

02_KNN

December 2, 2021

```
[1]: # ZOOM de la SFGP - GTIAP
# 02 / 12 / 2021
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#
#
# Importation des bibliothèques de bases - Pandas et Numpy - pour manipuler les
↳ data
#
import pandas as pd
import numpy as np
#
# Graphes
#
import matplotlib
import matplotlib.pyplot as plt
from matplotlib.pyplot import plot
%matplotlib inline
#
# Packages des algorithmes de ML
#
import sklearn
```

```
[2]: #
# Versions utilisées
#
print('Version de matplotlib = ',matplotlib.__version__)
print('Version de sklearn = ',sklearn.__version__)
!python -V
```

Version de matplotlib = 3.3.4

Version de sklearn = 0.24.1

Python 3.8.8

```
[3]: #
# Importation du dataset
#
```

```
df = pd.read_csv('Dataset_Complet.csv', sep = ',')
```

```
[4]: #  
# Affectation des features et des targets  
#  
X = df[['Pressure_(bar)', 'Resident_Time_(s)', 'Temperature_(C)',  
↪ 'Time_(sec)']]  
y = df.drop(['Pressure_(bar)', 'Resident_Time_(s)', 'Temperature_(C)',  
↪ 'Time_(sec)'], axis=1)
```

```
[5]: #  
# Répartition des valeurs "train" et "test"  
#  
from sklearn.model_selection import train_test_split # méthode pour le train /  
↪ test  
#  
X_train , X_test , y_train , y_test = train_test_split( X , y , test_size=0.2 ,  
↪ shuffle=True , random_state=4 )
```

```
[6]: #  
# Méthode de ML : KNN  
#  
from sklearn.neighbors import KNeighborsRegressor # methode KNN regression  
#  
# Metrics -  
#  
from sklearn.metrics import mean_squared_error # erreur MSE  
from sklearn.metrics import mean_absolute_error # MAE
```

```
[7]: #  
# Affectation de l'estimateur  
#  
Modele_KNN = KNeighborsRegressor()
```

```
[8]: #  
# Entraînement de la méthode avec les paramètres par défaut  
#  
Modele_KNN.fit( X_train , y_train )
```

```
[8]: KNeighborsRegressor()
```

```
[9]: #  
# Affichage des scores  
#  
print('R2_score')  
print('KNN R2 train = ' , Modele_KNN.score(X_train,y_train))  
print('KNN R2 test = ' , Modele_KNN.score(X_test,y_test))
```

```
R2_score
KNN R2 train = 0.9995255798604405
KNN R2 test = 0.9989076549162693
```

```
[10]: #
# Affichage des scores
#
print('MAE')
print('KNN MAE train = ', mean_absolute_error(y_train,Modele_KNN.
↪predict(X_train)))
print('KNN MAE test = ', mean_absolute_error(y_test,Modele_KNN.
↪predict(X_test)))
#
# Affichage des scores
#
print('MSE')
print('KNN MSE train = ', mean_squared_error(y_train,Modele_KNN.
↪predict(X_train)))
print('KNN MSE train = ', mean_squared_error(y_train,Modele_KNN.
↪predict(X_train)))
```

```
MAE
KNN MAE train = 8.071257215842221e-05
KNN MAE test = 0.00013580976254289897
MSE
KNN MSE train = 4.95376100840807e-07
KNN MSE train = 4.95376100840807e-07
```

```
[11]: #
# Affichage des hyper-paramètres disponible pour la méthode
#
Modele_KNN.get_params()
```

```
[11]: {'algorithm': 'auto',
'leaf_size': 30,
'metric': 'minkowski',
'metric_params': None,
'n_jobs': None,
'n_neighbors': 5,
'p': 2,
'weights': 'uniform'}
```

```
[12]: #
# Analyse paramétrique sur un paramètre
#
n_neighbors = [x for x in range(1,15)]
#
for i in (n_neighbors):
```

```

Modele_KNN = KNeighborsRegressor(n_neighbors=i)
Modele_KNN.fit( X_train , y_train )
print('n_neighbors = ',i,'- KNN R2 train = ' , Modele_KNN.
↪score(X_train,y_train))

```

```

n_neighbors = 1 - KNN R2 train = 0.999999990297944
n_neighbors = 2 - KNN R2 train = 0.9998172707584947
n_neighbors = 3 - KNN R2 train = 0.9998356948649295
n_neighbors = 4 - KNN R2 train = 0.9997015380413568
n_neighbors = 5 - KNN R2 train = 0.9995255798604405
n_neighbors = 6 - KNN R2 train = 0.9992164852623338
n_neighbors = 7 - KNN R2 train = 0.9988390997460975
n_neighbors = 8 - KNN R2 train = 0.997979909120591
n_neighbors = 9 - KNN R2 train = 0.9967337699160571
n_neighbors = 10 - KNN R2 train = 0.9947424937770087
n_neighbors = 11 - KNN R2 train = 0.9920331430350725
n_neighbors = 12 - KNN R2 train = 0.9887870666377916
n_neighbors = 13 - KNN R2 train = 0.9854397935890996
n_neighbors = 14 - KNN R2 train = 0.9826805579861378

```

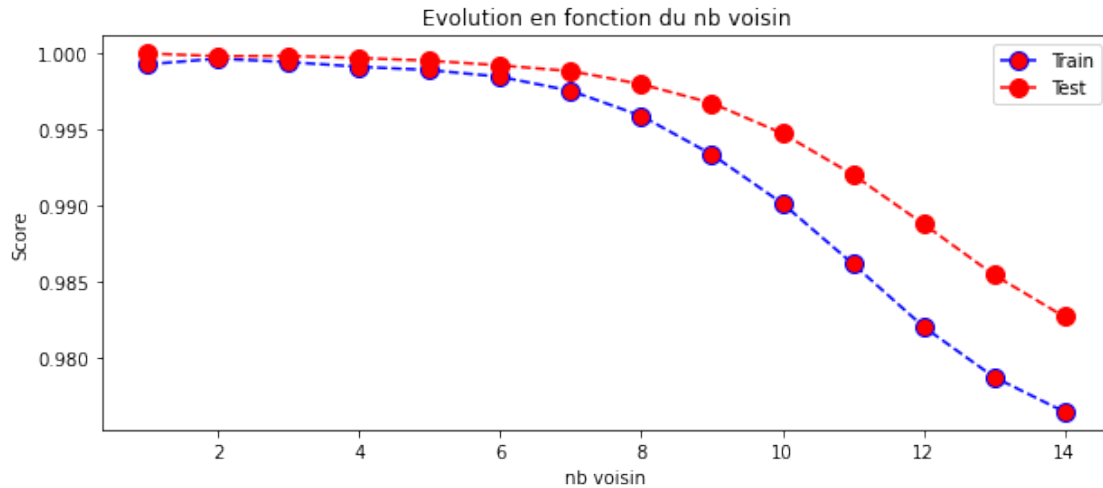
```

[13]: #
# Analyse paramétrique sur un paramètre - Résultat sous forme graphique
#
score_rate = []
score_train_rate = []
for i in range(1,15):
    Modele_KNN = KNeighborsRegressor(n_neighbors=i)
    Modele_KNN.fit( X_train , y_train )
    score_train_rate.append(Modele_KNN.score(X_train,y_train))
    score_rate.append(Modele_KNN.score(X_test,y_test))

plt.figure(figsize=(10,4))
plt.
↪plot(range(1,15),score_rate,color='blue',linestyle='dashed',marker='o',markerfacecolor='red)
plt.
↪plot(range(1,15),score_train_rate,color='red',linestyle='dashed',marker='o',markerfacecolor)
plt.title('Evolution en fonction du nb voisin')
plt.xlabel('nb voisin')
plt.ylabel('Score')
plt.legend()

```

[13]: <matplotlib.legend.Legend at 0x1ee84b9fbe0>



```
[ ]: #
# Cross-Validation - Influence des hyper-paramètres
#
```

```
[14]: #
# Affichage des hyper-paramètres disponible pour la méthode
#
Modele_KNN.get_params()
```

```
[14]: {'algorithm': 'auto',
'leaf_size': 30,
'metric': 'minkowski',
'metric_params': None,
'n_jobs': None,
'n_neighbors': 14,
'p': 2,
'weights': 'uniform'}
```

```
[15]: #
# Analyse paramétrique sur plusieurs paramètres
#
n_neighbors = [x for x in range(1,15)]
weights = ['uniform', 'distance']
metric = ['minkowski', 'euclidean', 'manhattan']
#
# Dictionnaire
#
param_grid = {'n_neighbors': n_neighbors, 'weights': weights, 'metric':
↳metric}
```

```
[16]: #
# Cross-Validation - GridSearchCV
#
from sklearn.model_selection import GridSearchCV
#
Modele_KNN = KNeighborsRegressor()
#
grid = GridSearchCV( estimator=Modele_KNN , param_grid = param_grid , cv=3 ,
↳ verbose=2 )
#
grid.fit(X_train,y_train)
```

Fitting 3 folds for each of 84 candidates, totalling 252 fits

```
[CV] END ...metric=minkowski, n_neighbors=1, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=1, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=1, weights=uniform; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=1, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=1, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=1, weights=distance; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=2, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=2, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=2, weights=uniform; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=2, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=2, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=2, weights=distance; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=3, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=3, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=3, weights=uniform; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=3, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=3, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=3, weights=distance; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=4, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=4, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=4, weights=uniform; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=4, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=4, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=4, weights=distance; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=5, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=5, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=5, weights=uniform; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=5, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=5, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=5, weights=distance; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=6, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=6, weights=uniform; total time= 0.0s
[CV] END ...metric=minkowski, n_neighbors=6, weights=uniform; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=6, weights=distance; total time= 0.0s
[CV] END ..metric=minkowski, n_neighbors=6, weights=distance; total time= 0.0s
```



```

[CV] END .metric=manhattan, n_neighbors=10, weights=distance; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=11, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=11, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=11, weights=uniform; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=11, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=11, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=11, weights=distance; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=12, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=12, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=12, weights=uniform; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=12, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=12, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=12, weights=distance; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=13, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=13, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=13, weights=uniform; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=13, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=13, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=13, weights=distance; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=14, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=14, weights=uniform; total time= 0.0s
[CV] END ..metric=manhattan, n_neighbors=14, weights=uniform; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=14, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=14, weights=distance; total time= 0.0s
[CV] END .metric=manhattan, n_neighbors=14, weights=distance; total time= 0.0s

```

```

[16]: GridSearchCV(cv=3, estimator=KNeighborsRegressor(),
                param_grid={'metric': ['minkowski', 'euclidean', 'manhattan'],
                              'n_neighbors': [1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12,
                                              13, 14],
                              'weights': ['uniform', 'distance']}},
                verbose=2)

```

```

[17]: #
      # Affiche les meilleurs hyperparametres trouvés
      #
      grid.best_params_

```

```

[17]: {'metric': 'manhattan', 'n_neighbors': 2, 'weights': 'distance'}

```

```

[18]: #
      # Avec ces hyperparamètres, affiche le meilleur score sur le train
      #
      grid.best_estimator_.score(X_train,y_train)

