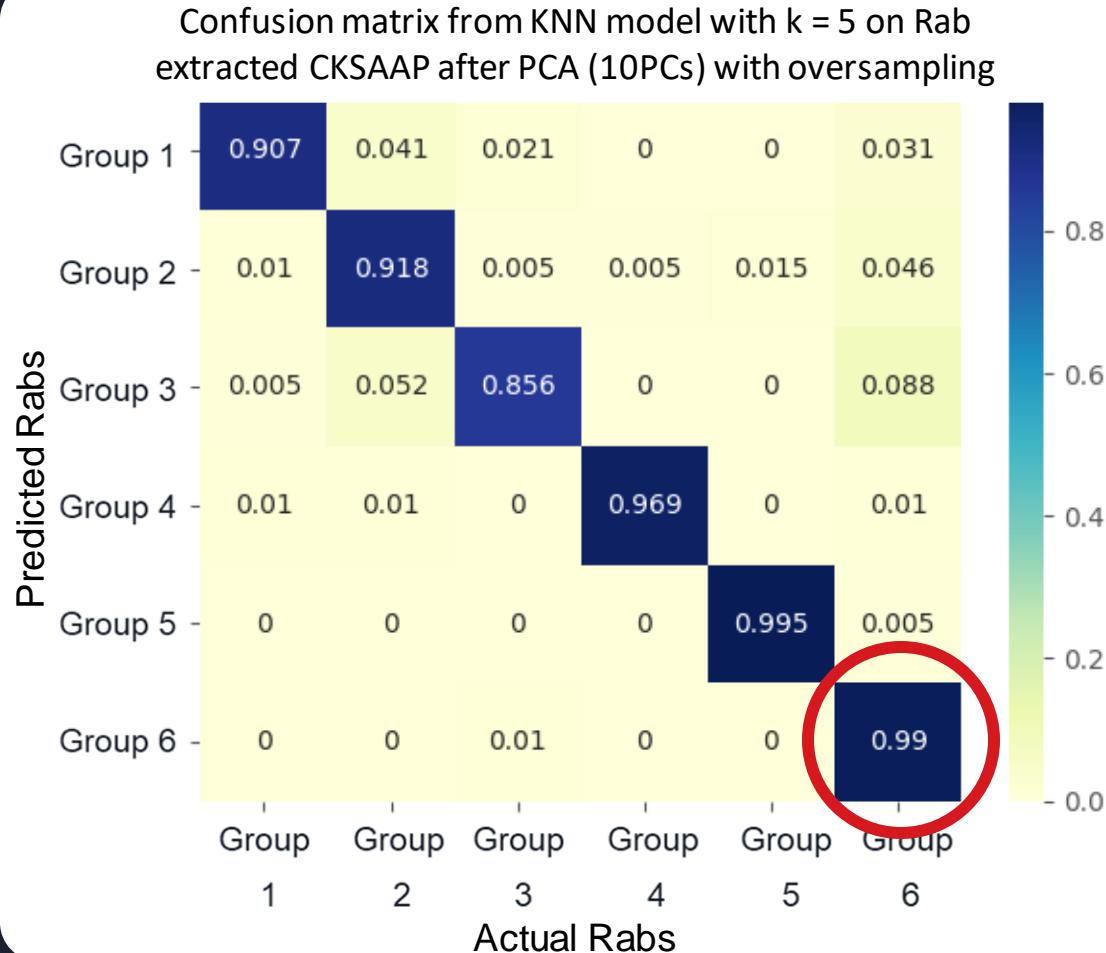
# What is going on with group 6?





