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## KNN (150 WORDS BY CODE)

```
TRAIN_R2_SCORE = 0.8638242265848174
```

```
TEST_R2_SCORE = 0.8516379991490913
```

```
BEST_PARAMS: [{'N_NEIGHBORS': 10}]
```

```
estimator KNeighborsClassifier()
```

```
params {'n_neighbors': [10]}
```

---

```
df.shape : (82265, 4052)
```

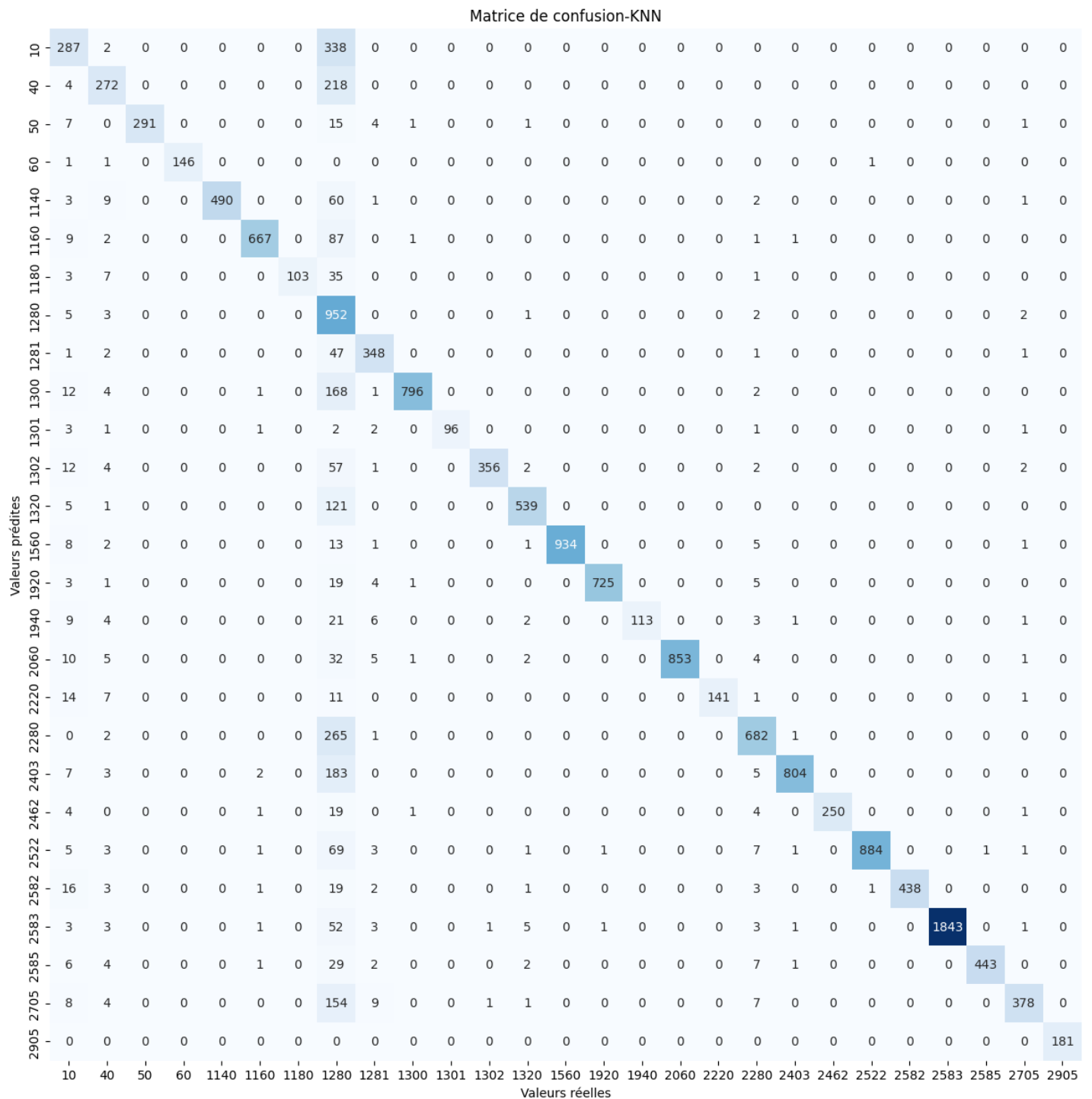
```
X_train.shape - X_test.shape - len(y_train) - len(y_test)
```

```
(65812, 4050) - (16453, 4050) - 65812 - 16453
```

```
=====CONFUSION MATRIX=====
```

3. Use SEABORN to draw confusion\_matrix-----

Confusion matrix as graph with Seaborn :



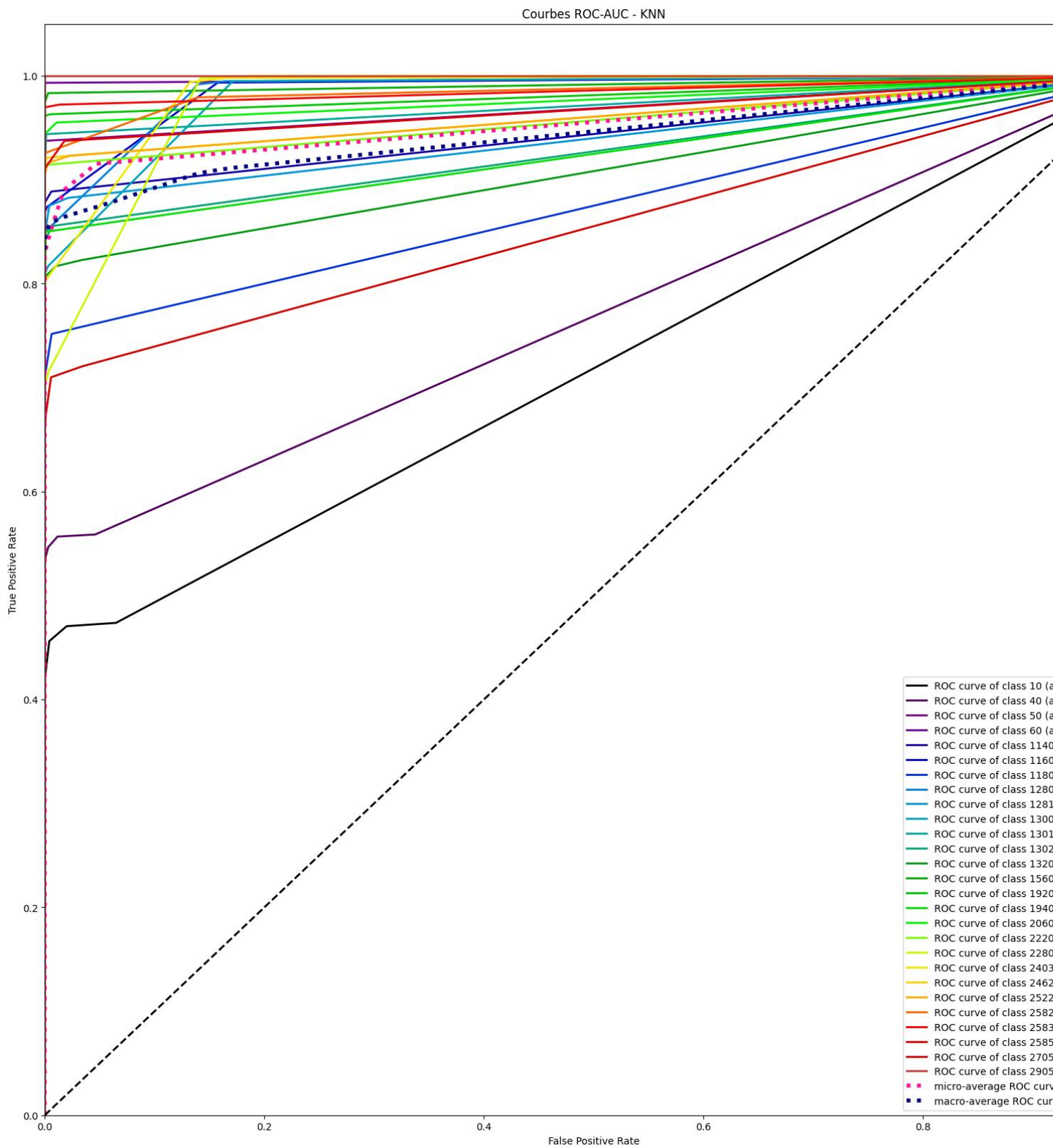
```
train_f1_score = [array([0.57901204, 0.68607825, 0.95127796, 0.98703404, 0.93035079,
0.93022476, 0.87568556, 0.48681333, 0.88793103, 0.90987821,
0.93099671, 0.92581944, 0.88107058, 0.98251479, 0.98434668,
0.89974293, 0.97252903, 0.94146744, 0.816935 , 0.87853233,
0.95158287, 0.95363889, 0.95509992, 0.98576165, 0.95005429,
0.82464956, 0.99855072]))]
```

```
test_f1_score = [array([0.53544776, 0.64531435, 0.95253682, 0.98983051, 0.9280303 ,
0.92382271, 0.81746032, 0.48190332, 0.8776797 , 0.89187675,
0.94581281, 0.89672544, 0.88071895, 0.98367562, 0.97643098,
```

0.82783883, 0.96602492, 0.89240506, 0.80282519, 0.88643881,  
0.94339623, 0.94900698, 0.95010846, 0.98031915, 0.94355698,  
0.79162304, 1.     ]]]

train\_mse\_result = 160215.49750805323

test\_mse\_result = 184986.59539293745



## KNN (150 WORDS BY CODE)

```
ESTIMATOR KNEIGHBORSCLASSIFIER()
```

```
PARAMS {'N_NEIGHBORS': [10]}
```

```
TRAIN_R2_SCORE = 0.8857199294961405
```

```
TEST_R2_SCORE = 0.8786847383455905
```

-----

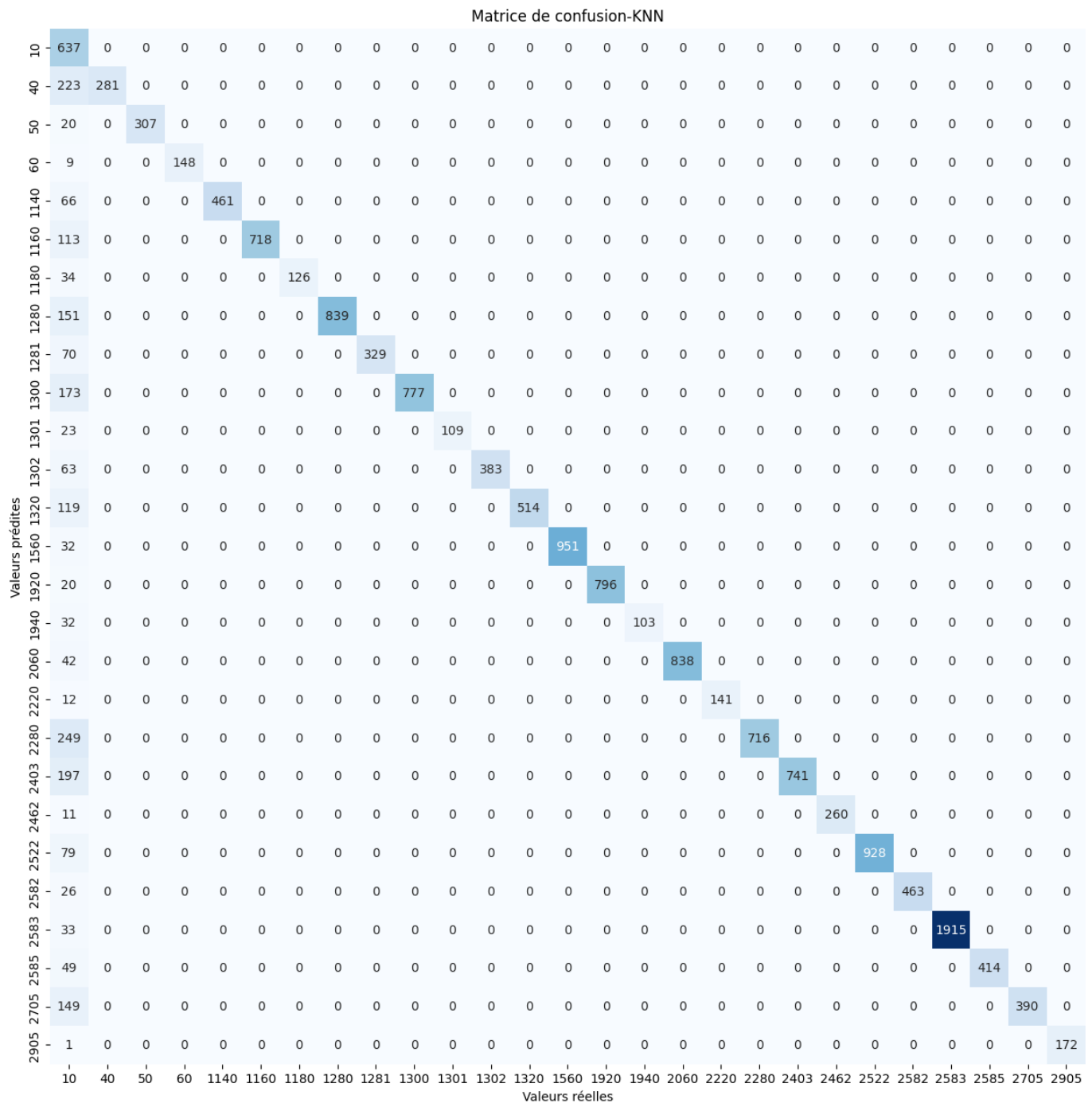
```
X_train.shape - X_test.shape - len(y_train) - len(y_test)
```

```
(65812, 4050) - (16453, 4050) - 65812 - 16453
```