```

```

[18]: 0.9999999995148973

```

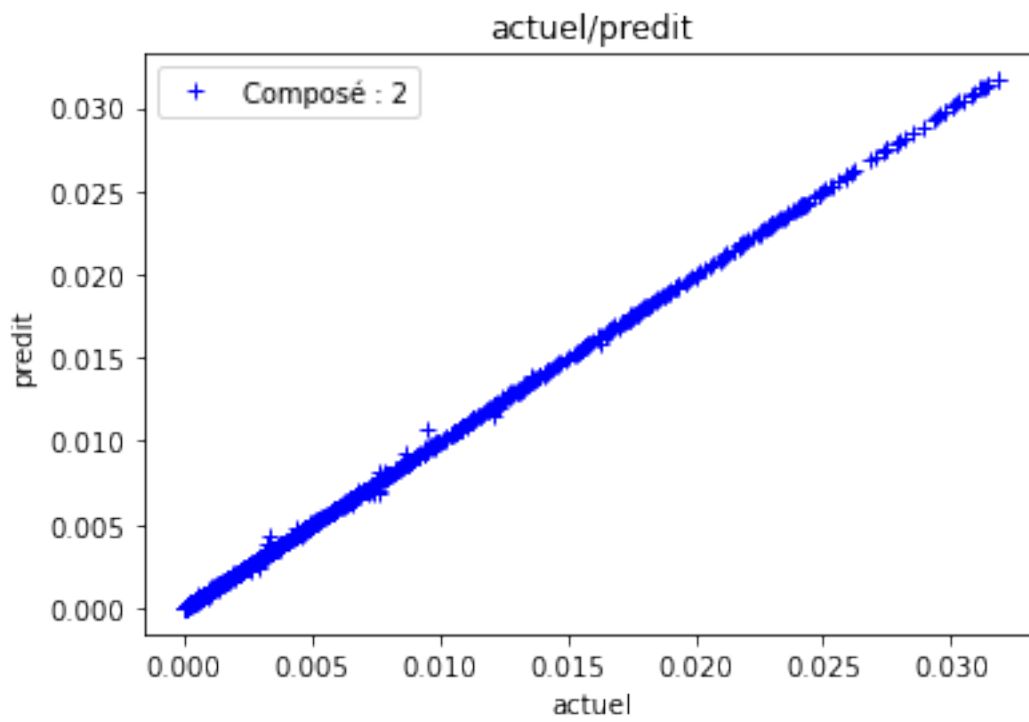
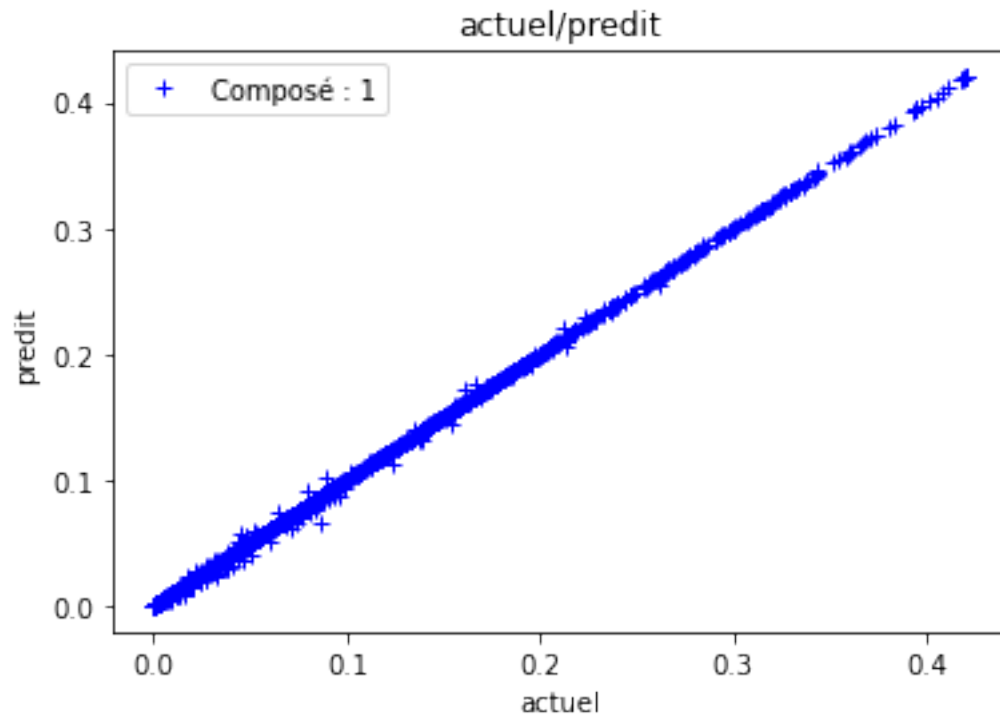
```
[19]: #  
# Avec ces hyperparamètres, affiche le meilleur score sur le test  
#  
grid.best_estimator_.score(X_test,y_test)
```

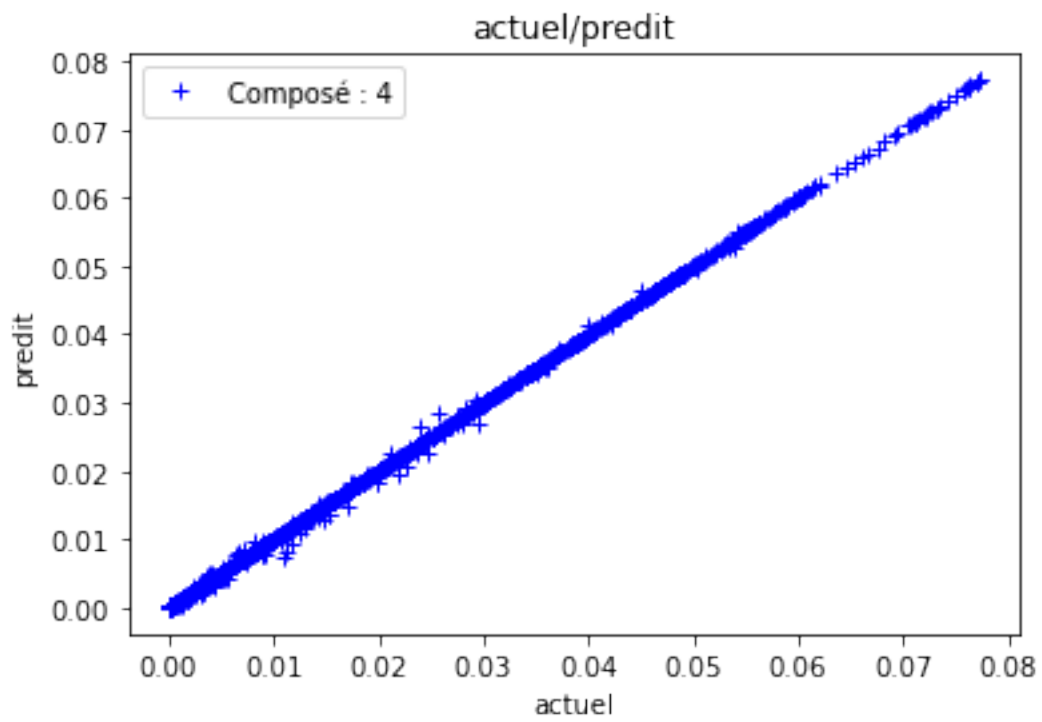
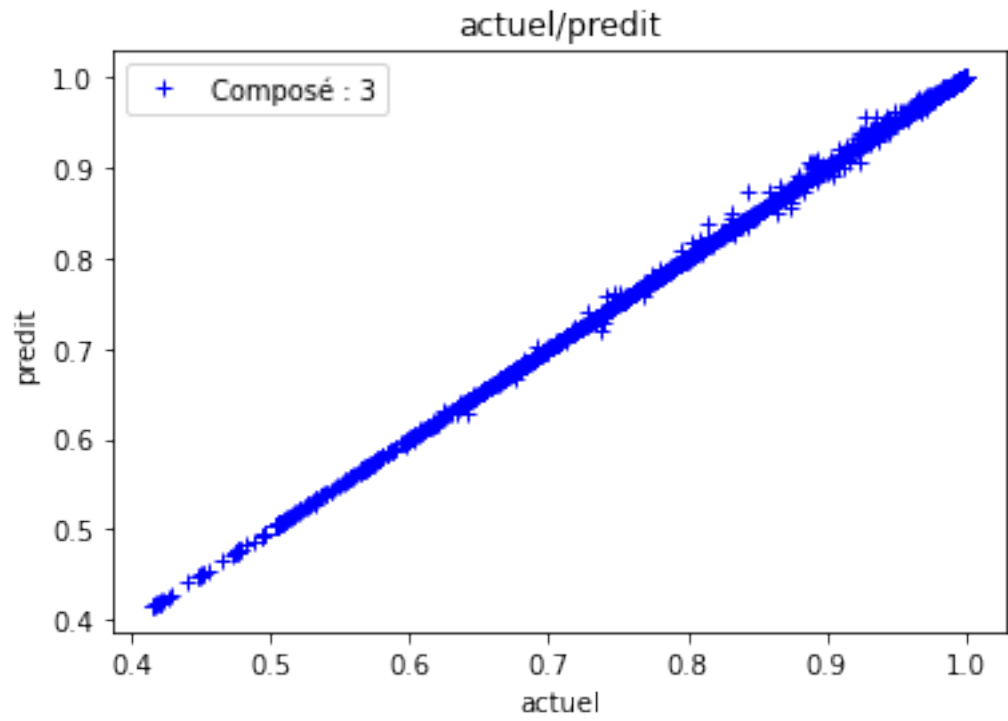
[19]: 0.9997595859006833