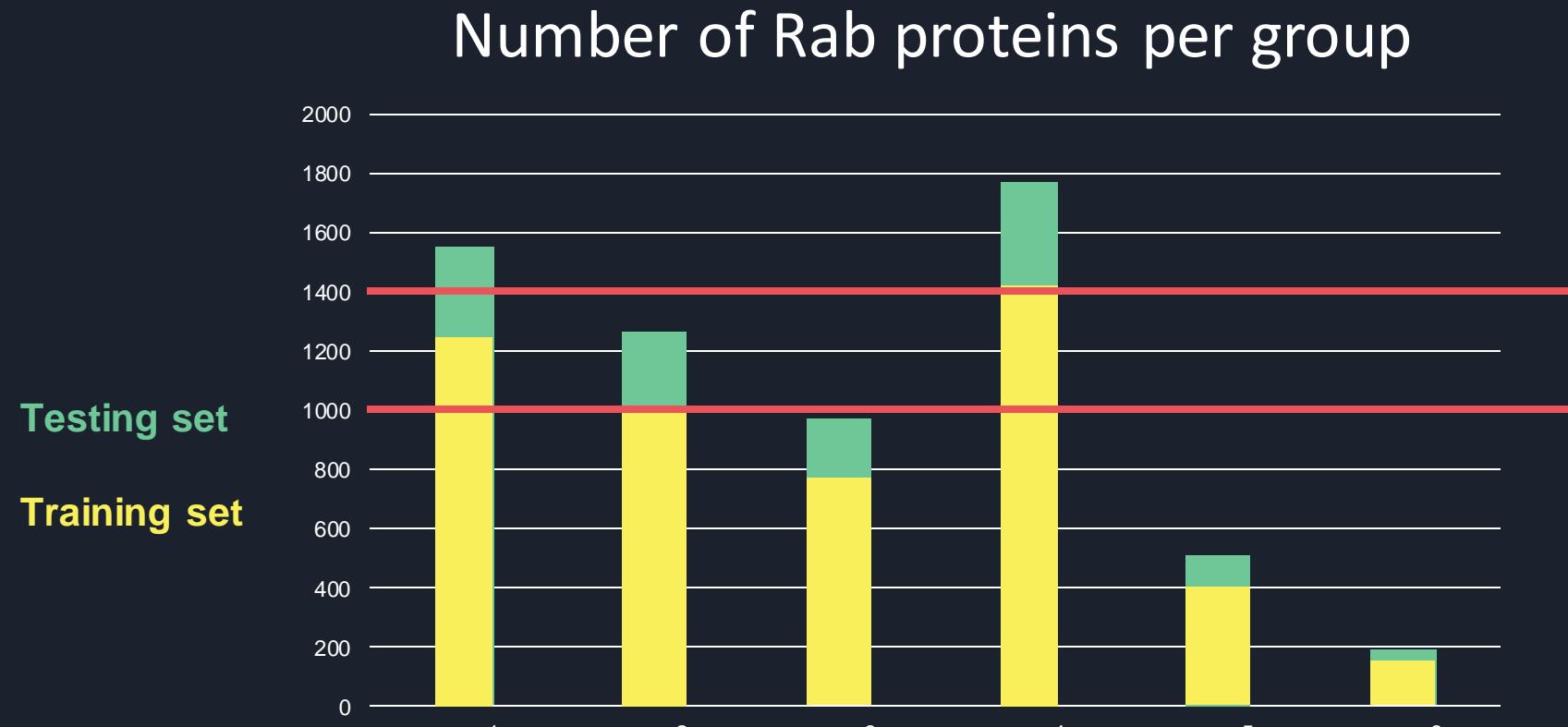
*overfitting*

# CKSAAP with ~~resampling~~





# Improving sampling

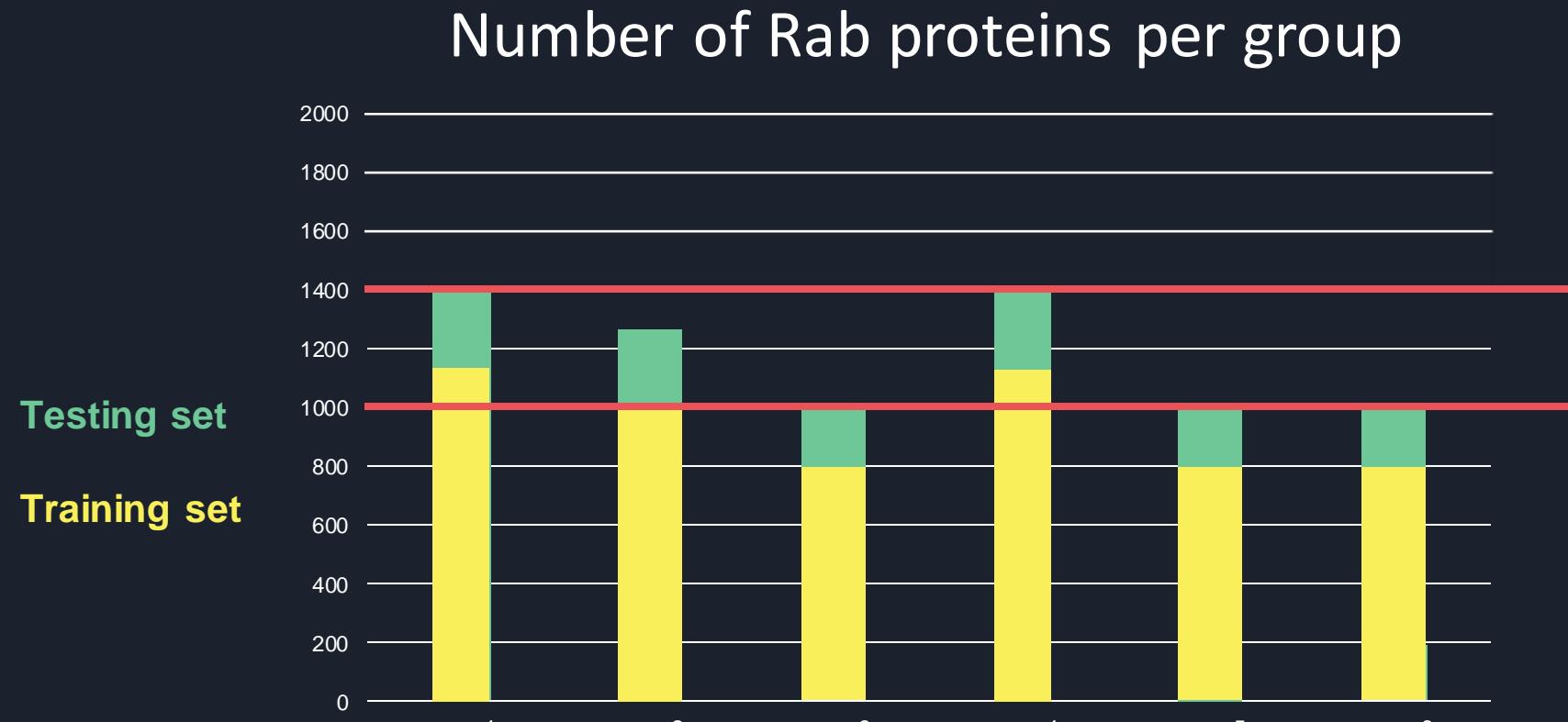