Fitting 3 folds for each of 1 candidates, totalling 3 fits

=====CONFUSION MATRIX=====

3. Use SEABORN to draw confusion\_matrix-----



```
train_f1_score = [array([0.39711423, 0.72972973, 0.96431404, 0.99018003, 0.94329389,
0.93843537, 0.87940631, 0.9235361 , 0.92734032, 0.91129685,
0.94292237, 0.92778741, 0.89739729, 0.99012947, 0.98742666,
0.92679002, 0.9777964 , 0.95019763, 0.83607313, 0.89386929,
0.97977528, 0.96024384, 0.97210136, 0.98762054, 0.96360759,
0.83718487, 0.99928418]))]
```

```
test_f1_score = [array([0.38960245, 0.71592357, 0.96845426, 0.9704918 , 0.93319838,
0.92704971, 0.88111888, 0.91744122, 0.90384615, 0.89982629,
0.90456432, 0.92400483, 0.89625109, 0.98345398, 0.98759305,
0.86554622, 0.97555297, 0.95918367, 0.85187388, 0.88266825,
```

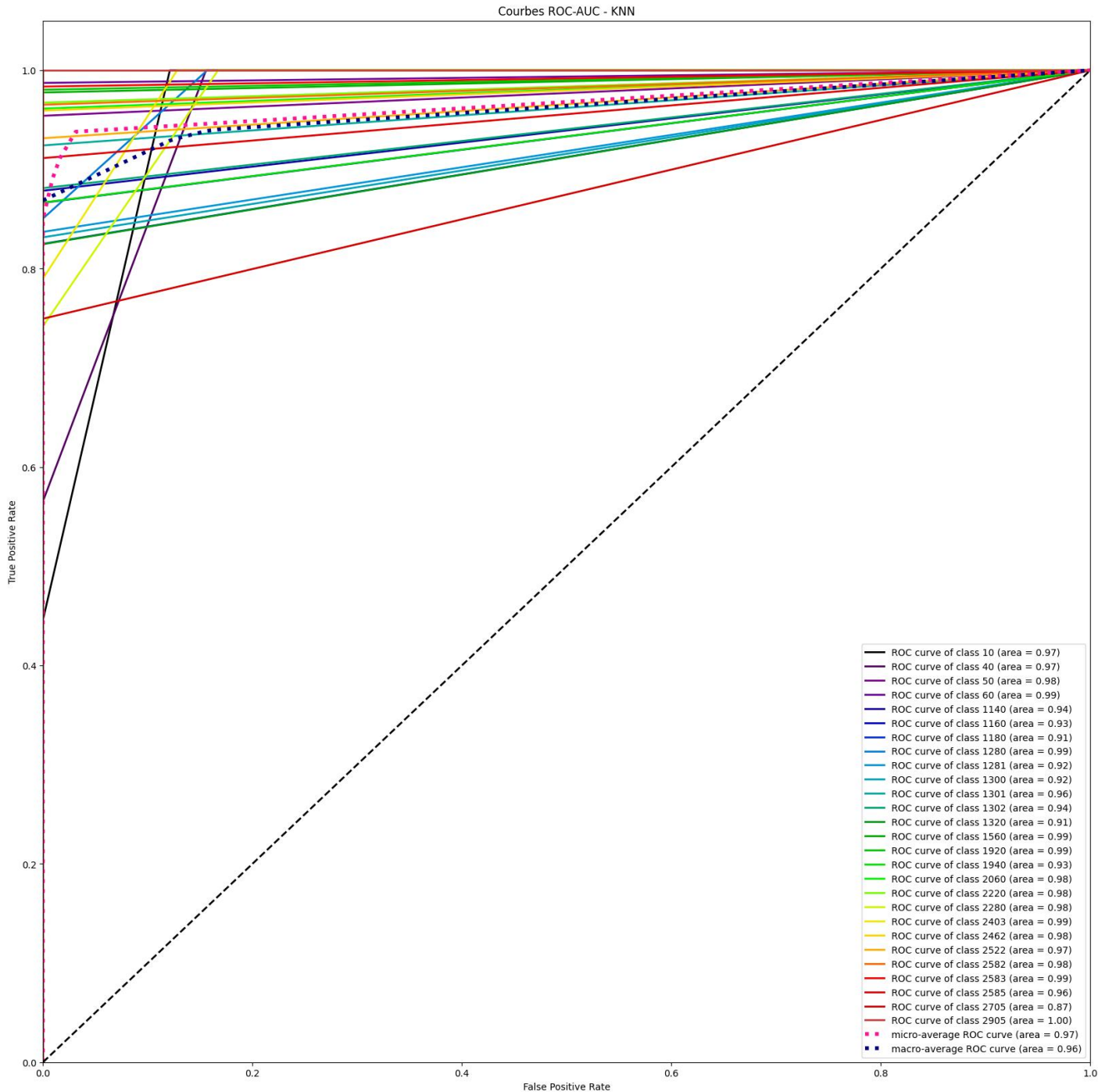


0.97928437, 0.95917313, 0.97268908, 0.99145742, 0.94412771,  
0.83961249, 0.99710145]]]

train\_mse\_result = 389357.61490305717

test\_mse\_result = 398629.87017565186

best\_params: [{'n\_neighbors': 10}]



## KNN (150 WORDS CODE) AVEC SCALING

ESTIMATOR      KNEIGHBORSCLASSIFIER()

PARAMS      {'N\_NEIGHBORS': [10, 12, 30]}

**TRAIN\_R2\_SCORE = 0.8887436941591199**

**TEST\_R2\_SCORE = 0.88160213942746**

-----  
df.shape : (82265, 4052)

Fitting 3 folds for each of 3 candidates, totalling 9 fits

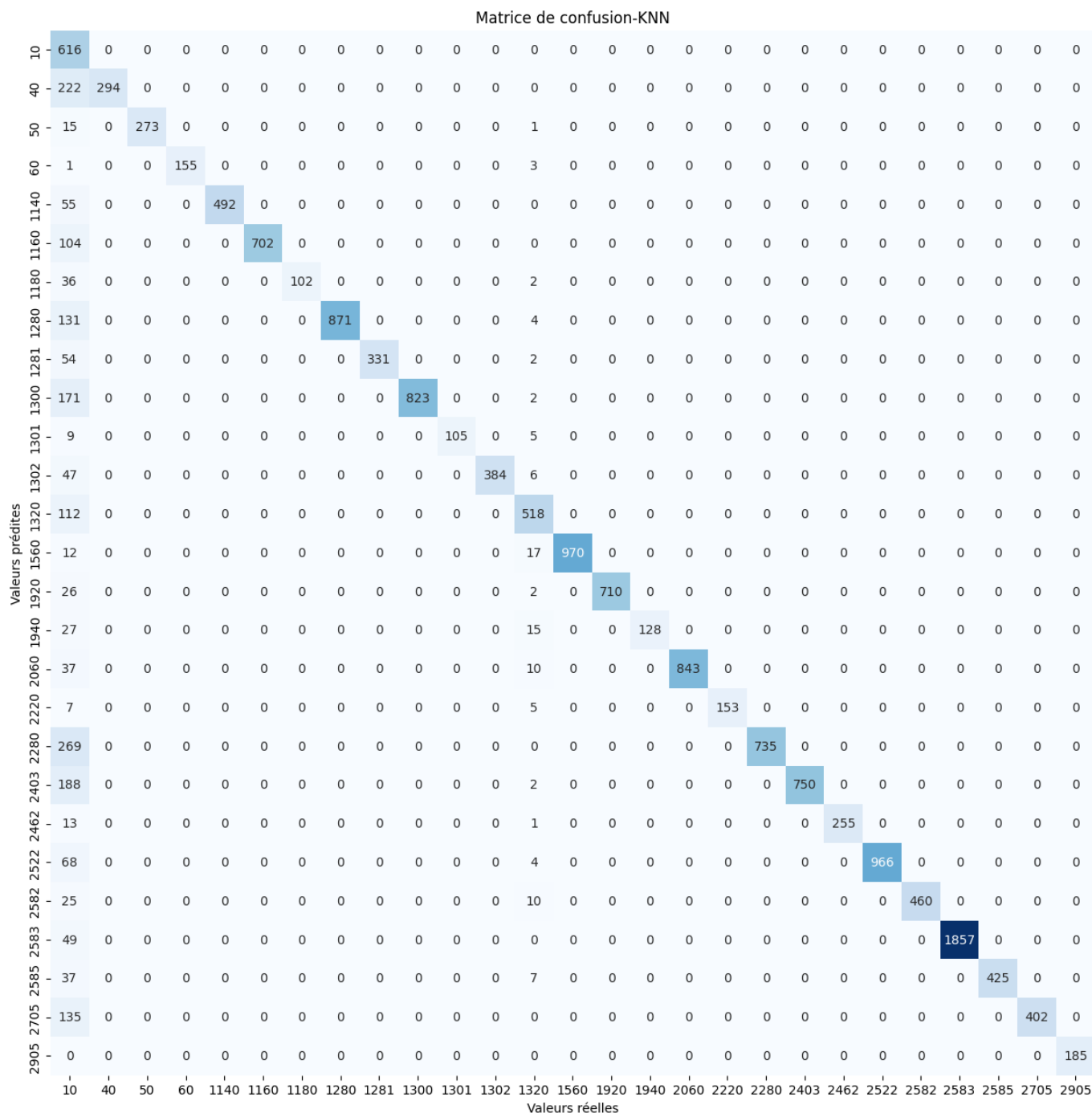
X\_train.shape - X\_test.shape - len(y\_train) - len(y\_test)

(65812, 4050) - (16453, 4050) - 65812 - 16453

=====CONFUSION MATRIX=====

Use SEABORN to draw confusion\_matrix-----

Confusion matrix as graph with Seaborn :



```
train_f1_score = [array([0.41769083, 0.73500967, 0.9689298 , 0.99673736, 0.9419387 ,
0.93805907, 0.90718039, 0.92072124, 0.92792491, 0.91580663,
0.95499451, 0.93367639, 0.83938852, 0.98852649, 0.9897277 ,
0.92972058, 0.97787735, 0.9542903 , 0.83787973, 0.89236564,
0.98163905, 0.95949739, 0.97016461, 0.98893276, 0.96046697,
0.85121825, 0.99854227]))]
```

```
test_f1_score = [array([0.39974043, 0.72592593, 0.97153025, 0.98726115, 0.94706449,
0.93103448, 0.84297521, 0.92807672, 0.92200557, 0.9048928 ,
0.9375 , 0.93544458, 0.83146067, 0.98527171, 0.98066298,
0.8590604 , 0.9728794 , 0.96226415, 0.8453134 , 0.88757396,
```