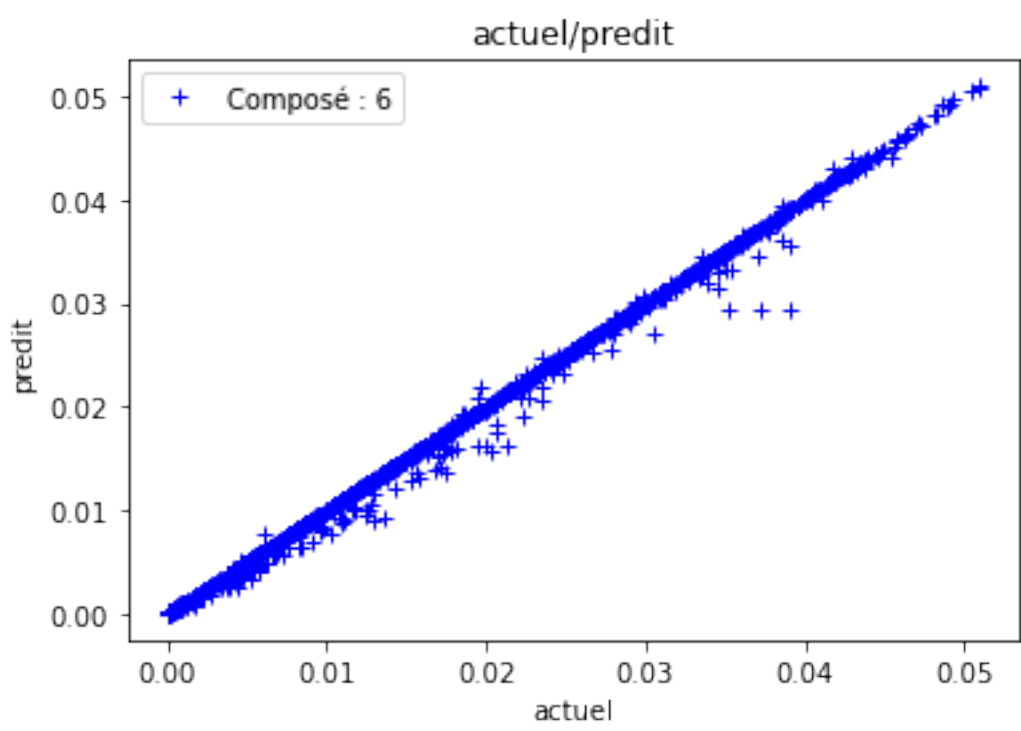
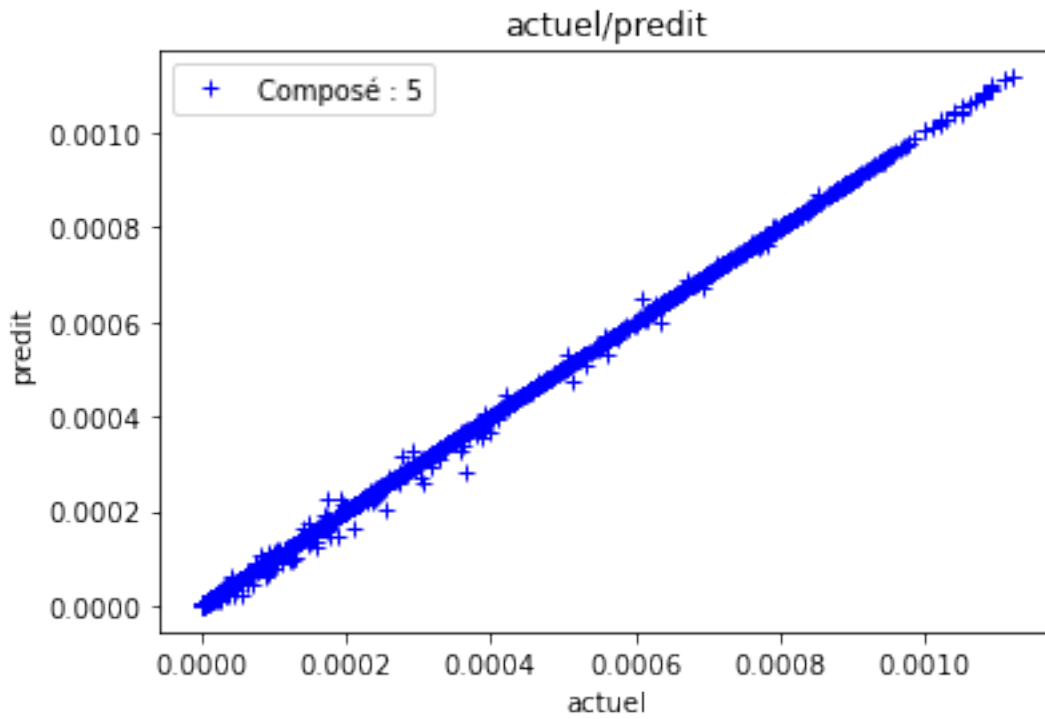
```
[20]: #  
# Génération d'une prédiction  
#  
grid.best_estimator_.predict([[0.003,.2,850,1.5]])
```

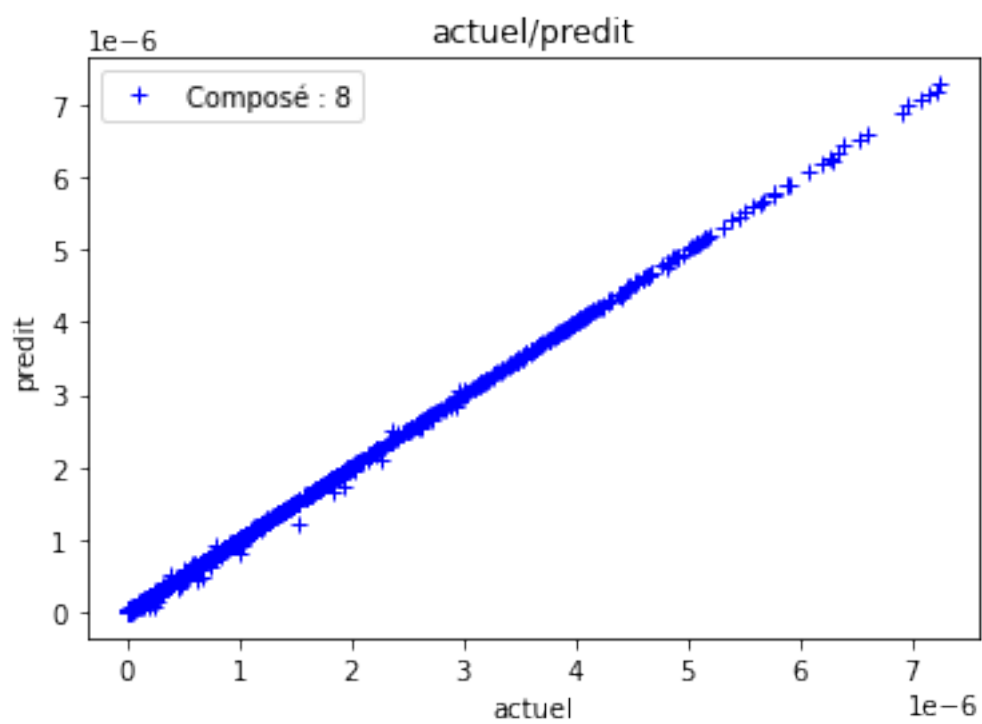
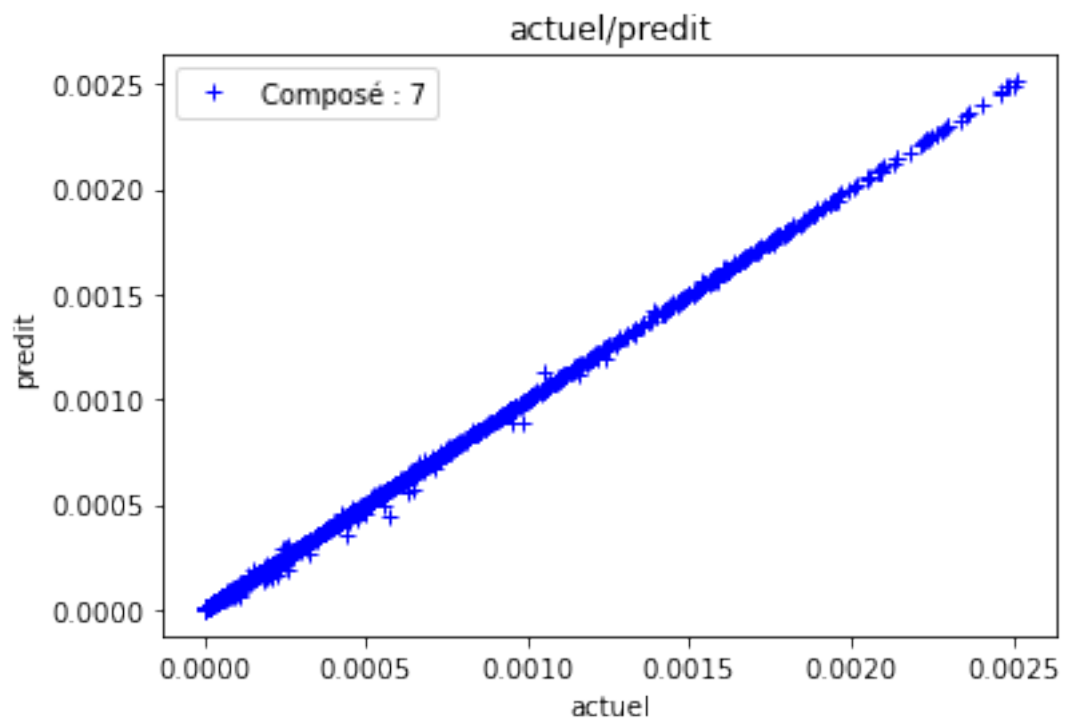
[20]: array([[5.83000000e-03, 1.04000000e-06, 9.74000000e-01, 3.23000000e-04,
1.07000000e-05, 1.56000000e-02, 1.63000000e-05, 3.11000000e-08,
3.53000000e-03, 1.52000000e-04, 1.25000000e-07, 1.14000000e-04,
5.58604167e-05, 2.62000000e-05, 1.20000000e-08, 4.78604167e-07,
5.17604167e-08, 7.36604167e-06, 1.42000000e-09, 1.35604167e-08,
2.17604167e-09, 3.10604167e-08, 2.89208333e-08, 1.38000000e-09,
1.62000000e-09, 8.29416667e-10, 7.08229167e-12, 3.01208333e-12,
1.15604167e-10, 2.34604167e-11, 3.02000000e-10]])

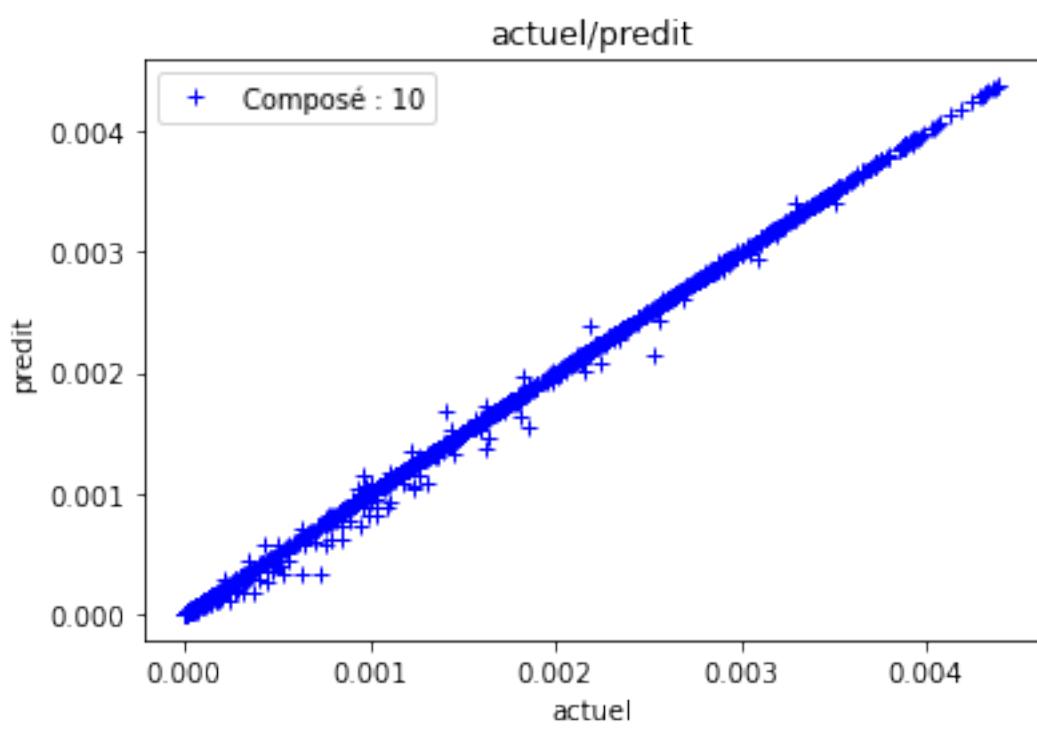
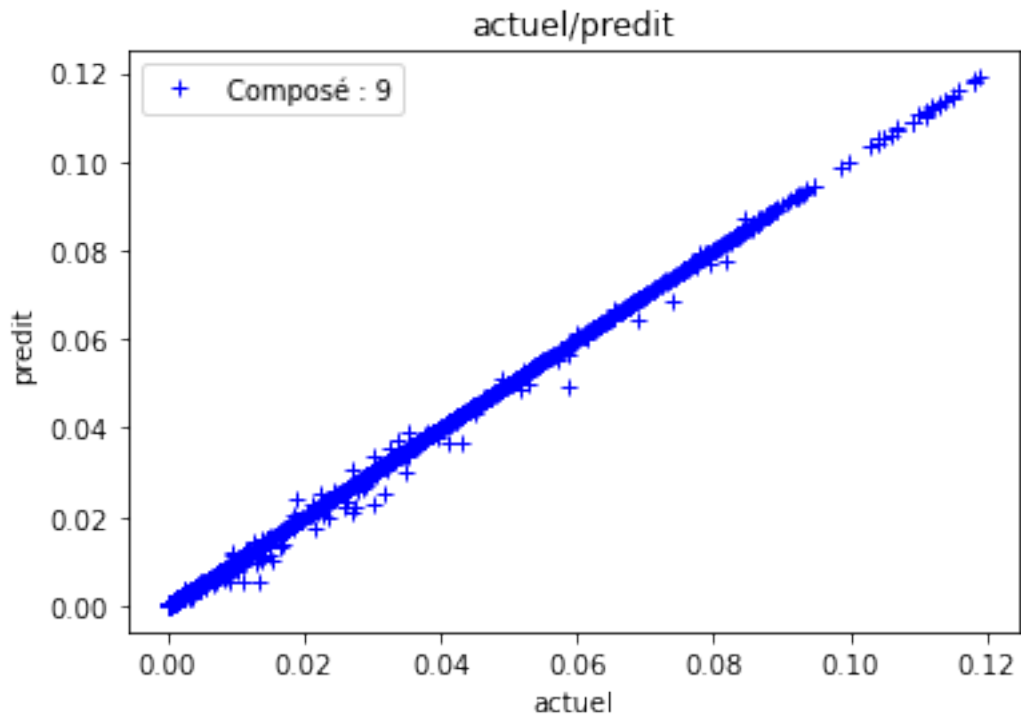
```
[21]: #  
# Courbes de comparaisons y_prédit vs y_test  
#  
from Ytest_Ypred_1 import Ytest_Ypred # Fonction personnelle  
#  
Ytest_Ypred( grid.best_estimator_, X_test , y_test )
```