Testing set

Training set



# Improving sampling

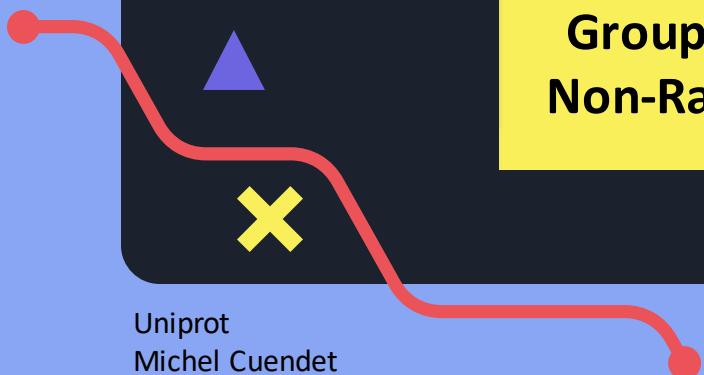
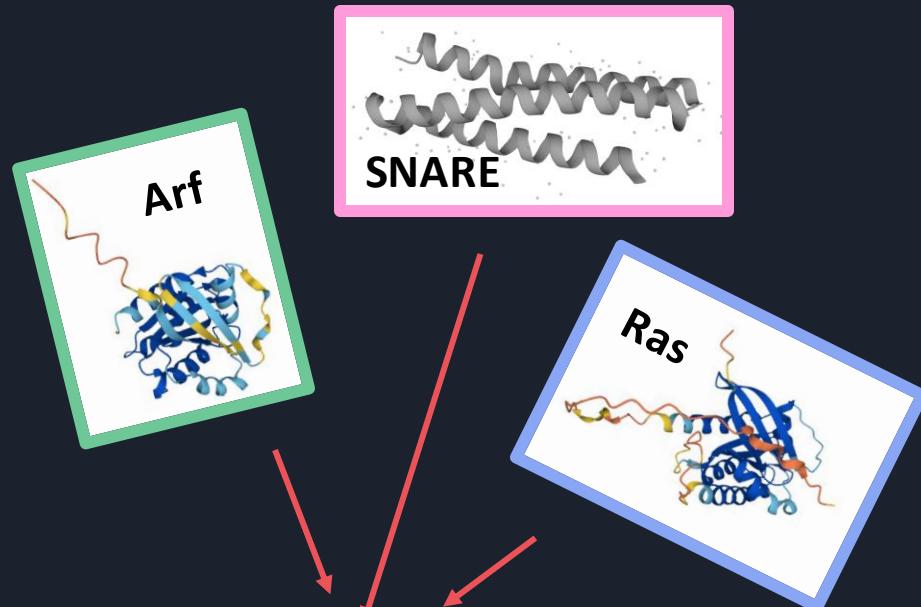