0.97328244, 0.96407186, 0.96335079, 0.98697847, 0.950783 ,

0.85623003, 1. ]]

train\_mse\_result = 361571.8485230657

test\_mse\_result = 382027.62973317935

**best\_params: [{'n\_neighbors': 10}]**

## KNN (100 WORDS BY CODE)

```
TRAIN_R2_SCORE = 0.8861802979450039
```

```
TEST_R2_SCORE = 0.8843028732925106
```

```
BEST_PARAMS: [{'ALGORITHM': 'AUTO', 'N_JOBS': -1, 'N_NEIGHBORS': 10, 'WEIGHTS': 'DISTANCE'}]
```

```
X_train.shape - X_test.shape - len(y_train) - len(y_test)
```

```
(67932, 2700) - (16984, 2700) - 67932 - 16984
```

```
estimator      KNeighborsClassifier()
```

```
params  {'n_neighbors': [10], 'weights': ['uniform', '...
```

-----

Fitting 3 folds for each of 2 candidates, totalling 6 fits

```
train_f1_score = [array([0.39089334, 0.76324655, 0.95494071, 0.99925981, 0.92794814,
```

```
    0.93105779, 0.8762421 , 0.90616622, 0.91707317, 0.91470786,
```

```
    0.95813953, 0.94    , 0.90372272, 0.98390572, 0.98402839,
```

```
    0.95230126, 0.97602475, 0.97179694, 0.81697044, 0.91878173,
```

```
    0.97751799, 0.95299539, 0.98472906, 0.98477977, 0.959442 ,
```

```
    0.85405961, 1.    ])]
```

```
test_f1_score = [array([0.39637953, 0.73316062, 0.94256259, 0.99678457, 0.93346981,
```

```
    0.94455578, 0.92830189, 0.9010503 , 0.87483871, 0.92225201,
```

```
    0.9453125 , 0.9376392 , 0.90306947, 0.98521698, 0.96850862,
```

```
    0.94642857, 0.98052921, 0.93103448, 0.81997372, 0.92016083,
```

```
    0.97472924, 0.94807892, 0.97773475, 0.98227216, 0.9600863 ,
```

```
    0.8380744 , 0.996997 ])]
```

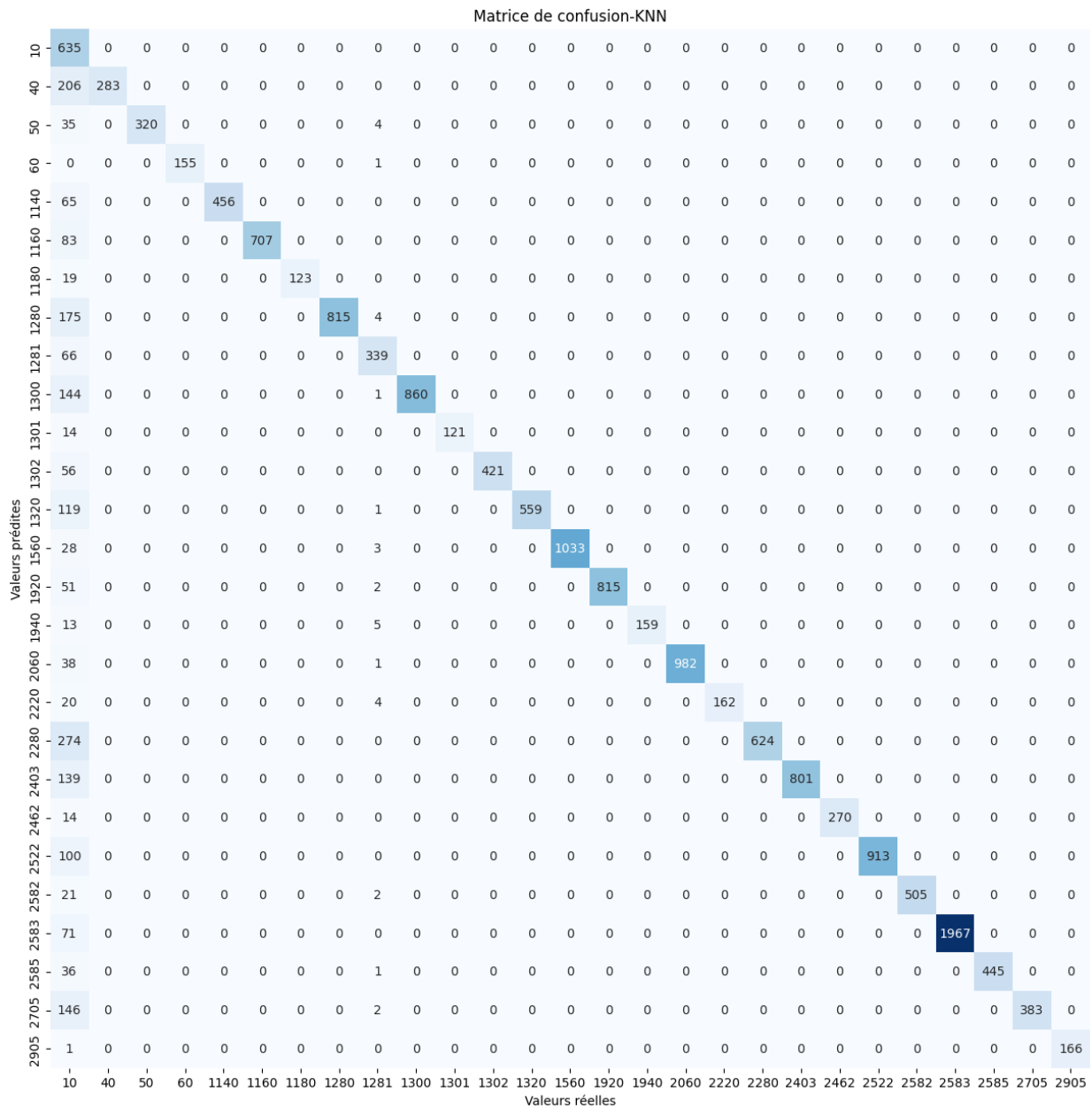
```
train_mse_result = 382045.8192162751
```

```
test_mse_result = 388084.74004945834
```

=====CONFUSION MATRIX=====

Use SEABORN to draw confusion\_matrix-----

Confusion matrix as graph with Seaborn :



KNN (100 WORDS BY CODE) APRES UNE PCA (REDUCTION DE 80% DES VARIABLES) – 2MIN

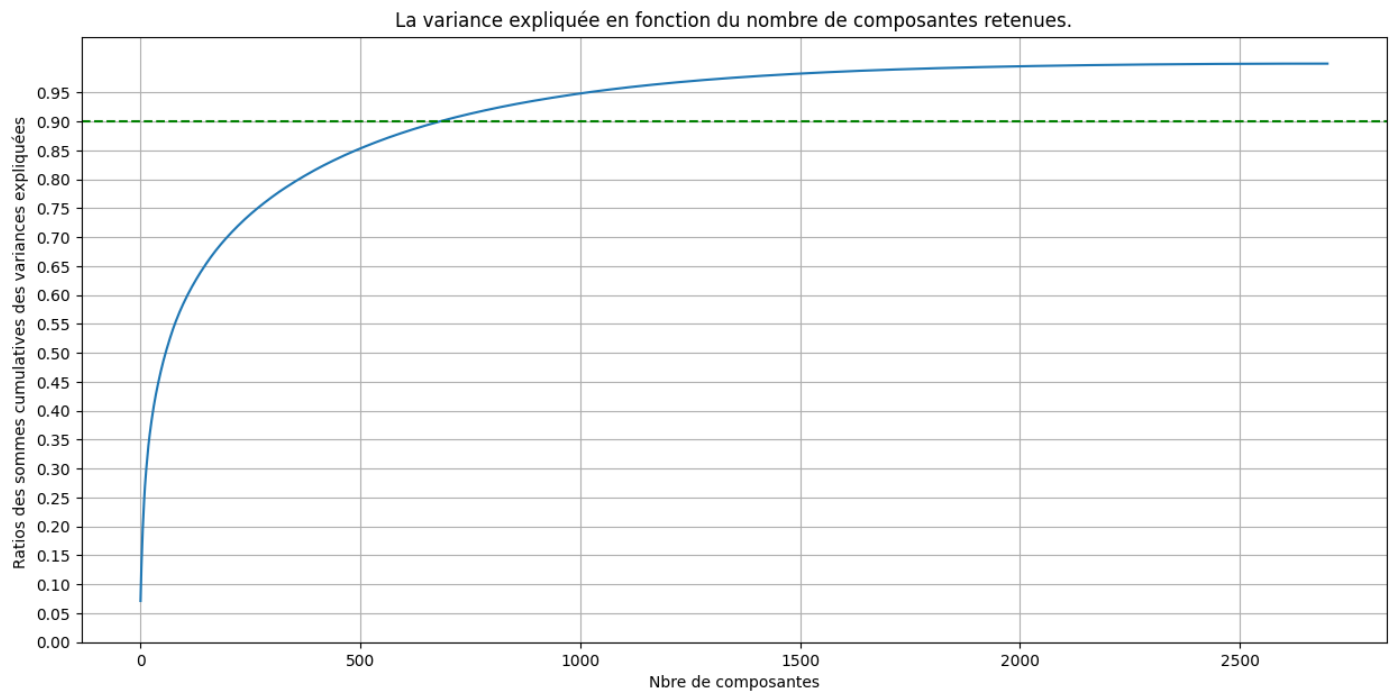
TRAIN\_R2\_SCORE = 0.8463046252962986

TEST\_R2\_SCORE = 0.8340120342794627

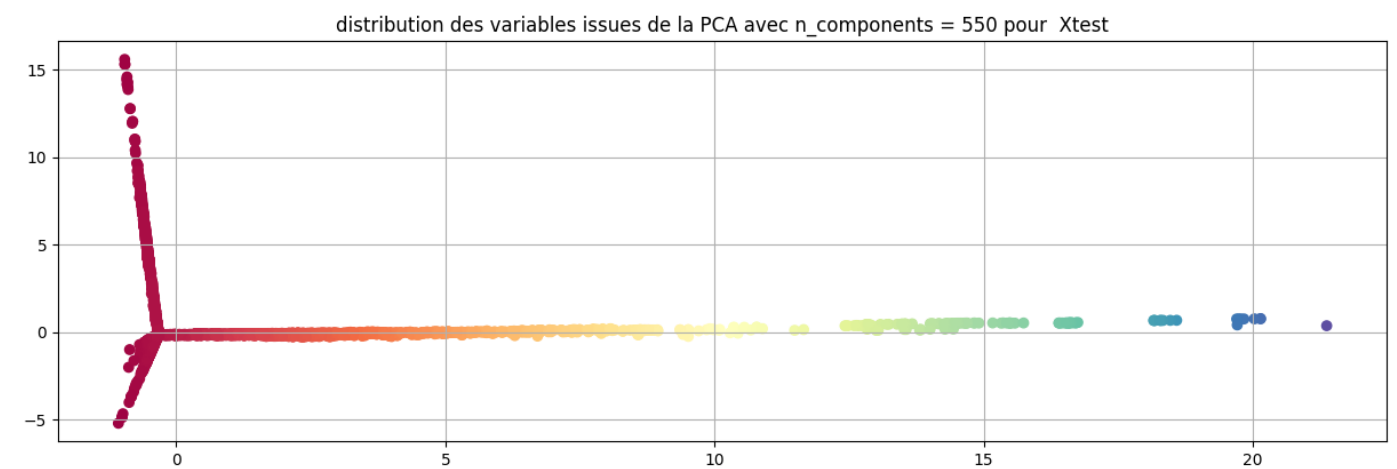
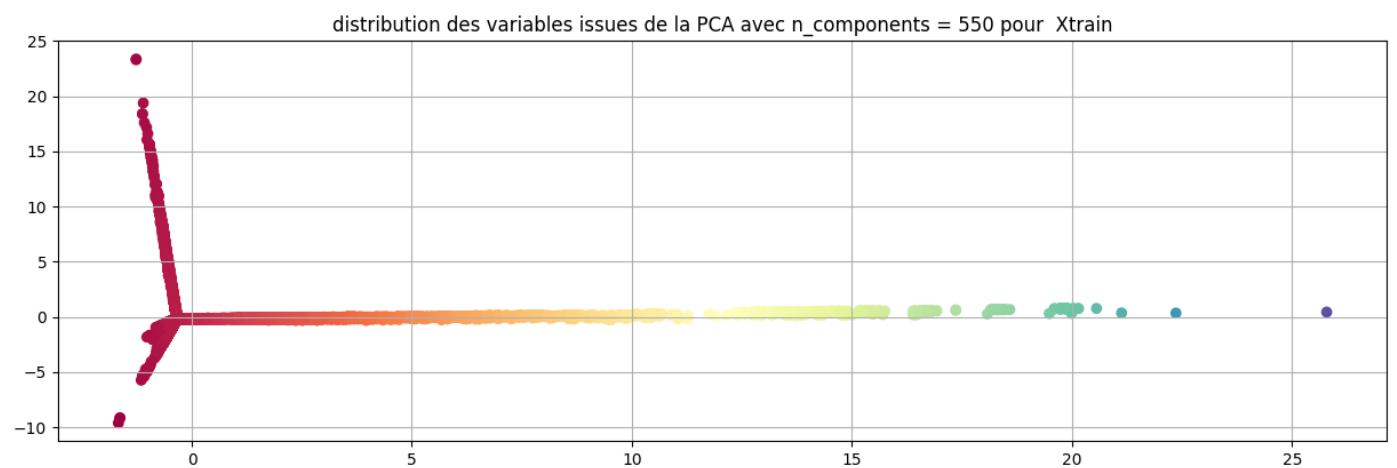
TRAIN\_MSE\_RESULT = 409521.15333069954