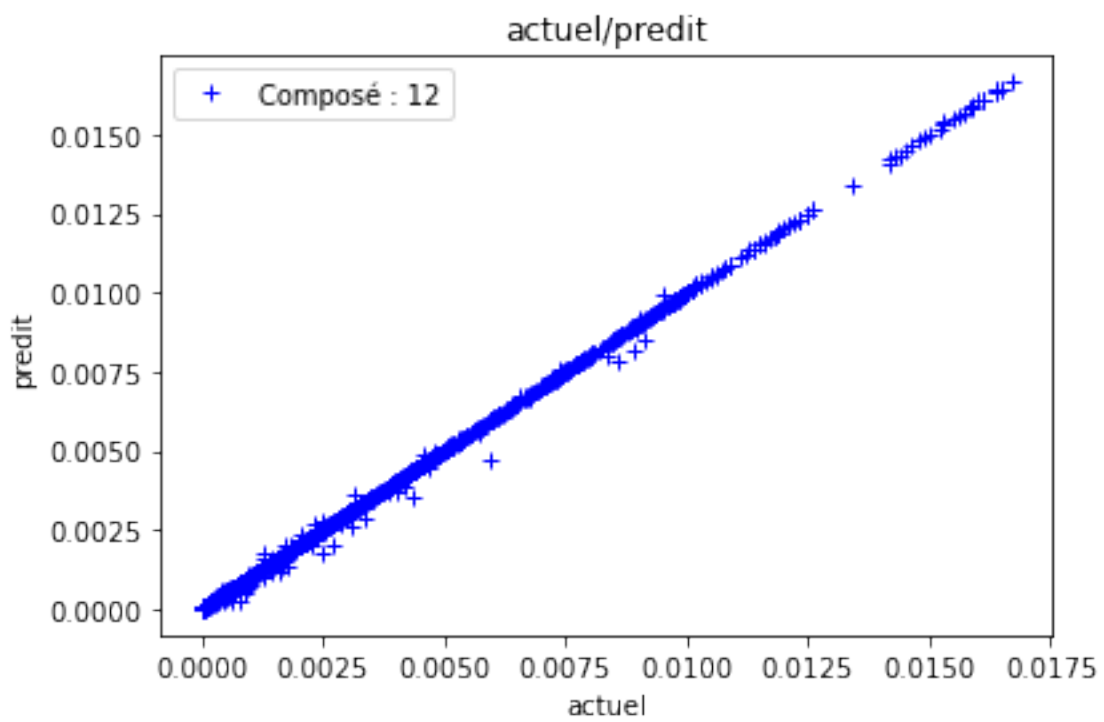
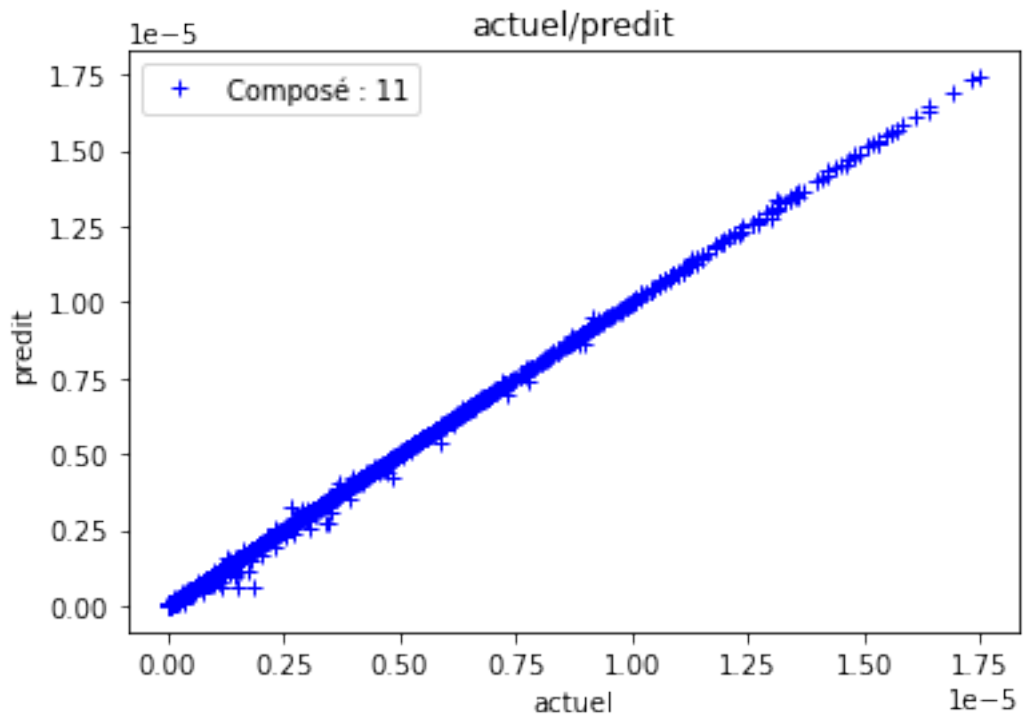


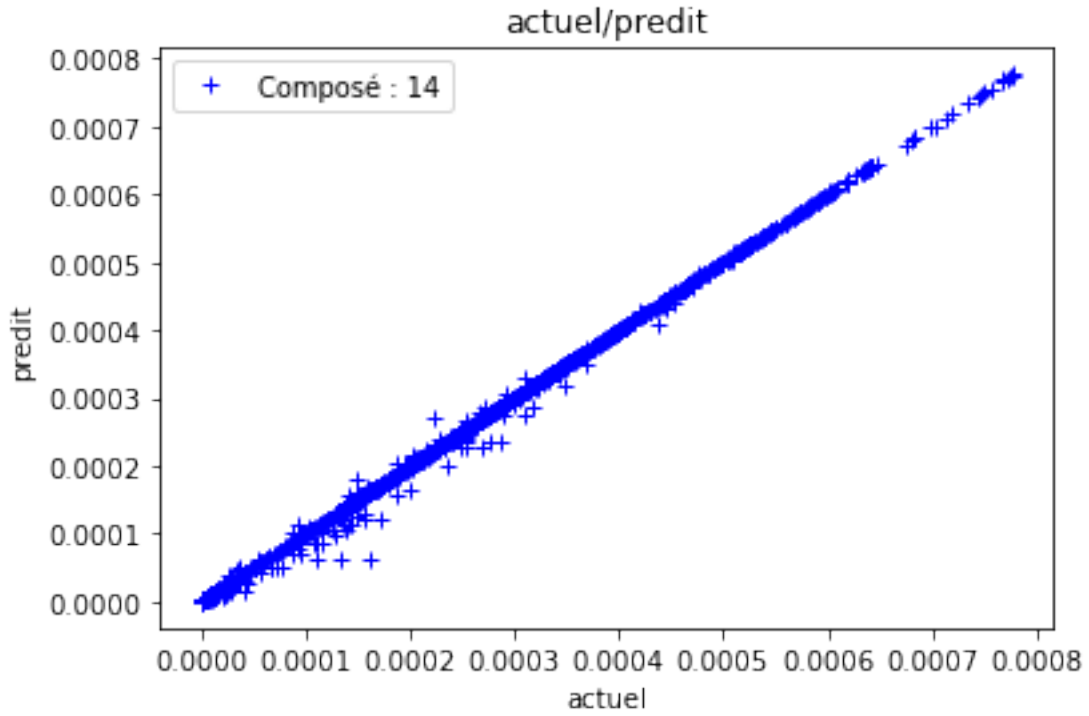
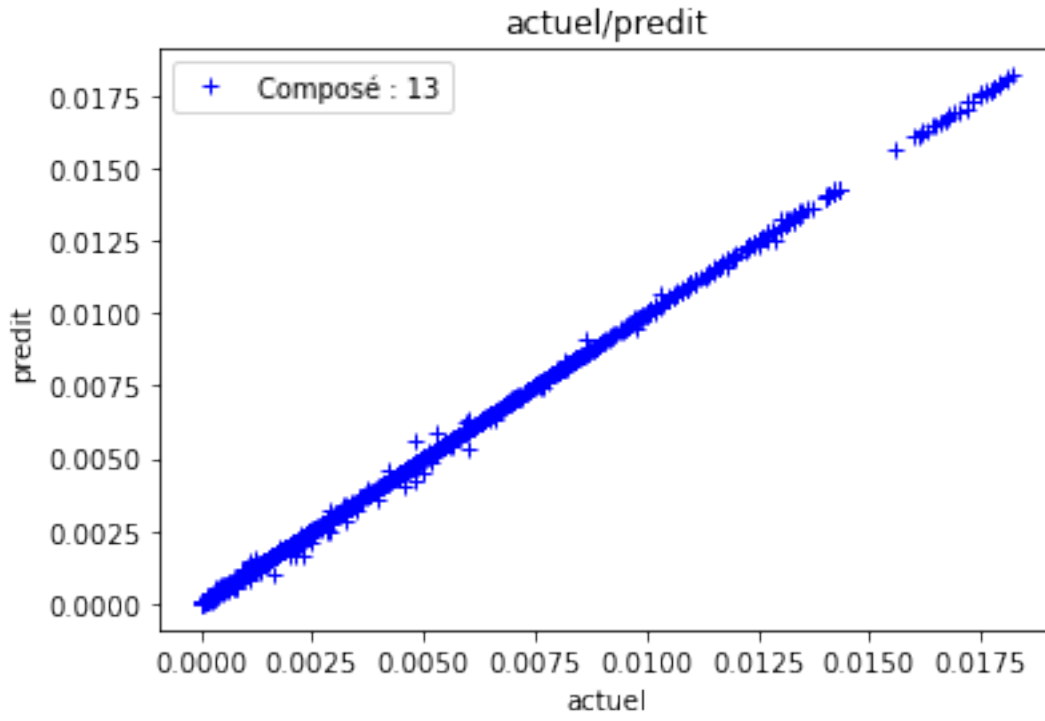