Testing set

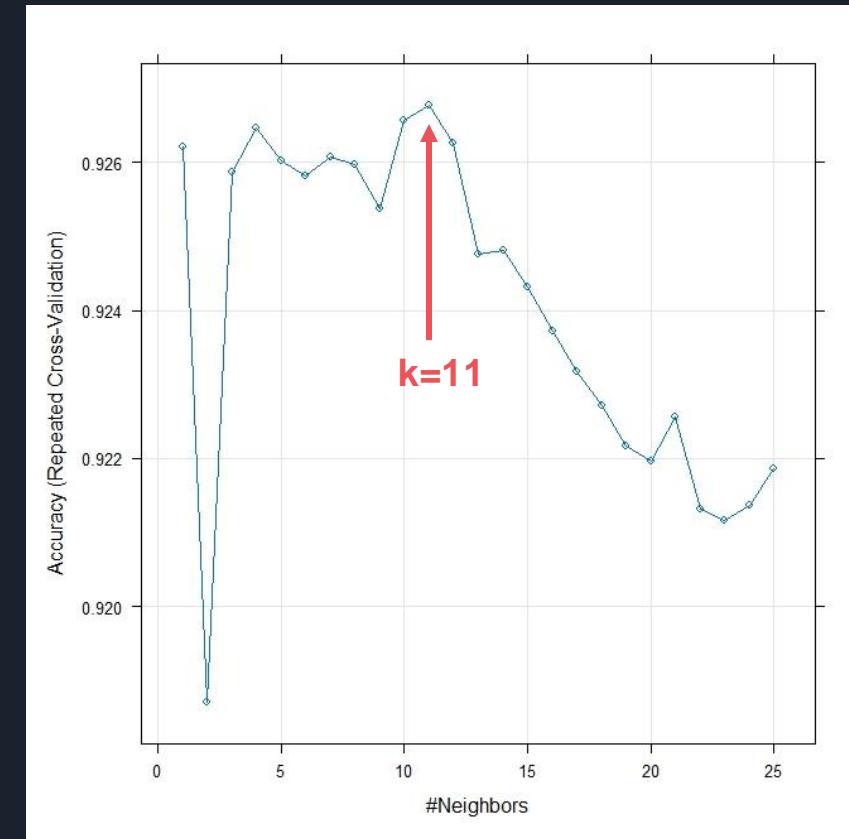
Training set



# Improving the model



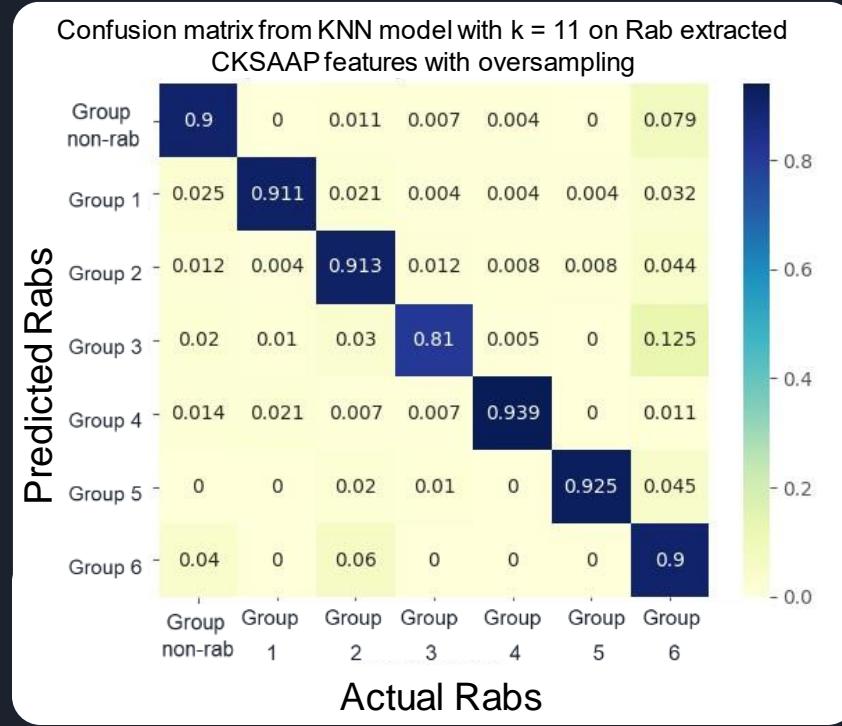
Uniprot  
Michel Cuendet



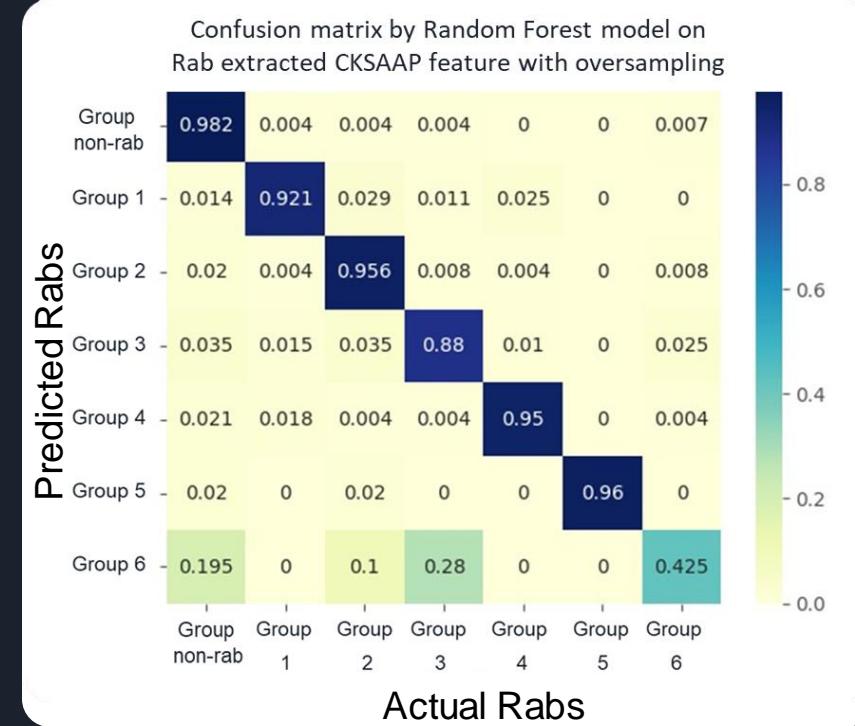


# CKSAAP predictor with improved model

## KNN



## Random forest





## Third feature: CTD

- **Physico-chemical** properties (CTD= Conjoint Triad Descriptors)