TEST\_MSE\_RESULT = 195946.71190664318

BEST\_PARAMS: [{'N\_NEIGHBORS': 10}]



Un minimum de **550** pour le # de composantes après réduction de dimensions PCA donnant un pourcentage de réduction de : **80.0 %**

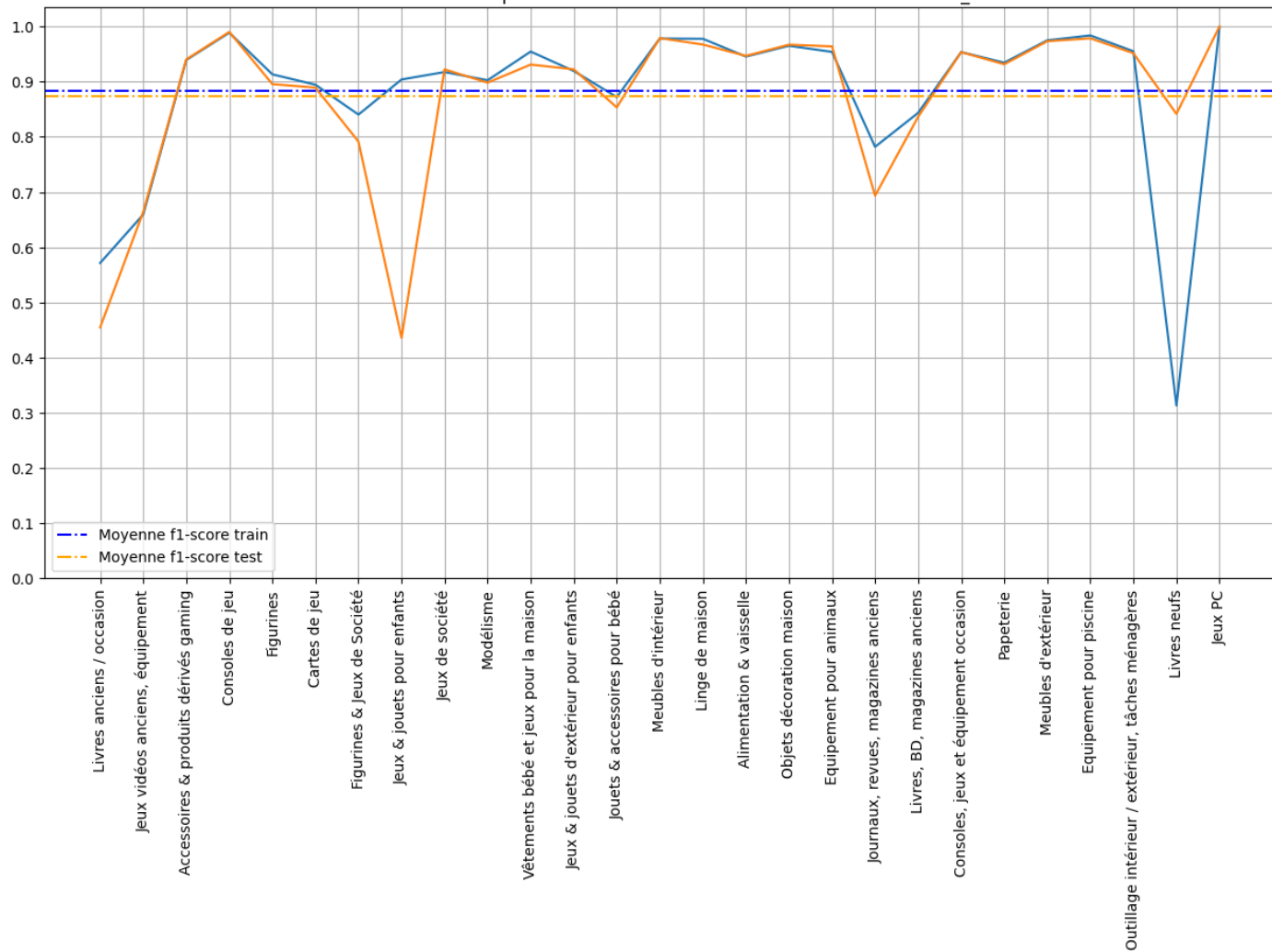


df.shape : (82265, 2702)

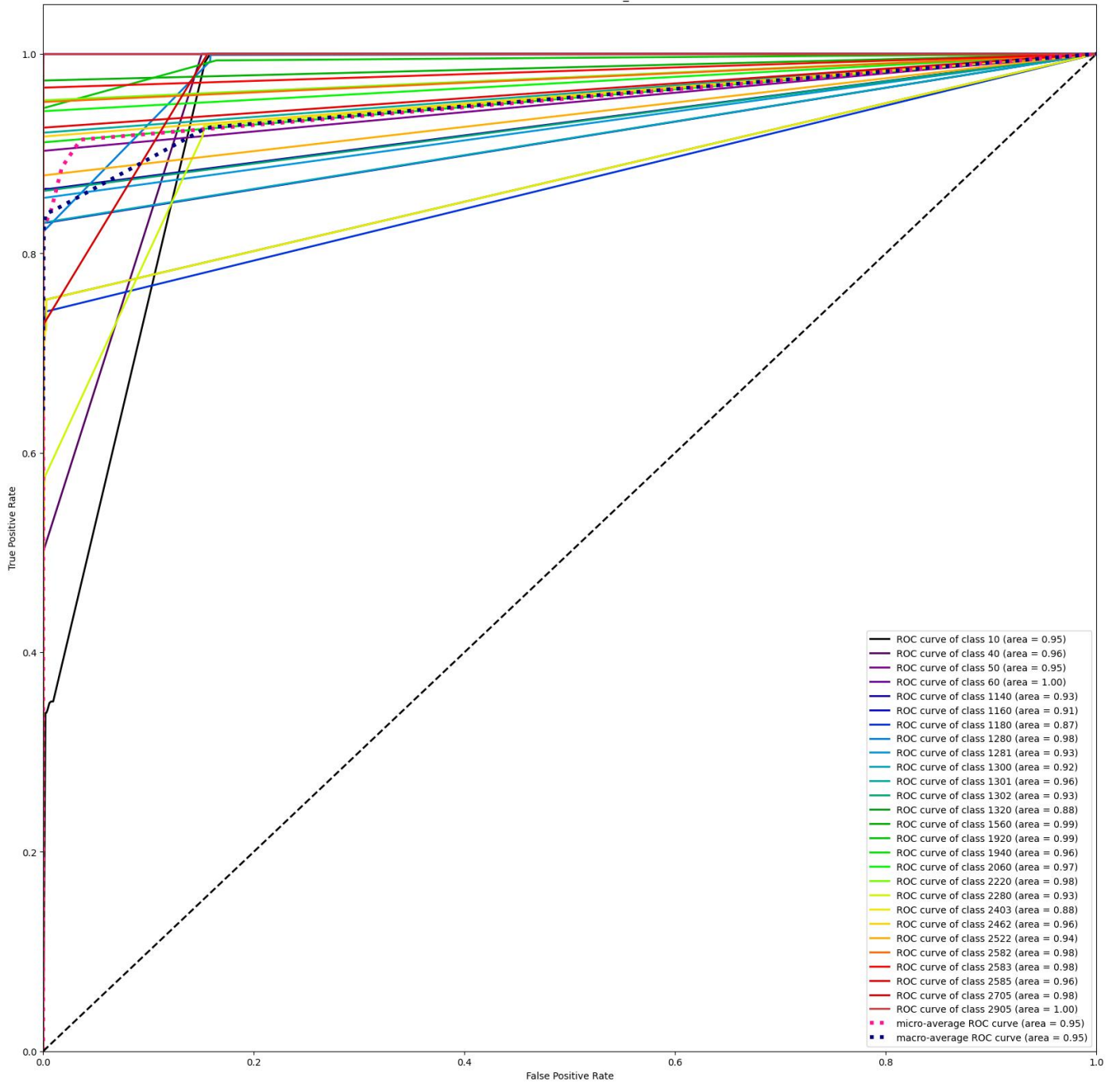




Courbes des f1-score pour les ensembles d'entraînement et de test - KNN\_PCA



Courbes ROC-AUC - KNN\_PCA



## RFC - RANDOMFORESTCLASSIFIER (300 WORDS BY CODE) – THE BEST

```
TRAIN_R2_SCORE = 0.9220203002491947
```

```
TEST_R2_SCORE = 0.9121740715978849
```

```
BEST_PARAMS: [{'MAX_FEATURES': 'SQRT', 'MIN_SAMPLES_SPLIT': 10}]
```

```
X_train.shape - X_test.shape - len(y_train) - len(y_test)
```

```
(65812, 8100) - (16453, 8100) - 65812 - 16453
```

```
estimator      RandomForestClassifier()
```

```
params  {'max_features': ['sqrt'], 'min_samples_split':...
```

Fitting 3 folds for each of 1 candidates, totalling 3 fits

```
train_f1_score = [array([0.49328594, 0.79447115, 0.98415153, 0.99516908, 0.96825397,  
0.95305318, 0.92307692, 0.94928335, 0.95154472, 0.92915893,  
0.9894958 , 0.95852018, 0.93389297, 0.99550302, 0.99721813,  
0.98020586, 0.98858892, 0.98505114, 0.89335485, 0.9163918 ,  
0.98128708, 0.97629708, 0.99320071, 0.99503514, 0.97842105,  
0.87859506, 1.    ])]
```