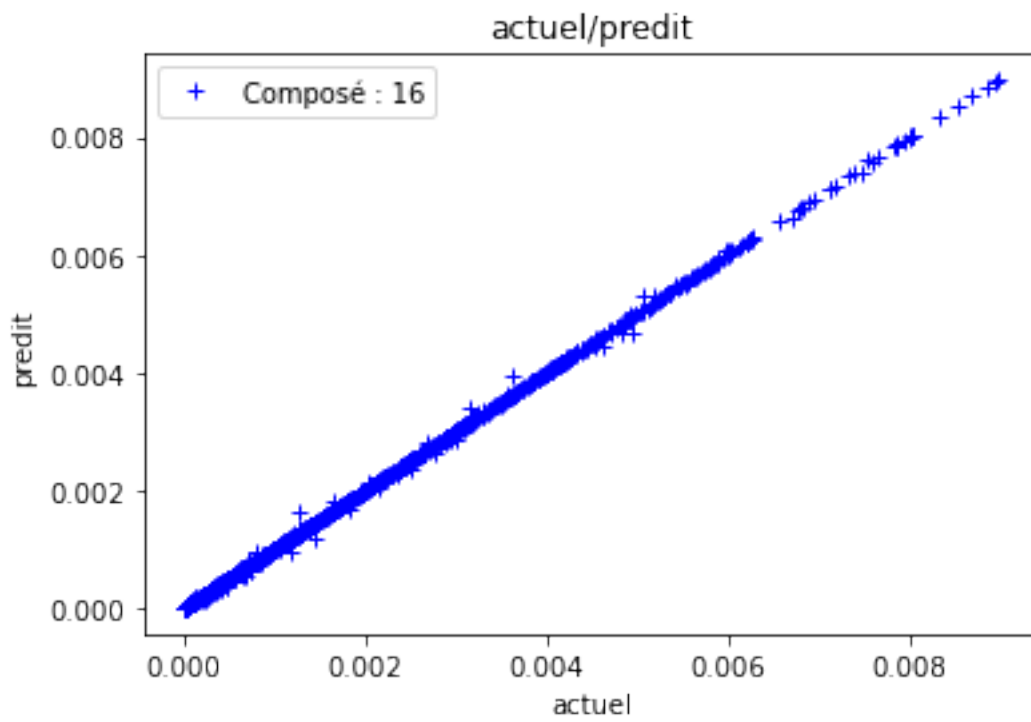
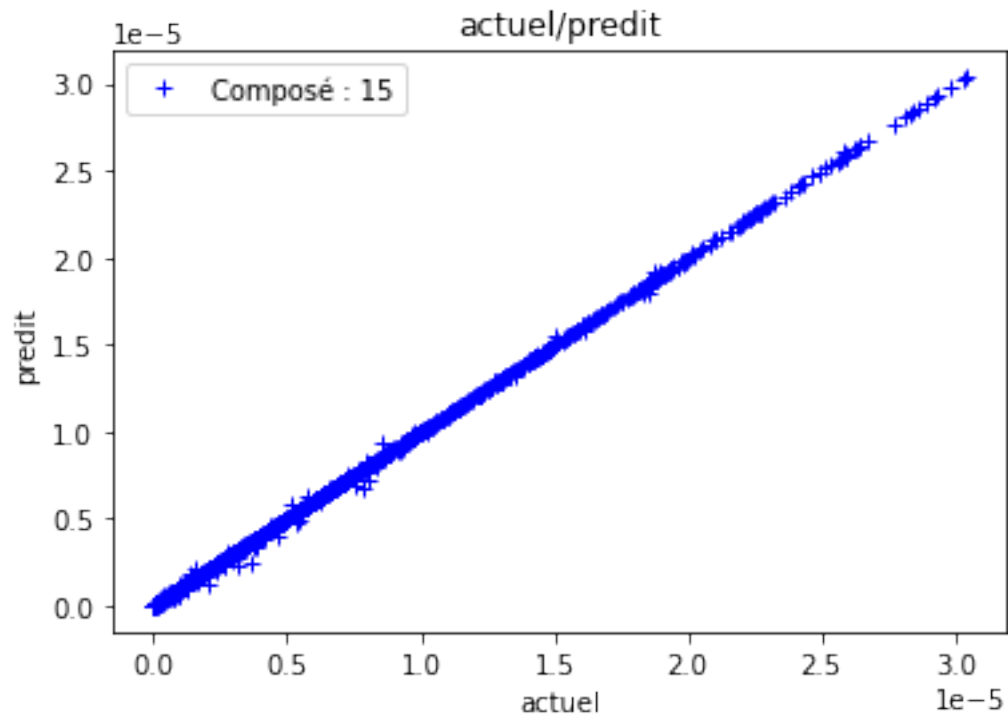


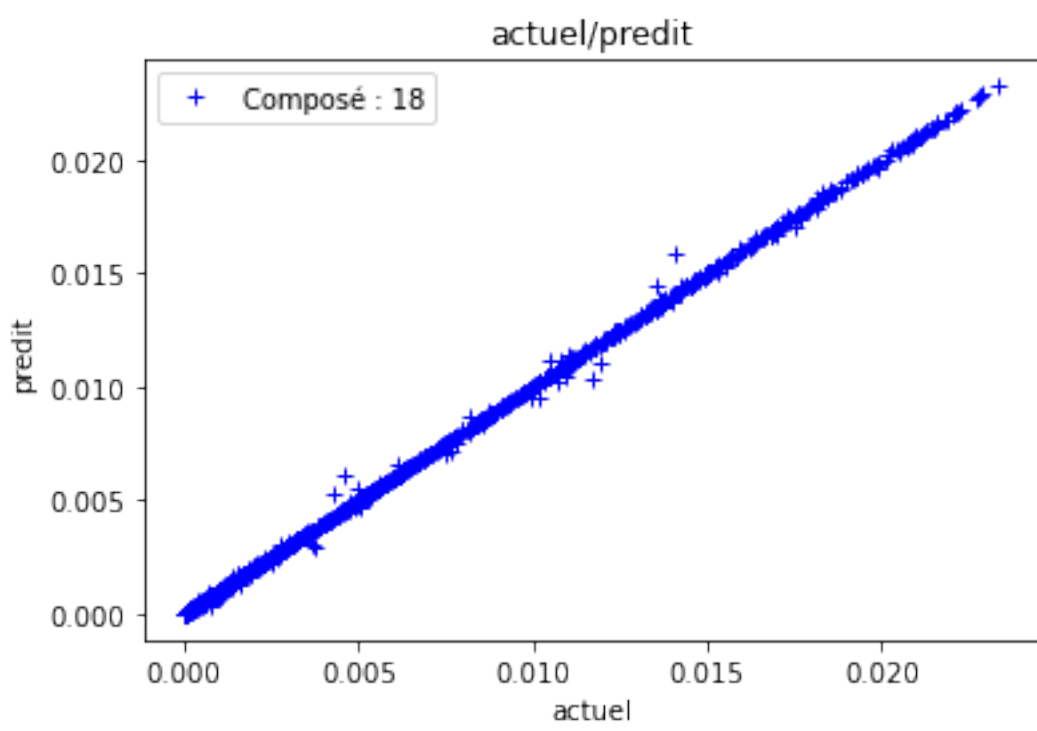
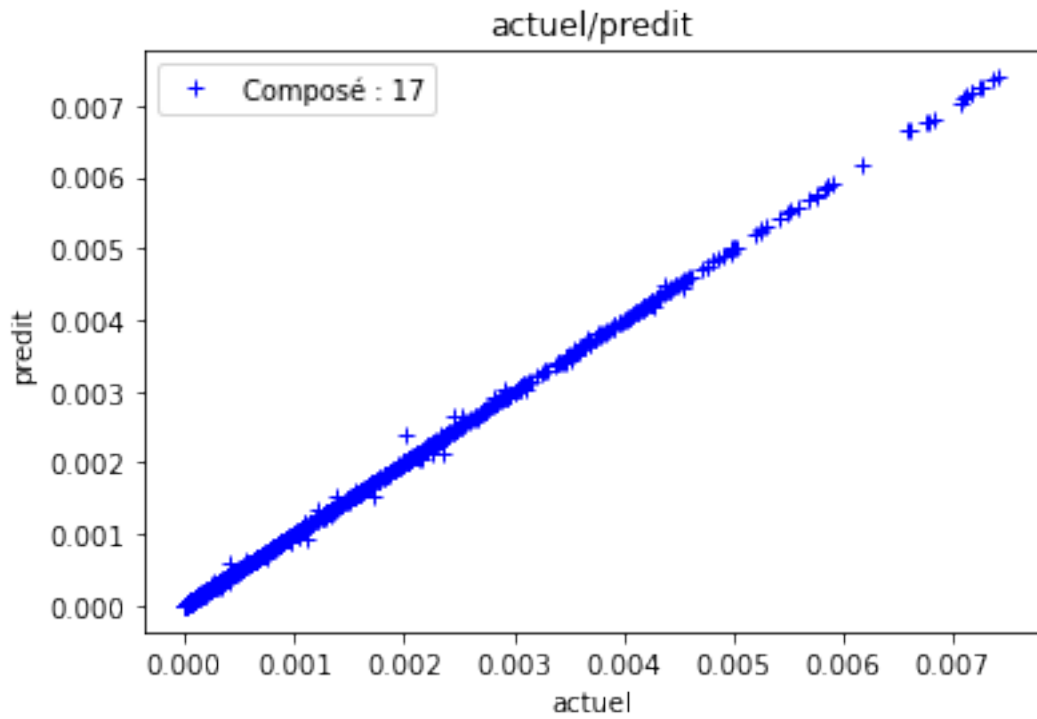