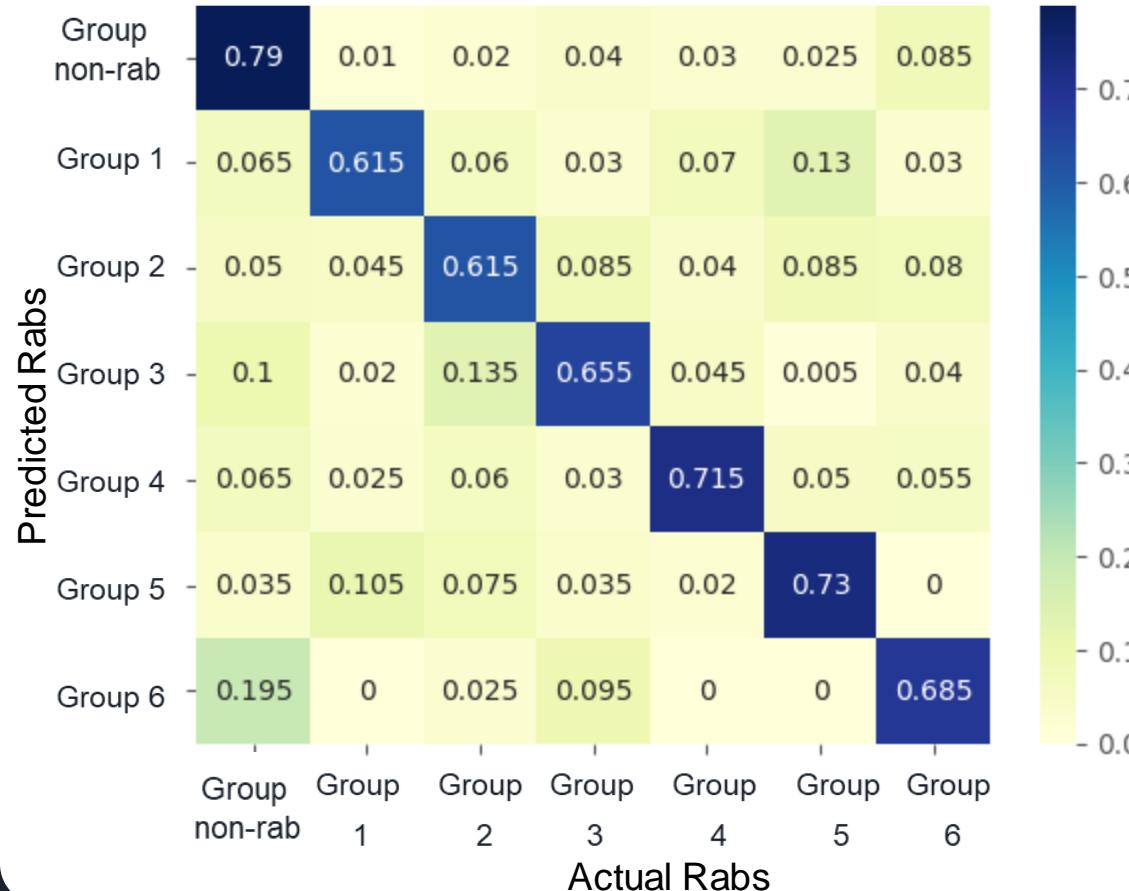
	Hydro-phobicity	Polarity	Van der Waals volume	Charge	...
rab1	0,212	0,66454	0,117	0,5888	
rab2	0,42588	0,1588	0,3665	0,05	
rab3	0,61	0,258	0,557	0,255	





# CTD with KNN

Confusion matrix from KNN model with  $k = 11$  on Rab extracted CTD after PCA (10PCs) feature with oversampling