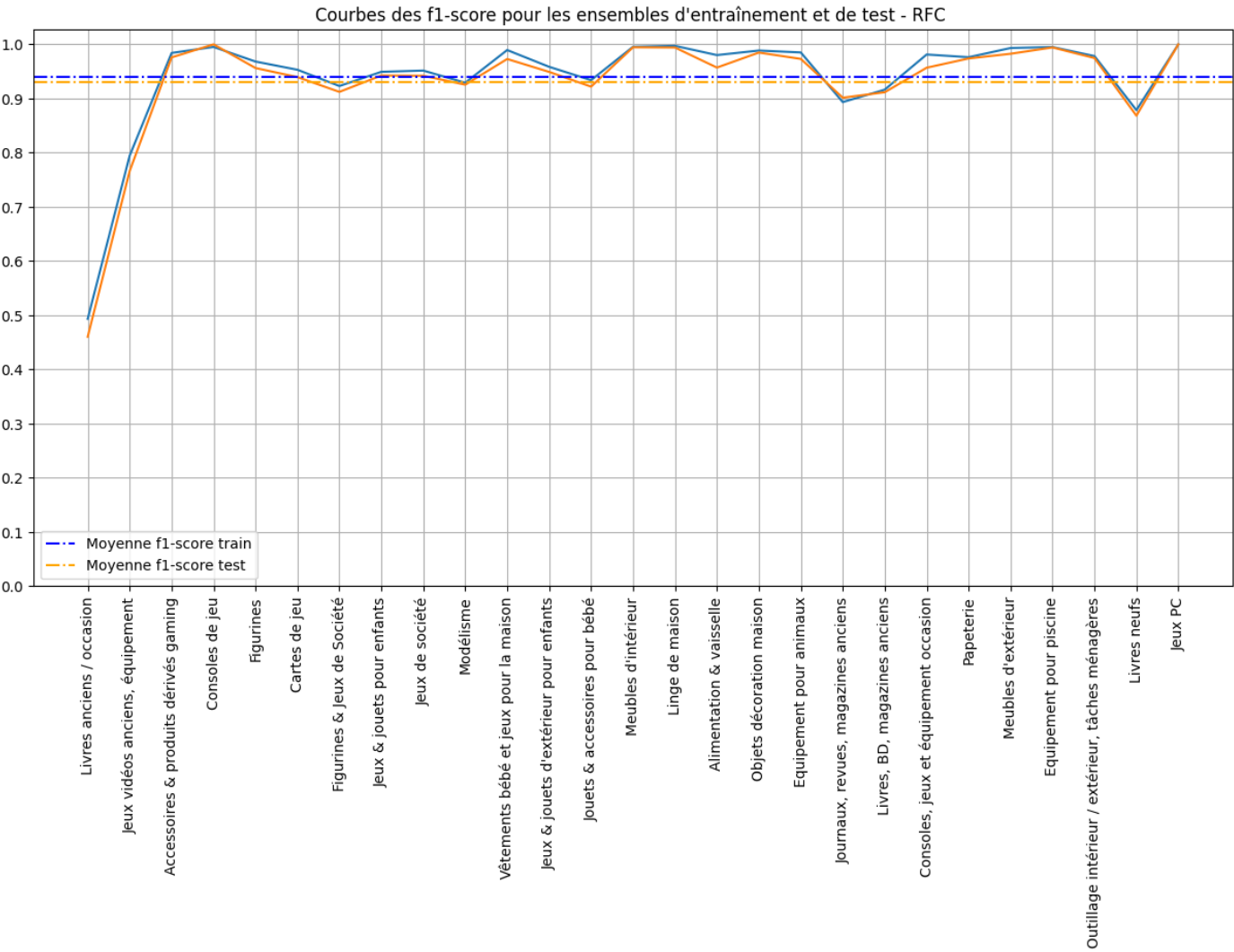
```
test_f1_score = [array([0.46021666, 0.76601307, 0.97592295, 1.    , 0.95626243,  
0.93954135, 0.9122807 , 0.94246575, 0.94200849, 0.9255079 ,  
0.97297297, 0.94911243, 0.92193919, 0.99454094, 0.99413681,  
0.95709571, 0.98487395, 0.97313433, 0.90145577, 0.91160221,  
0.95683453, 0.97393015, 0.98263534, 0.99424987, 0.97473684,  
0.86831276, 1.    ])]
```

```
mean_train_f1_score= 0.9400928726242548
```

```
mean_test_f1_score= 0.9308067817223938
```

	precision	recall	f1-score	support
10	0.30	1.00	0.46	616
40	1.00	0.62	0.77	472
50	1.00	0.95	0.98	319
60	1.00	1.00	1.00	150
1140	1.00	0.92	0.96	525
1160	1.00	0.89	0.94	763
1180	1.00	0.84	0.91	155
1280	1.00	0.89	0.94	965
1281	1.00	0.89	0.94	374

1300	1.00	0.86	0.93	952
1301	1.00	0.95	0.97	114
1302	1.00	0.90	0.95	444
1320	1.00	0.86	0.92	656
1560	1.00	0.99	0.99	1013
1920	1.00	0.99	0.99	772
1940	1.00	0.92	0.96	158
2060	1.00	0.97	0.98	906
2220	1.00	0.95	0.97	172
2280	1.00	0.82	0.90	981
2403	1.00	0.84	0.91	985
2462	1.00	0.92	0.96	290
2522	1.00	0.95	0.97	1043
2582	1.00	0.97	0.98	498

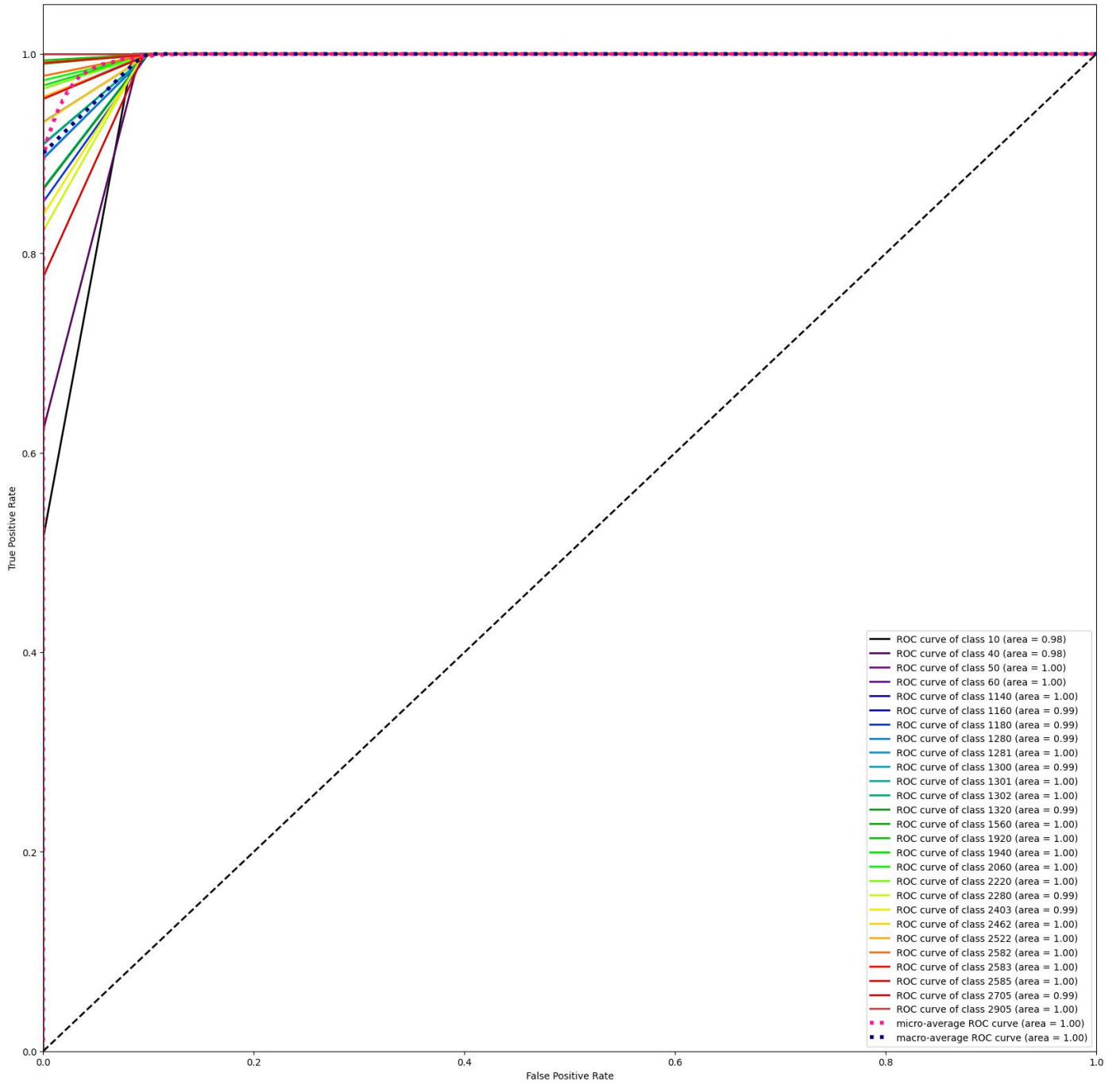


=====**CONFUSION MATRIX**=====

Matrice de confusion-RFC

Valeurs prédites	10	616	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	40	179	293	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	50	15	0	304	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	60	0	0	0	150	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1140	44	0	0	0	481	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1160	87	0	0	0	0	676	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1180	25	0	0	0	0	0	130	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1280	105	0	0	0	0	0	0	860	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1281	41	0	0	0	0	0	0	0	333	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1300	132	0	0	0	0	0	0	0	0	820	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1301	6	0	0	0	0	0	0	0	0	0	108	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1302	43	0	0	0	0	0	0	0	0	0	0	401	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1320	95	0	0	0	0	0	0	0	0	0	0	0	561	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1560	11	0	0	0	0	0	0	0	0	0	0	0	0	1002	0	0	0	0	0	0	0	0	0	0	0	0	
	1920	9	0	0	0	0	0	0	0	0	0	0	0	0	0	763	0	0	0	0	0	0	0	0	0	0	0	
	1940	13	0	0	0	0	0	0	0	0	0	0	0	0	0	0	145	0	0	0	0	0	0	0	0	0	0	
	2060	27	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	879	0	0	0	0	0	0	0	0	0	
	2220	9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	163	0	0	0	0	0	0	0	0	
	2280	176	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	805	0	0	0	0	0	0	0	
	2403	160	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	825	0	0	0	0	0	0	
	2462	24	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	266	0	0	0	0	0	
	2522	53	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	990	0	0	0	0	
	2582	17	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	481	0	0	0	
	2583	22	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1902	0	0	
	2585	24	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	463	0	
	2705	128	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	422	
	2905	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	169	
		10	40	50	60	1140	1160	1180	1280	1281	1300	1301	1302	1320	1560	1920	1940	2060	2220	2280	2403	2462	2522	2582	2583	2585	2705	2905
		Valeurs réelles																										

Courbes ROC-AUC - RFC



## LREG (100 WORDS BY CODE) – 4MIN

BEST\_PARAMS: [{'C': 30}]

TRAIN\_R2\_SCORE = 0.8658603294232055

TEST\_R2\_SCORE = 0.8622135780708685

X\_train.shape - X\_test.shape - len(y\_train) - len(y\_test)