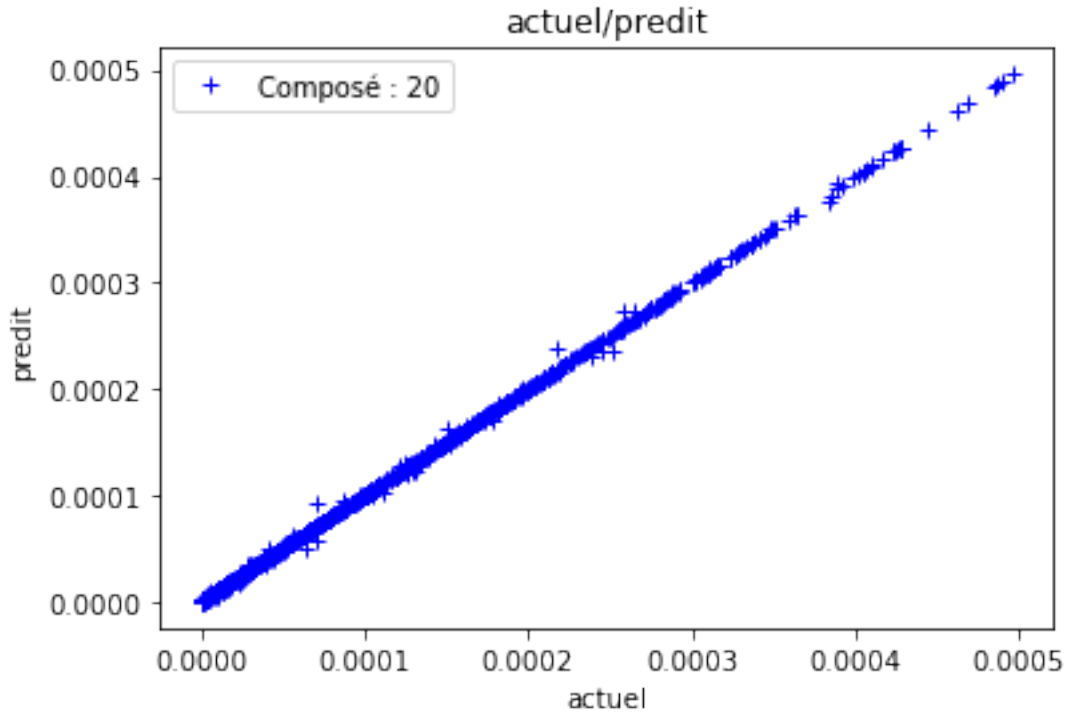
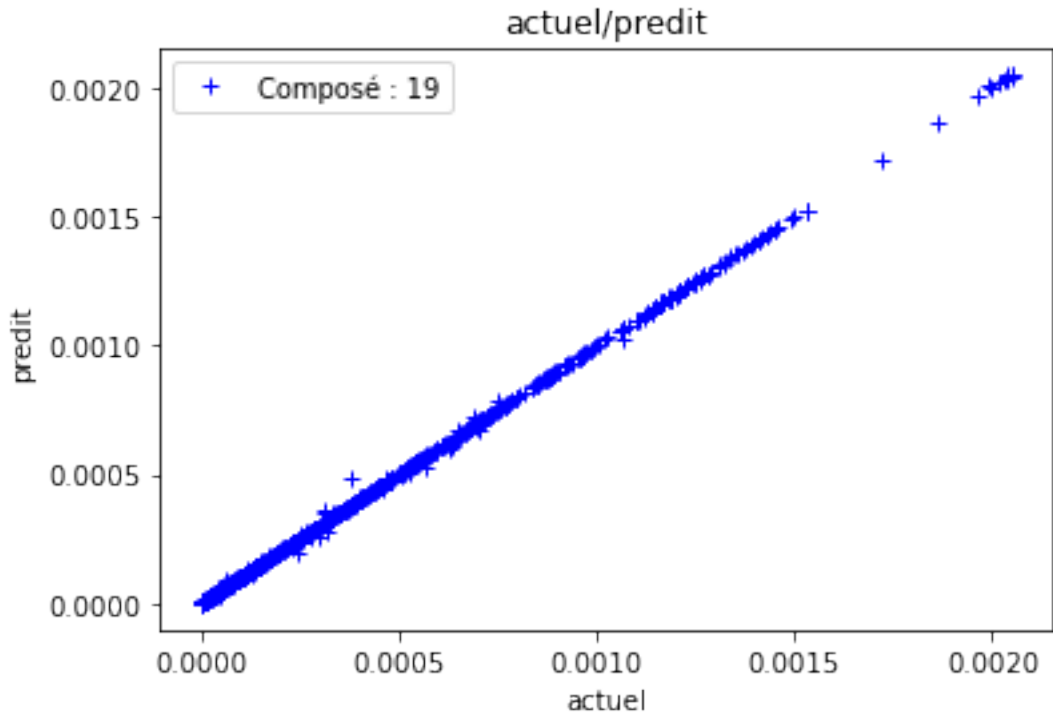


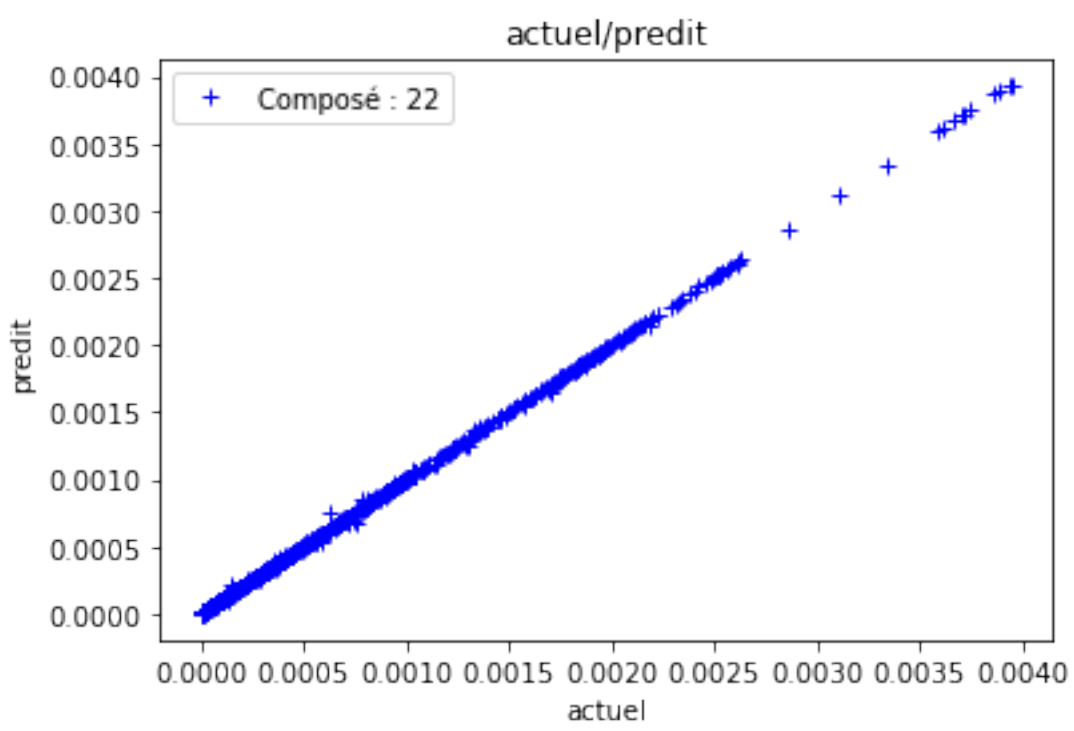
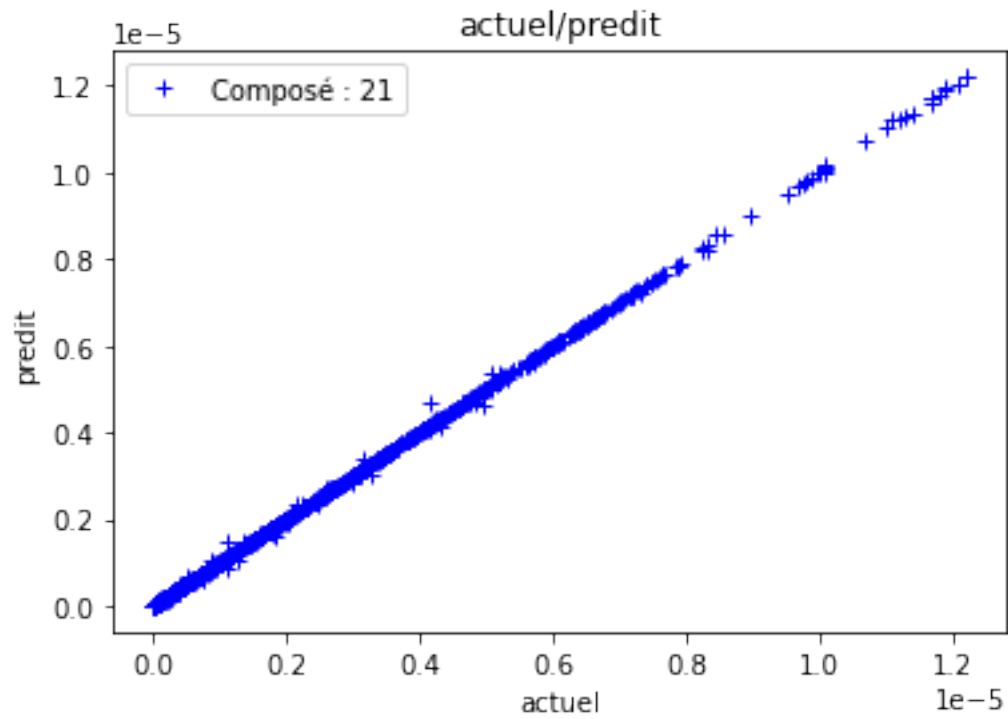


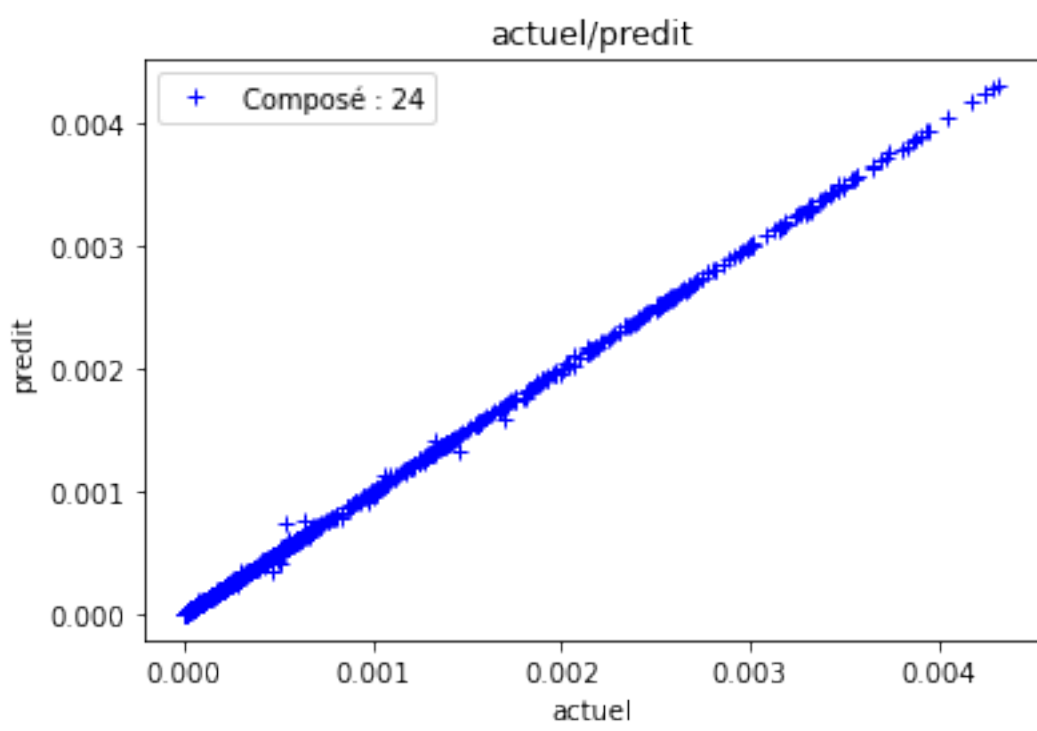
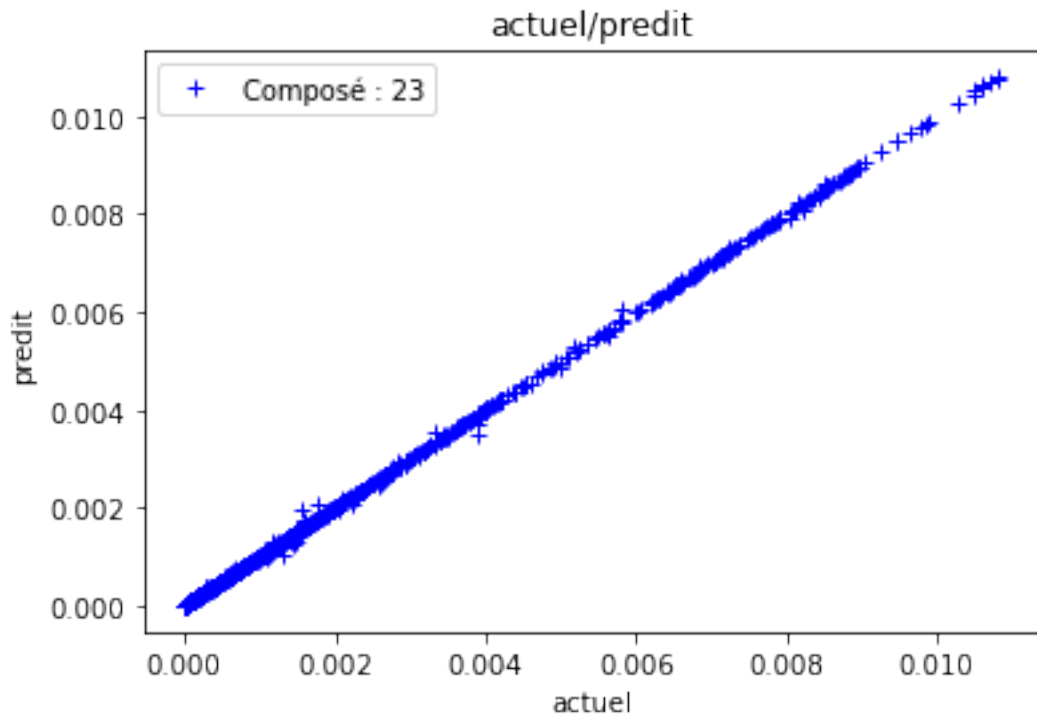