Accuracy: 69%

Precision: 0.692

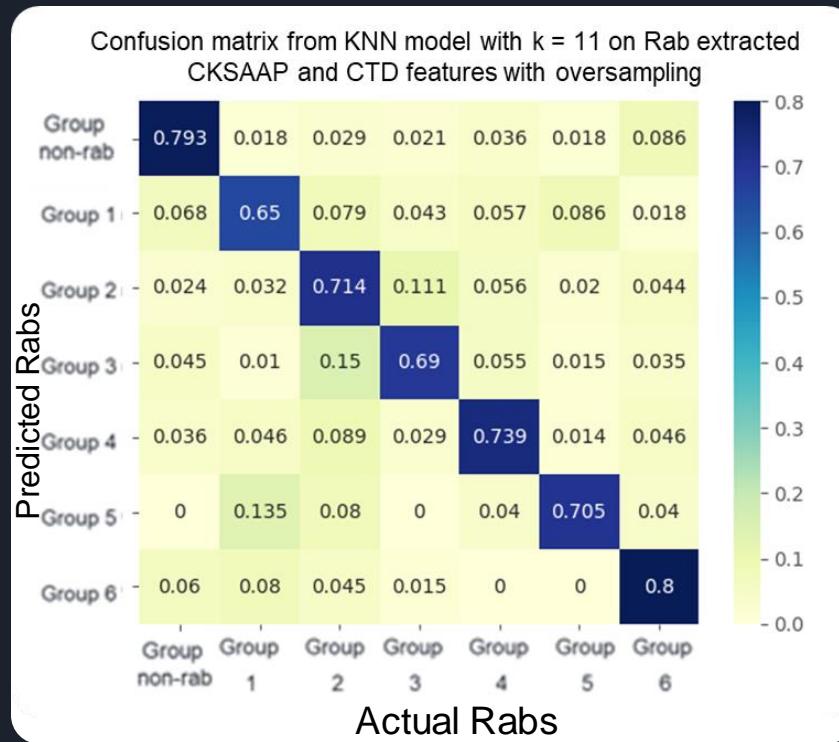
Recall: 0.686

F1-Score: 0.686



# CKSAAP and CTD predictors combined

## KNN



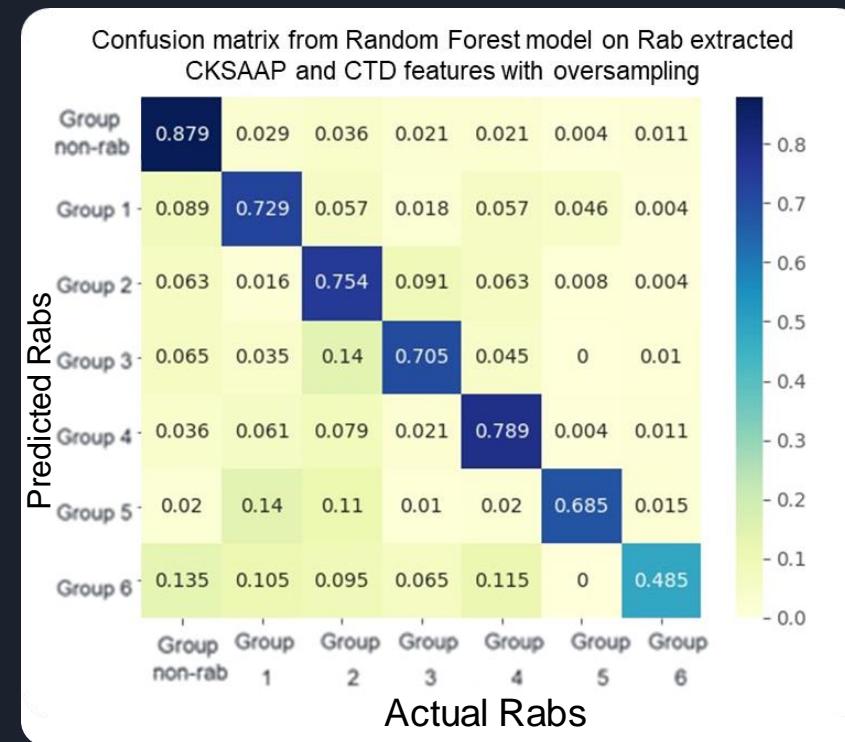
Accuracy: 73%

Precision: 0.731

Recall: 0.727

F1-Score: 0.727

## Random forest



Accuracy: 73%

Precision: 0.747

Recall: 0.730

F1-Score: 0.728



## Fourth feature: KSCT

- **Physico-chemical** properties (KSCT= K spaced conjoint triad)

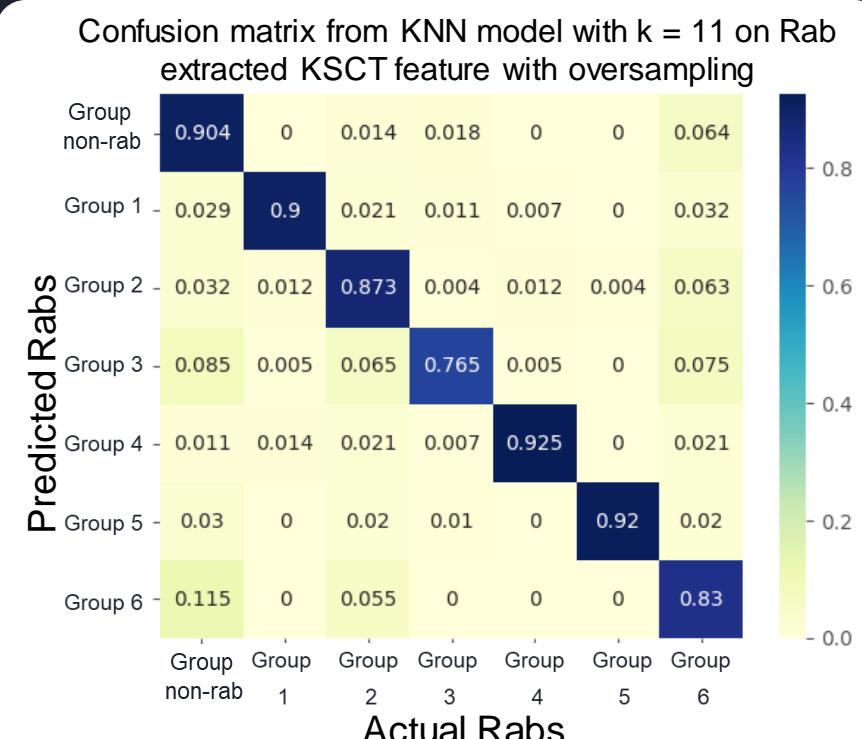
	Hydro-phobicity	Polarity	Van der Waals volume	Charge	...
rab1	0,212	0,66454	0,11	0,5888	
rab2	0,02588	0,0588	0,3665	0,005	
rab3	0,6	0,258	0,557	0,255	