(65812, 2700) - (16453, 2700) - 65812 - 16453

estimator LogisticRegression()

params {'C': [5, 10, 20]}

train\_f1\_score = [array([0.36140046, 0.66355763, 0.94627105, 0.99273608, 0.93363162,  
0.9073154 , 0.89071038, 0.9119452 , 0.91848373, 0.90918919,  
0.9622438 , 0.92756133, 0.87660327, 0.98651802, 0.98189068,  
0.95114007, 0.97431555, 0.96634615, 0.79063803, 0.85167173,  
0.96040987, 0.93652531, 0.98114169, 0.98680361, 0.96119882,  
0.83593131, 1. ])]

test\_f1\_score = [array([0.35135908, 0.66854725, 0.94719472, 0.99665552, 0.92307692,  
0.90294752, 0.89285714, 0.90837104, 0.92063492, 0.91075515,  
0.94444444, 0.92493947, 0.85813751, 0.98750625, 0.97203728,  
0.95016611, 0.97103918, 0.97005988, 0.7997558 , 0.85863268,  
0.94927536, 0.93408278, 0.97636177, 0.98254892, 0.96162047,  
0.83966245, 1. ])]

train\_mse\_result = 456855.5308302437

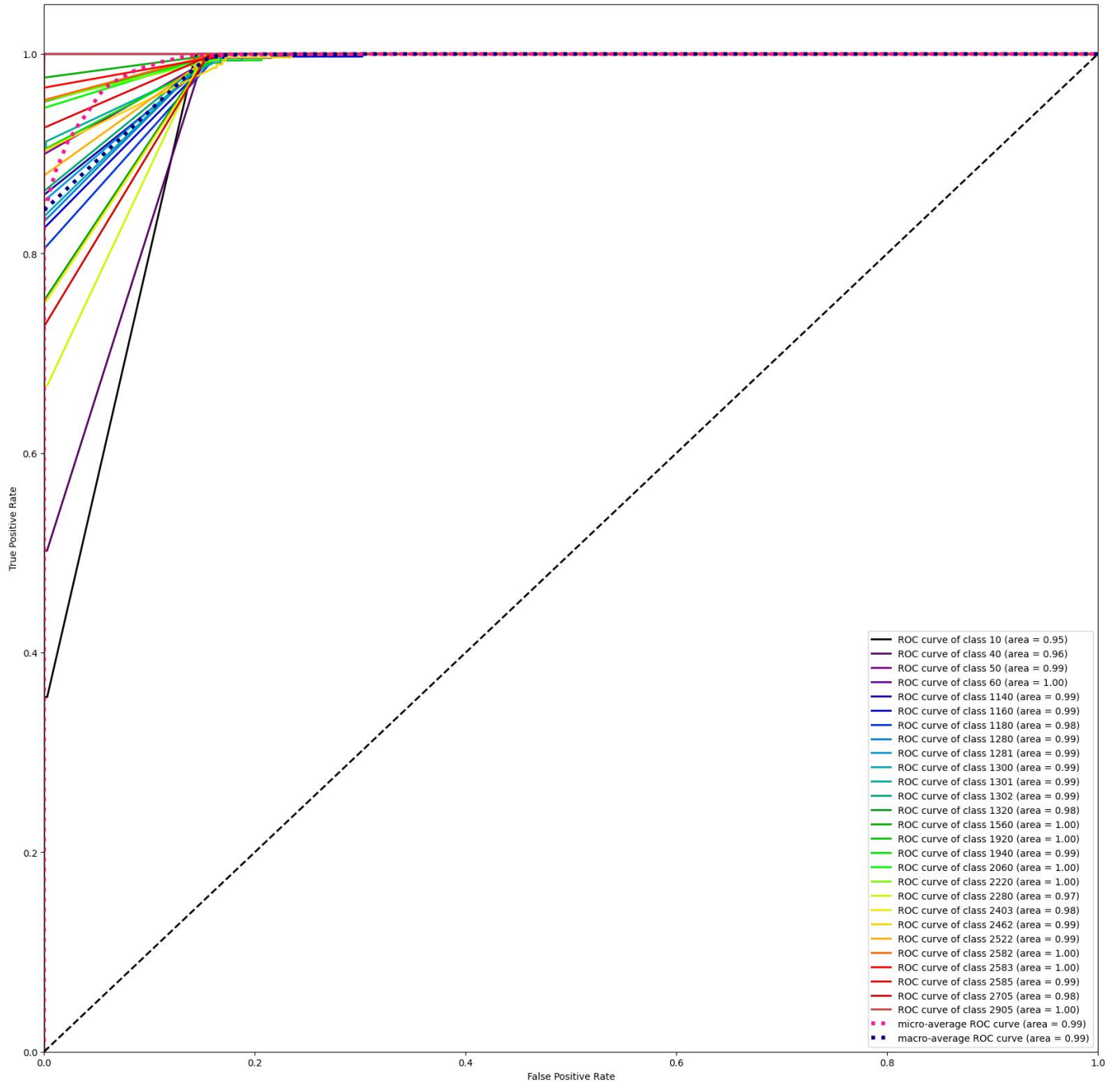
Valeurs prédites

Valeurs réelles

10	614	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	2	0	0	0	0	0	0	0	0	0	
40	235	237	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
50	32	0	287	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
60	1	0	0	149	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1140	75	0	0	0	450	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1160	135	0	0	0	0	628	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1180	30	0	0	0	0	0	125	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1280	162	0	0	0	0	0	0	803	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1281	55	0	0	0	0	0	0	0	319	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1300	156	0	0	0	0	0	0	0	0	796	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1301	12	0	0	0	0	0	0	0	0	0	102	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1302	62	0	0	0	0	0	0	0	0	0	0	382	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1320	163	0	0	0	0	0	0	0	0	0	0	0	493	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1560	25	0	0	0	0	0	0	0	0	0	0	0	0	988	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1920	42	0	0	0	0	0	0	0	0	0	0	0	0	0	730	0	0	0	0	0	0	0	0	0	0	0	0	0	
1940	15	0	0	0	0	0	0	0	0	0	0	0	0	0	0	143	0	0	0	0	0	0	0	0	0	0	0	0	
2060	51	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	855	0	0	0	0	0	0	0	0	0	0	0	
2220	10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	162	0	0	0	0	0	0	0	0	0	0	
2280	326	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	655	0	0	0	0	0	0	0	0	0	
2403	244	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	741	0	0	0	0	0	0	0	0	
2462	28	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	262	0	0	0	0	0	0	0	
2522	129	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	914	0	0	0	0	0	0	
2582	23	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	475	0	0	0	0	0	
2583	66	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1858	0	0	0	0	
2585	36	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	451	0	0	0	
2705	152	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	398	0	0	
2905	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	169	0	
	10	40	50	60	1140	1160	1180	1280	1281	1300	1301	1302	1320	1560	1920	1940	2060	2220	2280	2403	2462	2522	2582	2583	2585	2705	2905		
	Valeurs réelles																												



Courbes ROC-AUC - LREG



## RBF (100 WORDS BY CODE)

TRAIN\_R2\_SCORE = 0.8660274721935209

TEST\_R2\_SCORE = 0.8619704613140461

BEST\_PARAMS: [{'MAX\_FEATURES': 'SQRT', 'MIN\_SAMPLES\_SPLIT': 10}]

X\_train.shape - X\_test.shape - len(y\_train) - len(y\_test)

(65812, 2700) - (16453, 2700) - 65812 - 16453

estimator                RandomForestClassifier()

params    {'name': 'RBF', 'estimator': ensemble.RandomForestClassifier(), 'params': {'max\_features': ["sqrt", None],

          'min\_samples\_split': [1, 10]}

},

          {'name': 'SVC', 'estimator': svm.SVC(),

          'params': {'kernel': ('linear', 'rbf'), 'C': [1, 10]}

          }

train\_f1\_score = [array([0.36168826, 0.66088117, 0.94627105, 0.99273608, 0.93363162,

          0.9073154 , 0.89071038, 0.9119452 , 0.91848373, 0.90918919,

          0.9622438 , 0.92756133, 0.87660327, 0.98651802, 0.98189068,

          0.94857143, 0.97431555, 0.96634615, 0.79063803, 0.85341426,

          0.96040987, 0.93725222, 0.98140127, 0.98680361, 0.96203209,

          0.83623877, 1.     ]])

test\_f1\_score = [array([0.35169854, 0.66288952, 0.94719472, 1.     , 0.92307692,

          0.90373563, 0.89285714, 0.90775325, 0.91907514, 0.90700344,

          0.93457944, 0.92493947, 0.85813751, 0.98801199, 0.97272122,

          0.93602694, 0.97103918, 0.96072508, 0.8014661 , 0.86192952,

          0.95306859, 0.93408278, 0.97636177, 0.98281787, 0.95940171,

          0.84332282, 1.     ]])

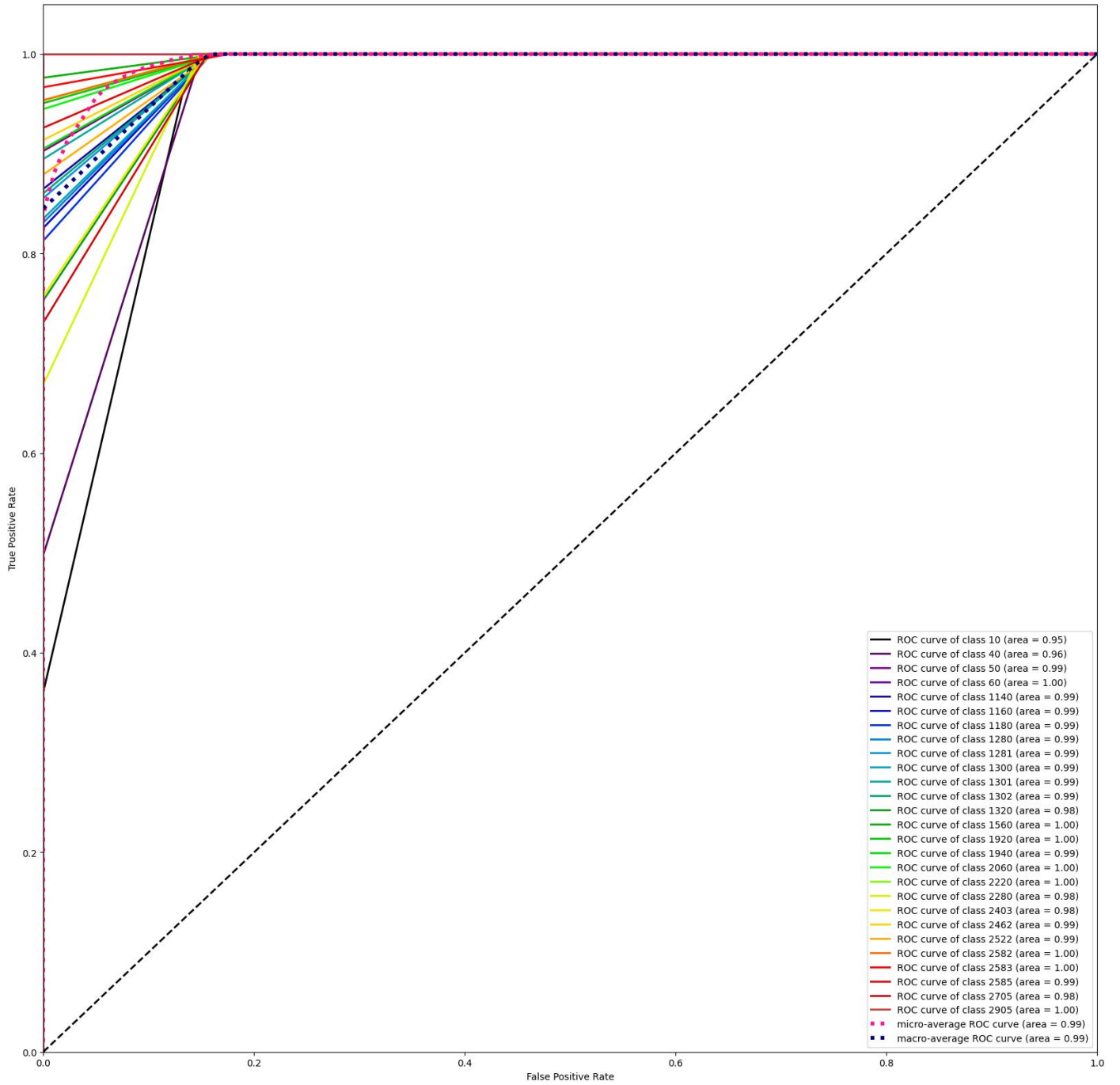
train\_mse\_result = 455162.75148909015

test\_mse\_result = 475895.7078344375

### Matrice de confusion-RBF

[illegible]