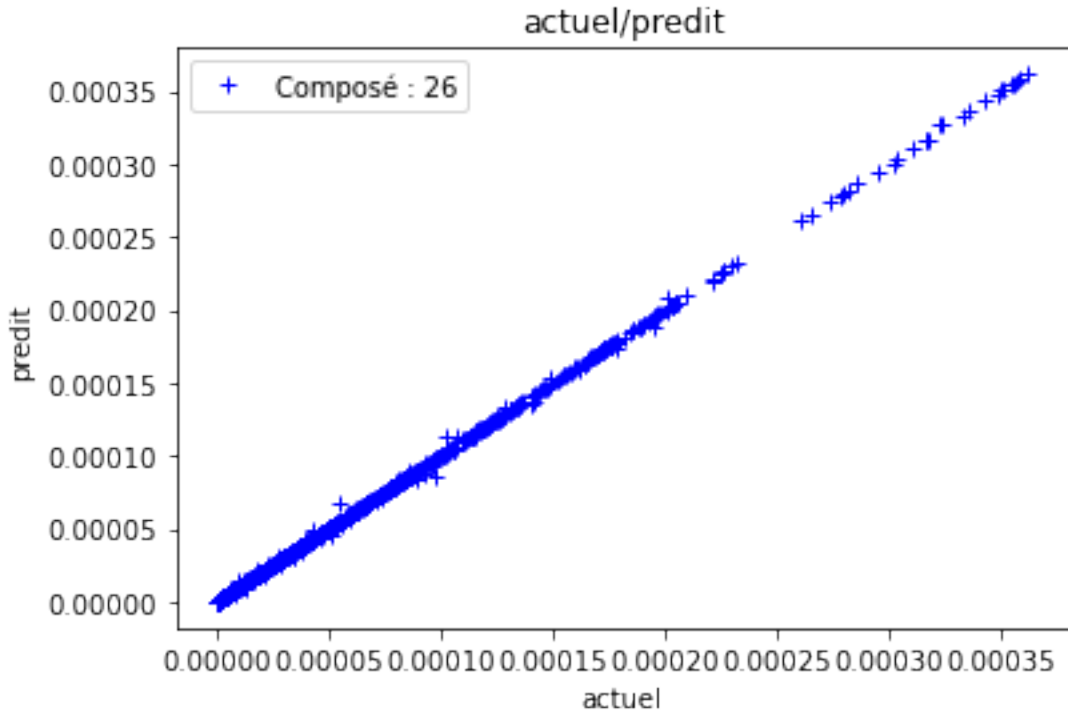
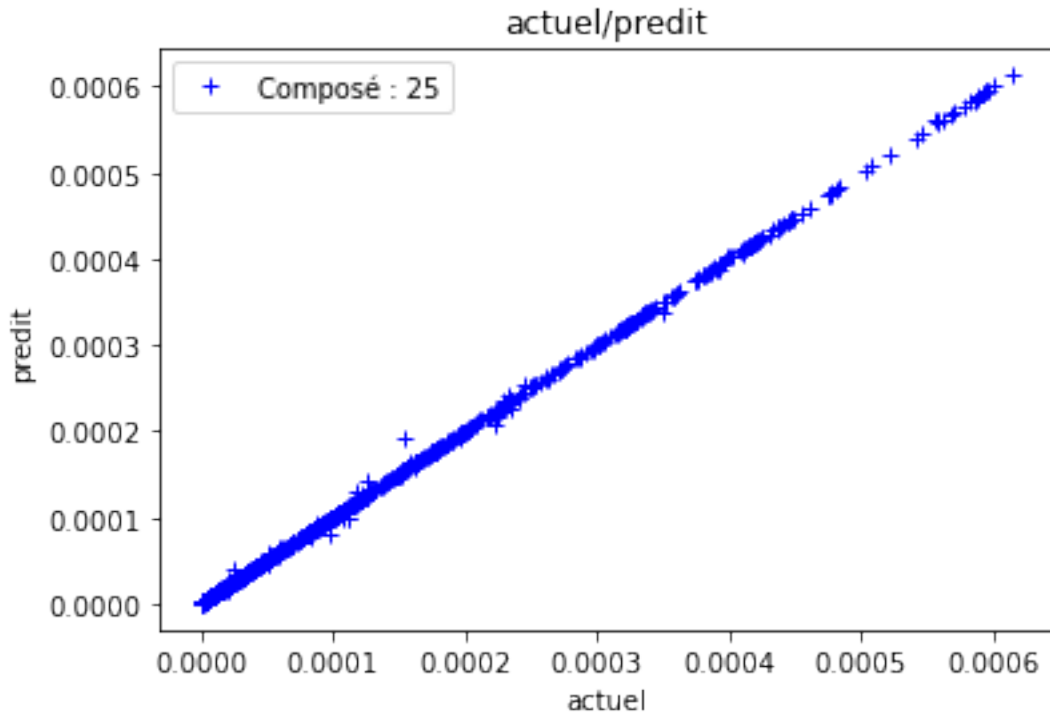


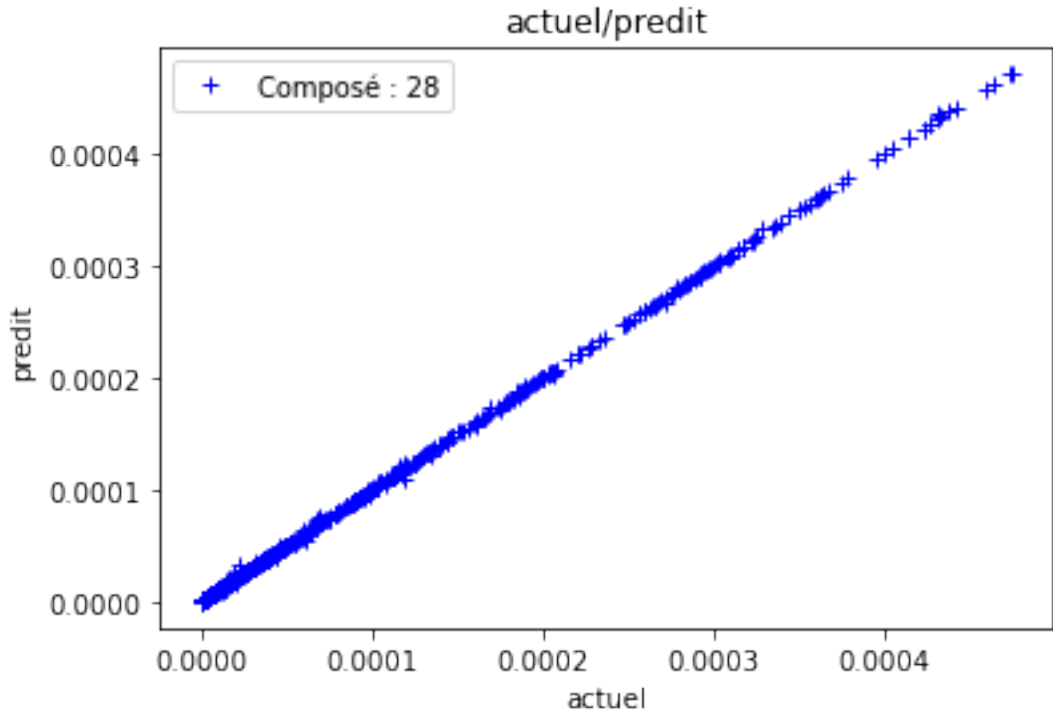
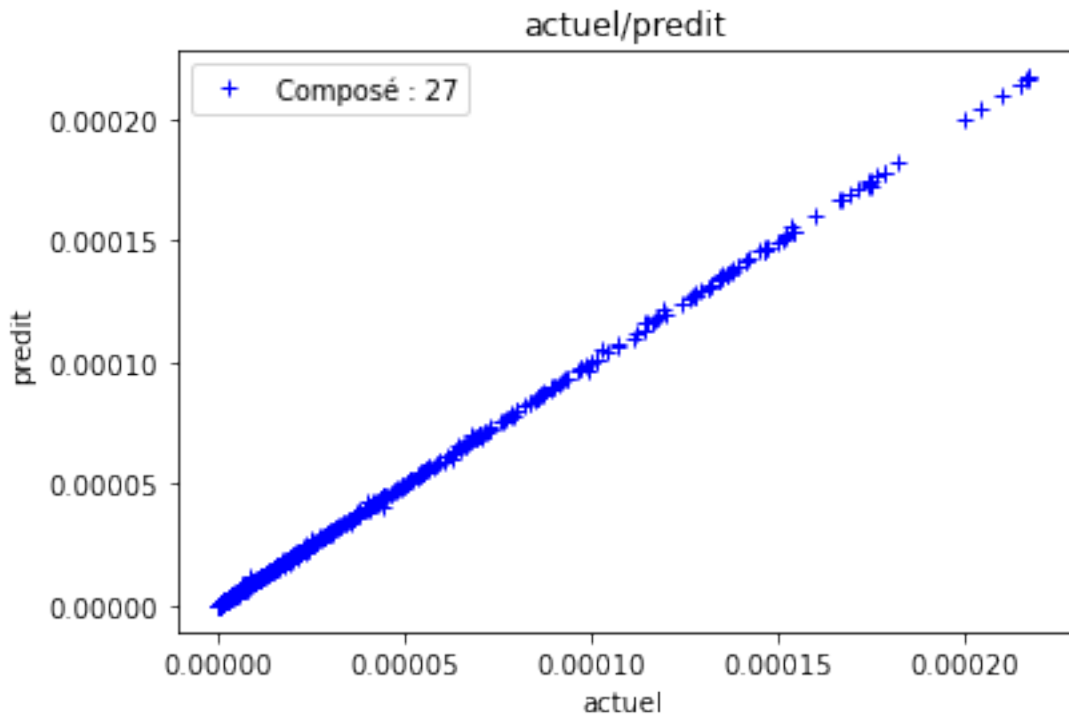