# KSCT predictor

## KNN



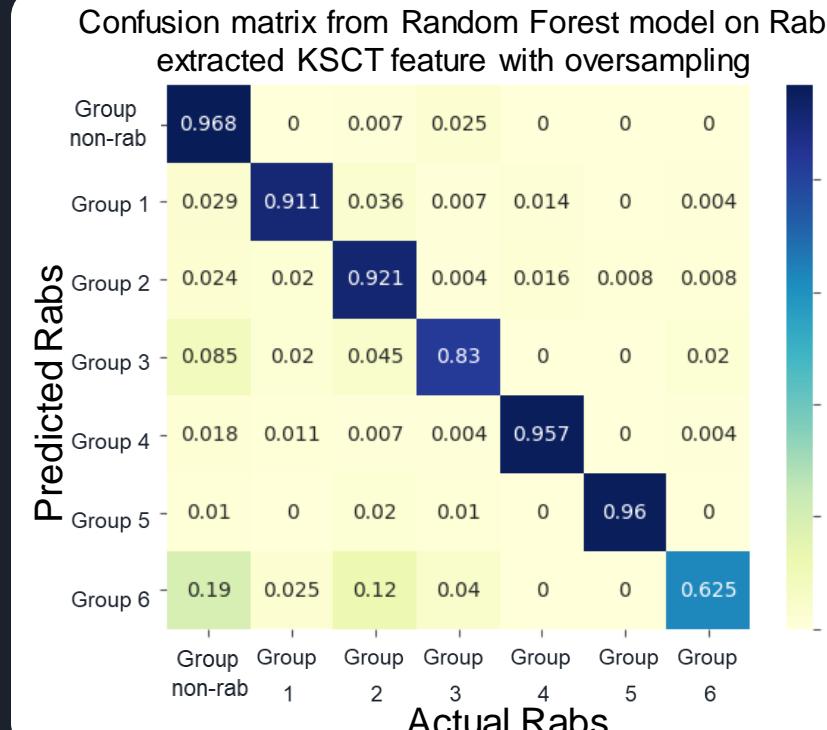
Accuracy: 87.9%

Precision: 0.888

Recall: 0.879

F1-Score: 0.881

## Random forest



Accuracy: 89.2%

Precision: 0.900

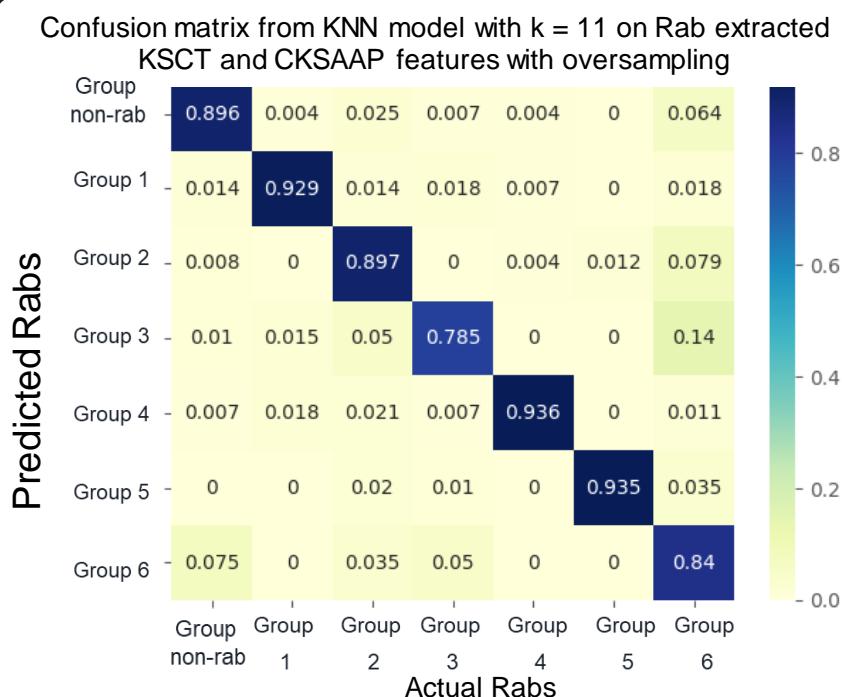
Recall: 0.892

F1-Score: 0.890



# KSCT and CKSAAP predictors combined

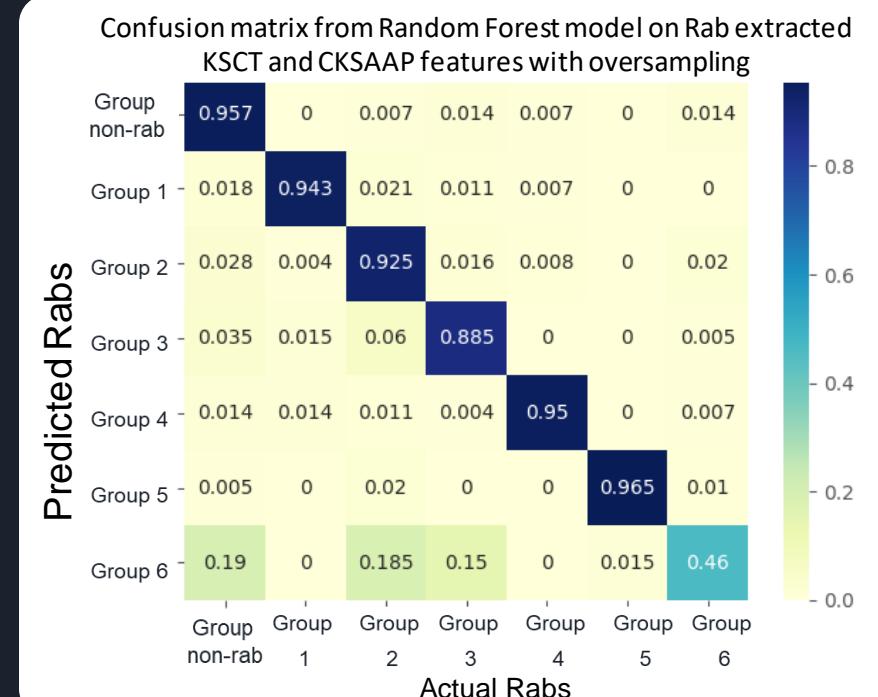
## KNN



**Accuracy: 89.3%**  
Precision: 0.901

Recall: 0.893  
F1-Score: 0.895

## Random forest



**Accuracy: 88.2%**  
Precision: 0.888

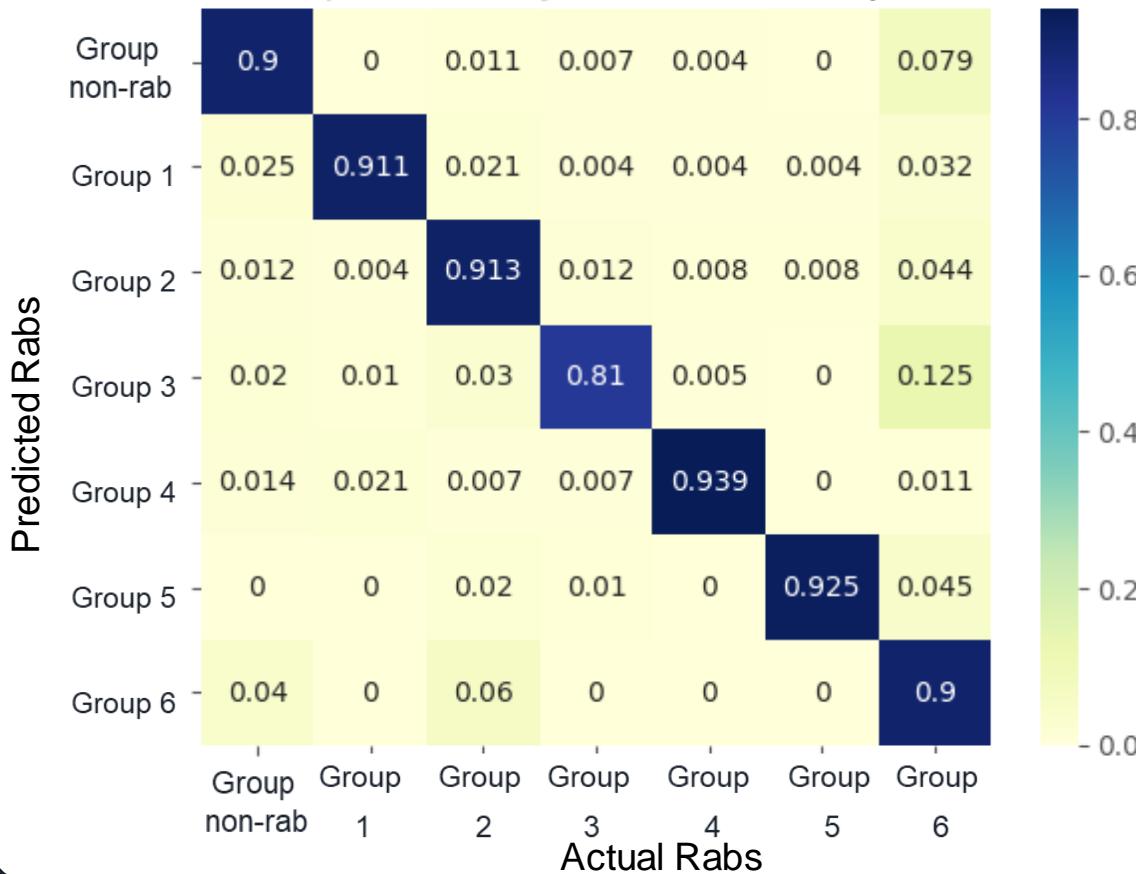
Recall: 0.882  
F1-Score: 0.876



# Best of the best: CKSAAP with oversampling

Confusion matrix from KNN model with  $k = 11$  on Rab extracted

CKSAAP features with oversampling



Accuracy: 90.2%

Precision: 0.912

Recall: 0.902

F1-Score: 0.905



# Take home messages

Best model: KNN with CKSAAP feature



Combos of features do not add to the model





# Challenges



- **Unequal distribution** in classes
- **Unintuitive results**: best model is not a combination (?)
- **Lots of testing** and probing, limited time
- **Google colab**... not so easy to colab
- **Different kind of coding** but lots of help from our mentor and forums/Chat GPT





# Feedback

- Highly concentrated working period blur lines between work and personal time
- Grade disproportionately based on presentation, rather than work done
- Little time between intermediate and final presentation but helpful and inspiring
- Very helpful mentor
- Apply theory seen in class on an interesting project
- Learned a lot about coding and problem solving



# Thanks!

Questions?

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