Courbes ROC-AUC - RBF



## NAIVE BAYES (100 WORDS BY CODE) – 11SEC TEMPS D'EXECUTION

PARAMS {'ALPHA': [1]}

TRAIN\_R2\_SCORE = 0.8464261836747098

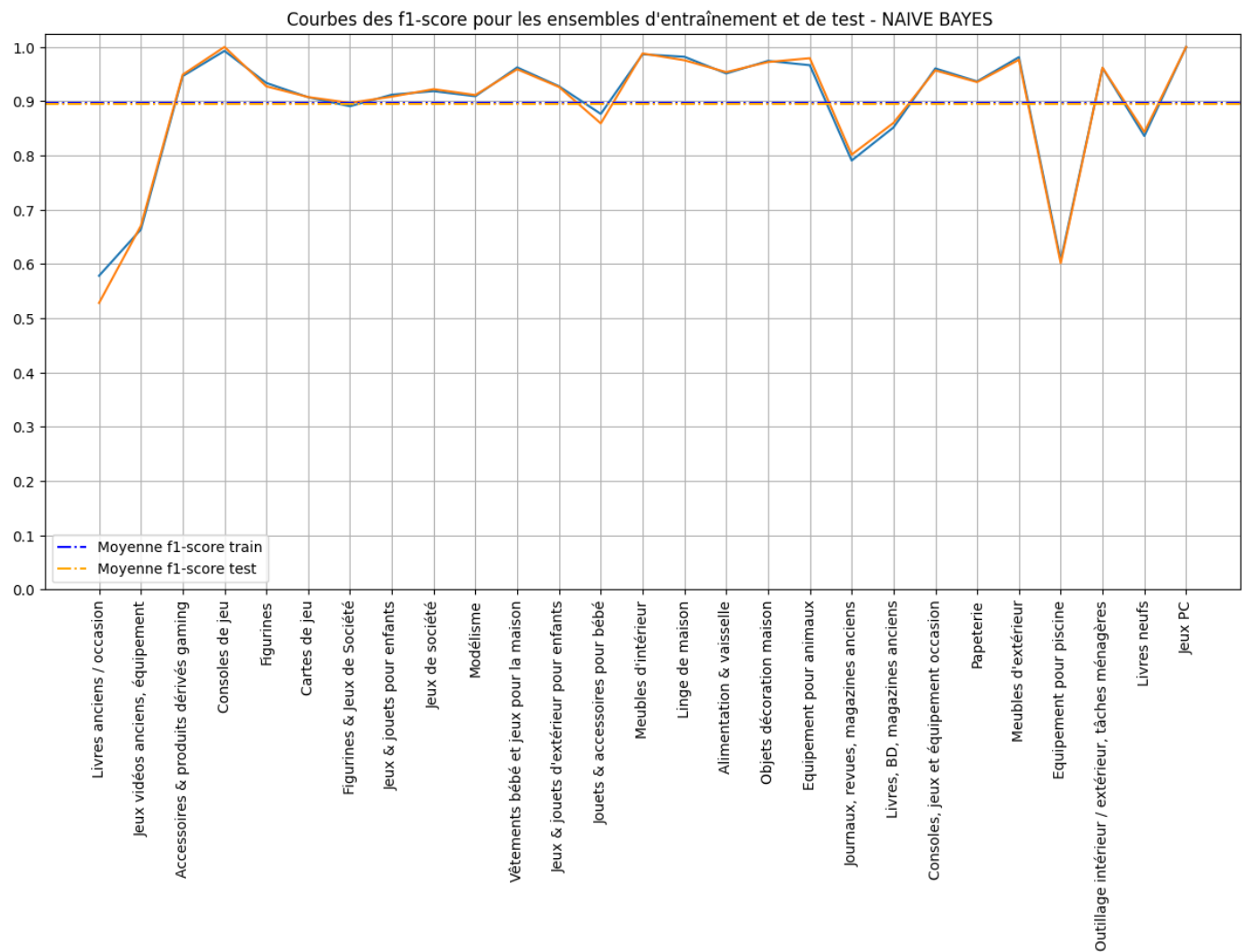
TEST\_R2\_SCORE = 0.8450738467148848

MEAN\_TRAIN\_F1\_SCORE= 0.8964021730543796

MEAN\_TEST\_F1\_SCORE= 0.895199049313926

X\_train.shape - X\_test.shape - len(y\_train) - len(y\_test)

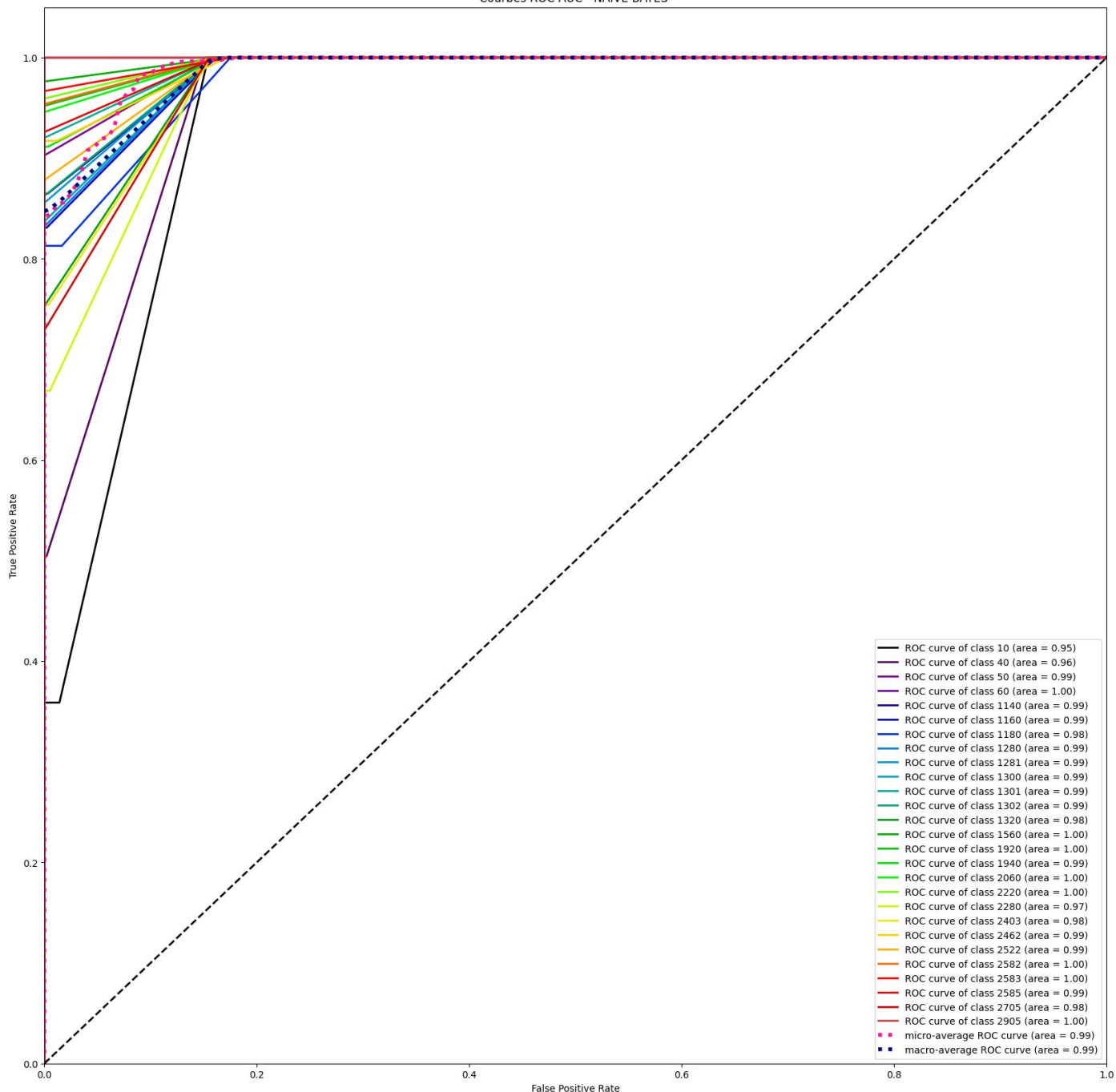
(65812, 2700) - (16453, 2700) - 65812 - 16453



### Matrice de confusion-NAIVE BAYES

10	221	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	395	0	0	0
40	0	238	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	234	0	0	0
50	0	0	288	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	31	0	0	0
60	0	0	0	150	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1140	0	0	0	0	454	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	71	0	0	0
1160	0	0	0	0	0	634	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	129	0	0	0
1180	0	0	0	0	0	0	126	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	29	0	0	0
1280	0	0	0	0	0	0	0	803	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	162	0	0	0
1281	0	0	0	0	0	0	0	0	320	0	0	0	0	0	0	0	0	0	0	0	0	0	0	54	0	0	0
1300	0	0	0	0	0	0	0	0	0	797	0	0	0	0	0	0	0	0	0	0	0	0	0	155	0	0	0
1301	0	0	0	0	0	0	0	0	0	0	105	0	0	0	0	0	0	0	0	0	0	0	0	9	0	0	0
1302	0	0	0	0	0	0	0	0	0	0	0	383	0	0	0	0	0	0	0	0	0	0	0	61	0	0	0
1320	0	0	0	0	0	0	0	0	0	0	0	0	494	0	0	0	0	0	0	0	0	0	0	162	0	0	0
1560	0	0	0	0	0	0	0	0	0	0	0	0	0	989	0	0	0	0	0	0	0	0	0	24	0	0	0
1920	0	0	0	0	0	0	0	0	0	0	0	0	0	0	735	0	0	0	0	0	0	0	0	37	0	0	0
1940	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	144	0	0	0	0	0	0	0	14	0	0	0
2060	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	857	0	0	0	0	0	0	49	0	0	0
2220	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	165	0	0	0	0	0	7	0	0	0
2280	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	656	0	0	0	0	325	0	0	0
2403	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	743	0	0	0	242	0	0	0
2462	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	266	0	0	24	0	0	0
2522	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	916	0	127	0	0	0
2582	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	475	23	0	0	0
2583	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1924	0	0	0
2585	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	36	451	0	0
2705	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	149	0	401	0
2905	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	169

Courbes ROC-AUC - NAIVE BAYES



RF (100 WORDS BY CODE):

FITTING 3 FOLDS FOR EACH OF 1 CANDIDATES, TOTALLING 3 FITS

TRAIN\_R2\_SCORE = 0.8693095484106242

TEST\_R2\_SCORE = 0.8640977329362426

train\_mse\_result = 446904.01537713484

test\_mse\_result = 465541.4230231569

best\_params: [{'max\_features': 'sqrt', 'min\_samples\_split': 100}]

=====CONFUSION MATRIX=====

3. Use SEABORN to draw confusion\_matrix-----

Confusion matrix as graph with Seaborn :

### Matrice de confusion-RF

10	603	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
40	234	267	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
50	30	0	274	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
60	1	0	0	135	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1140	72	0	0	0	436	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1160	128	0	0	0	0	710	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1180	37	0	0	0	0	0	110	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1280	176	0	0	0	0	0	0	757	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1281	58	0	0	0	0	0	0	0	350	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1300	170	0	0	0	0	0	0	0	0	825	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1301	14	0	0	0	0	0	0	0	0	0	119	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1302	68	0	0	0	0	0	0	0	0	0	0	399	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
1320	145	0	0	0	0	0	0	0	0	0	0	0	514	0	0	0	0	0	0	0	0	0	0	0	0	0		
1560	23	0	0	0	0	0	0	0	0	0	0	0	0	925	0	0	0	0	0	0	0	0	0	0	0	0		
1920	40	0	0	0	0	0	0	0	0	0	0	0	0	0	755	0	0	0	0	0	0	0	0	0	0	0		
1940	16	0	0	0	0	0	0	0	0	0	0	0	0	0	0	145	0	0	0	0	0	0	0	0	0	0		
2060	40	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	834	0	0	0	0	0	0	0	0	0		
2220	13	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	160	0	0	0	0	0	0	0	0		
2280	319	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	634	0	0	0	0	0	0	0		
2403	230	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	760	0	0	0	0	0	0		
2462	18	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	247	0	0	0	0	0		
2522	132	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	901	0	0	0	0		
2582	29	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	479	0	0	0		
2583	52	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1847	0	0		
2585	26	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	448	0		
2705	165	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	396		
2905	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	187		
	10	40	50	60	1140	1160	1180	1280	1281	1300	1301	1302	1320	1560	1920	1940	2060	2220	2280	2403	2462	2522	2582	2583	2585	2705	2905	
	Valeurs réelles																											
	Valeurs prédites																											



## SVC (100 WORDS BY CODE)

TRAIN\_R2\_SCORE = 0.8660274721935209

TEST\_R2\_SCORE = 0.8574120221236249

BEST\_PARAMS: [{'C': 10, 'KERNEL': 'LINEAR'}]

params {'kernel': ('linear', 'rbf'), 'C': [10, 20]}

---

X\_train.shape - X\_test.shape - len(y\_train) - len(y\_test)