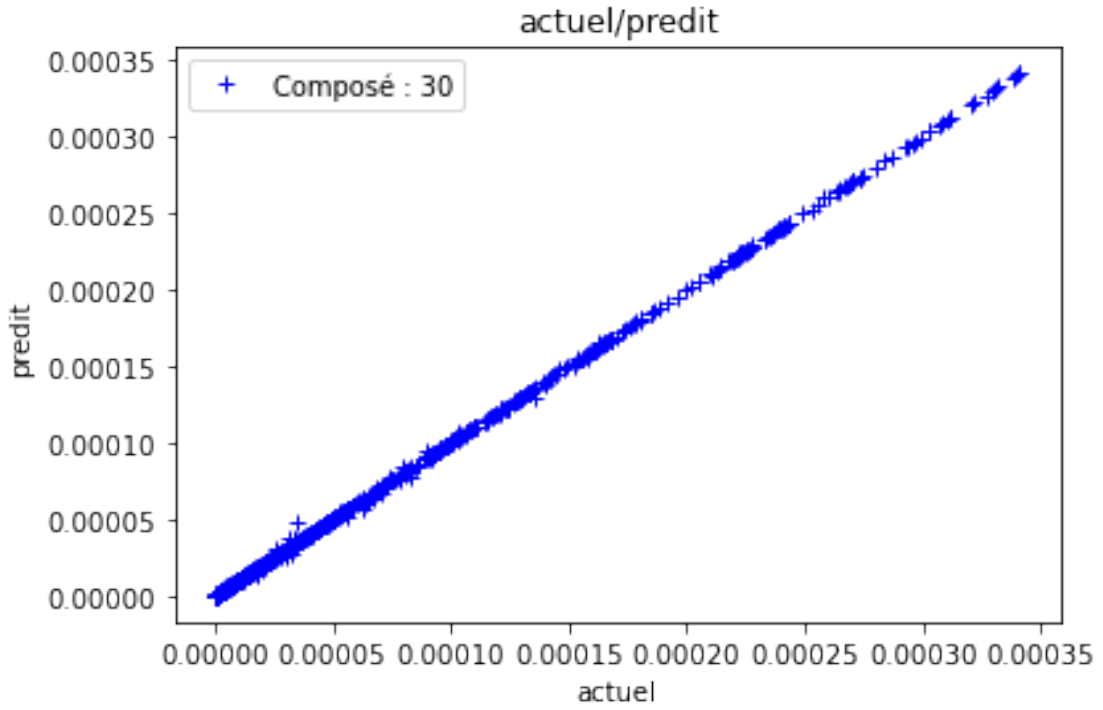
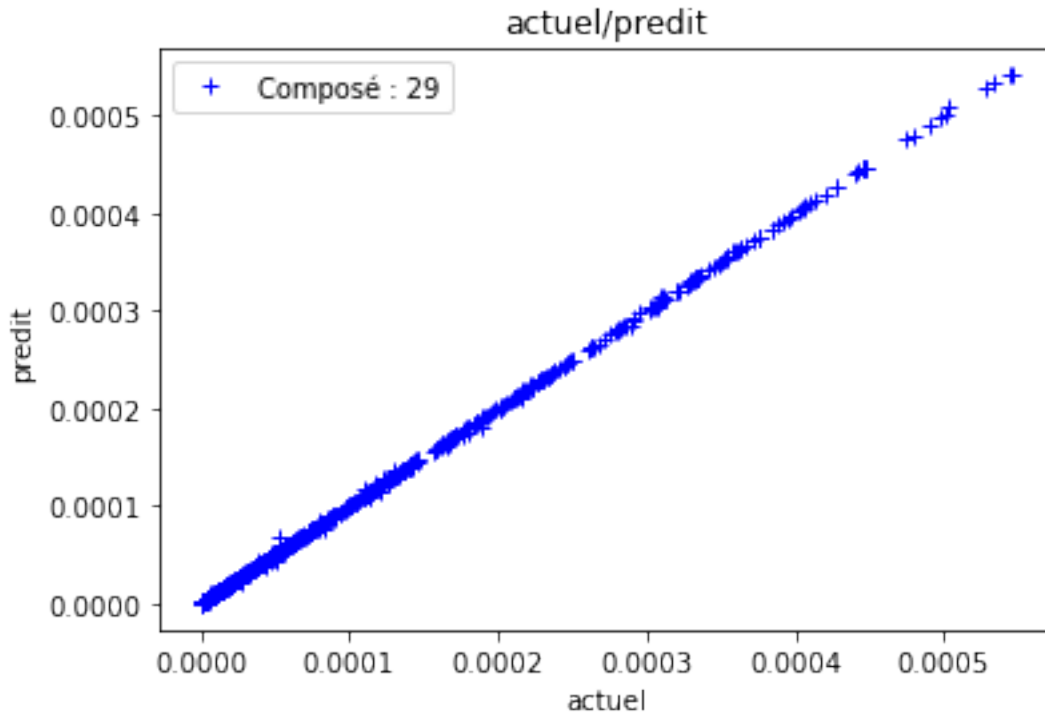


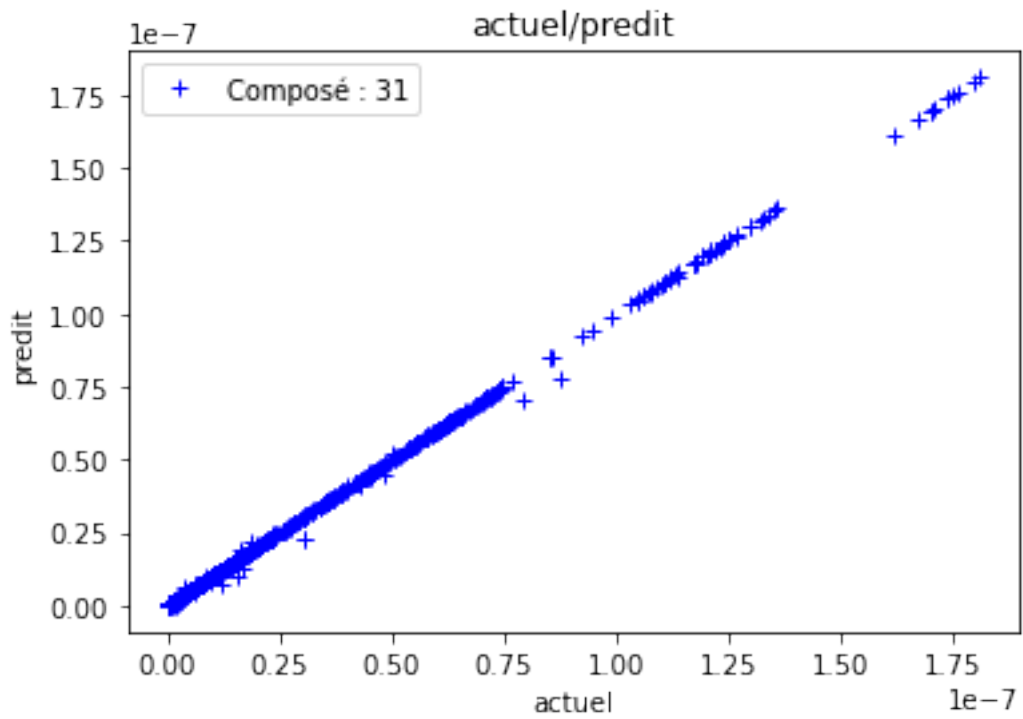






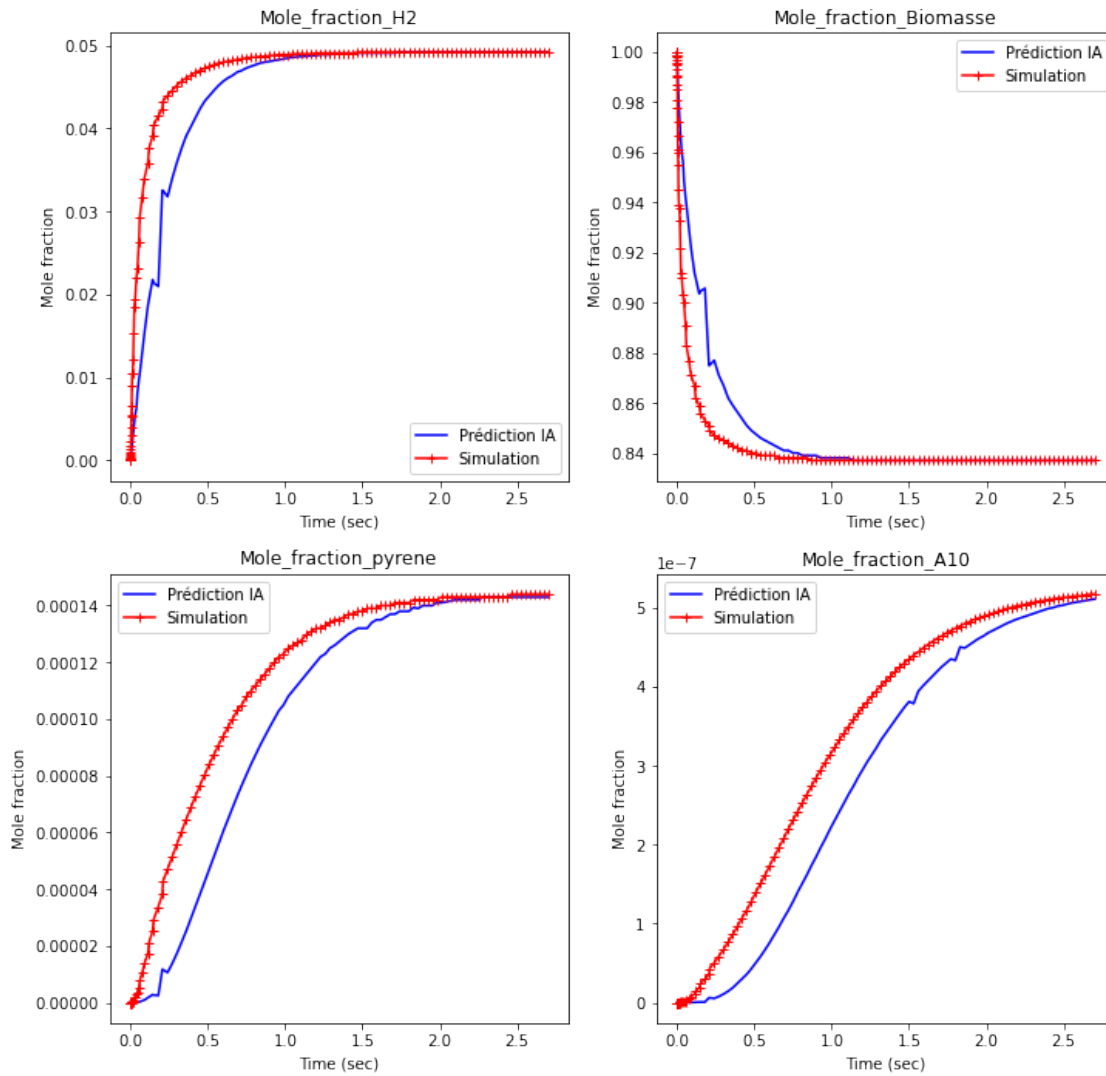






```
[22]: #
# Courbes de comparaison  $y_{test}=f(X_{test})$  vs  $y_{predit}=f(X_{prédit})$ 
#
from Courbes_Exp_Simu import comparaison_Exp_Sim_4 # Fonction personnelle
#
comparaison_Exp_Sim_4( grid.best_estimator_ , df , [0,2,16,29])
```

Comparaison Simulaton / Prédiction IA



```
[23]: #
# Joblib - exportation du modèle ML sous forme d'un fichier de compilation
#
import joblib
#
joblib.dump(grid.best_estimator_,'KNN.joblib')
```

```
[23]: ['KNN.joblib']
```

```
[24]: #
# Essai pour utilisation future
#
load_KNN = joblib.load('./KNN.joblib')
```

```
[25]: #  
      # Tests sur les métriques  
      #  
      print('R2_score')  
      print('KNN R2 train = ' , load_KNN.score(X_train,y_train))  
      print('KNN R2 test = ' , load_KNN.score(X_test,y_test))
```

```
R2_score  
KNN R2 train = 0.9999999995148973  
KNN R2 test = 0.9997595859006833
```