(65812, 2700) - (16453, 2700) - 65812 - 16453

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train\_f1\_score = [array([0.36168826, 0.66088117, 0.94627105, 0.99273608, 0.93363162,  
0.9073154 , 0.89071038, 0.9119452 , 0.91848373, 0.90918919,  
0.9622438 , 0.92756133, 0.87660327, 0.98651802, 0.98189068,  
0.94857143, 0.97431555, 0.96634615, 0.79063803, 0.85341426,  
0.96040987, 0.93725222, 0.98140127, 0.98680361, 0.96203209,  
0.83623877, 1. ])]

test\_f1\_score = [array([0.34432644, 0.66099291, 0.94719472, 0.99328859, 0.91975309,  
0.90215827, 0.88489209, 0.9052751 , 0.91751085, 0.90574713,  
0.90909091, 0.92363636, 0.85614647, 0.98293173, 0.96722408,  
0.93243243, 0.96928328, 0.94478528, 0.8007335 , 0.86192952,  
0.94545455, 0.93244626, 0.97425335, 0.98201058, 0.95605573,  
0.83474576, 1. ])]

train\_mse\_result = 455162.75148909015

test\_mse\_result = 492912.0065641524

### Matrice de confusion-SVC

[illegible]

## KNN (300 WORD BY CODE)

```
TRAIN_R2_SCORE = 0.9067799185558865
```

```
TEST_R2_SCORE = 0.9002613505135841
```

```
ESTIMATOR KNEIGHBORSCLASSIFIER()
```

```
PARAMS {'N_NEIGHBORS': [10]}
```

```
X_train.shape - X_test.shape - len(y_train) - len(y_test)
```

```
(65812, 8100) - (16453, 8100) - 65812 - 16453
```

Fitting 3 folds for each of 1 candidates, totalling 3 fits

```
train_f1_score = [array([0.75349301, 0.8144208, 0.9837587, 0.98947368, 0.42818645,  
0.96005218, 0.90762332, 0.94754279, 0.95120364, 0.92673847,  
0.97002141, 0.95146727, 0.93545683, 0.99200619, 0.99376026,  
0.94339623, 0.98420685, 0.964687, 0.89900759, 0.92226501,  
0.98637602, 0.97737438, 0.98398983, 0.99484071, 0.96810207,  
0.799908, 0.97447119])]
```

```
test_f1_score = [array([0.74541752, 0.8035488, 0.98245614, 0.97260274, 0.41079812,  
0.9569378, 0.90070922, 0.94072448, 0.95384615, 0.9218573,  
0.95412844, 0.94033413, 0.92193919, 0.98801199, 0.99282453,  
0.88732394, 0.97972973, 0.94153846, 0.91482301, 0.92876563,  
0.97707231, 0.97795198, 0.96465696, 0.99503787, 0.96051227,  
0.77019749, 0.95031056])]
```

```
train_mse_result = 96349.62113292409
```

```
test_mse_result = 103203.23928766791
```

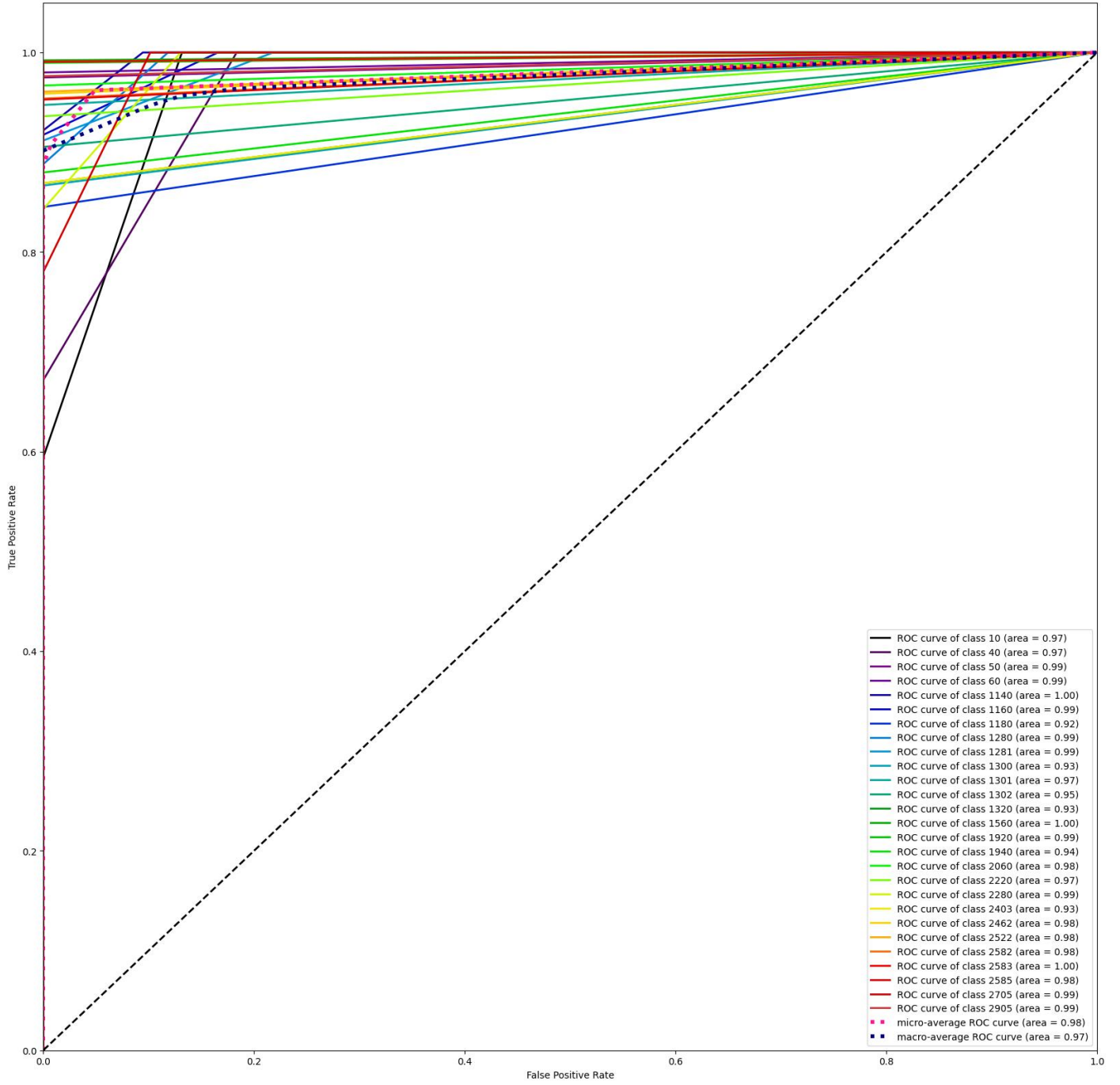
```
best_params: [{'n_neighbors': 10}]
```

=====CONFUSION MATRIX=====

### Matrice de confusion-KNN

[illegible]

Courbes ROC-AUC - KNN



## NAIVE BAYES (300 WORDS BY CODE) – 30SEC TEMPS D'EXECUTION

ESTIMATOR `MULTINOMIALNB()`

PARAMS `{'ALPHA': [1]}`

**TRAIN\_R2\_SCORE = 0.9078283595696833**

**TEST\_R2\_SCORE = 0.9073117364614356**

MEAN\_TRAIN\_F1\_SCORE= 0.9392493667037067

MEAN\_TEST\_F1\_SCORE= 0.9379498799117439

train\_mse\_result = 229948.6231842217

test\_mse\_result = 236261.5843919042

best\_params: `[{'alpha': 1}]`

train\_f1\_score = `[array([0.74661315, 0.79447115, 0.98415153, 0.99435939, 0.96825397,`  
0.95305318, 0.92307692, 0.94928335, 0.95154472, 0.92915893,  
0.9894958 , 0.95852018, 0.93389297, 0.99550302, 0.99721813,  
0.98020586, 0.98858892, 0.98505114, 0.89335485, 0.9162604 ,  
0.98128708, 0.97629708, 0.99320071, 0.71987437, 0.97842105,  
0.87859506, 1. ])]

test\_f1\_score = `[array([0.72066459, 0.77561608, 0.98569157, 1. , 0.96653543,`  
0.95264242, 0.92733564, 0.94593119, 0.9596662 , 0.92853123,  
0.97757848, 0.95652174, 0.92810458, 0.99503968, 0.9974026 ,  
0.98717949, 0.98827471, 0.98823529, 0.90696379, 0.91826659,  
0.97707231, 0.9784525 , 0.98883249, 0.71630678, 0.97796432,  
0.87983707, 1. ])]

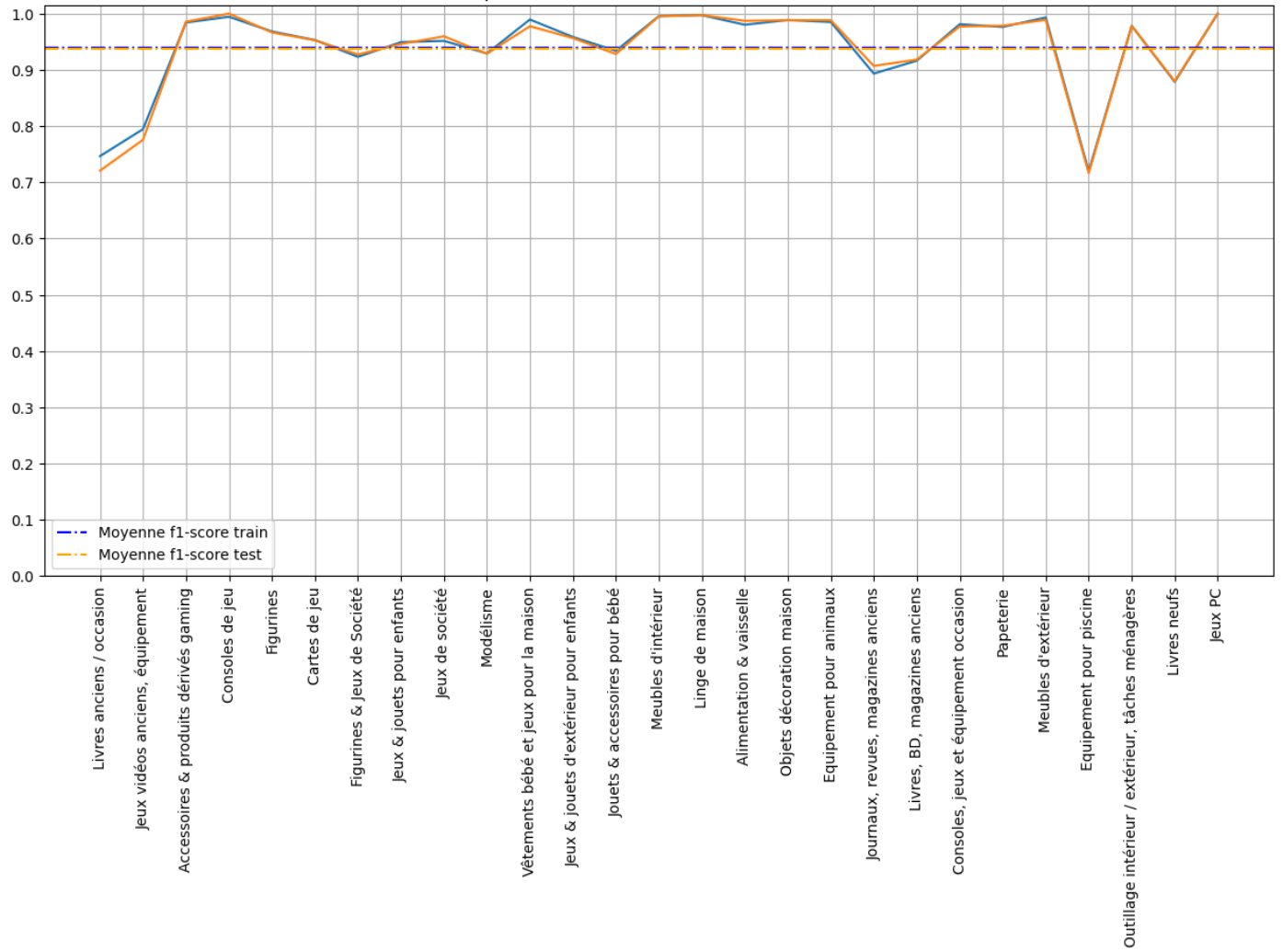
mean\_train\_f1\_score= 0.9392493667037067

mean\_test\_f1\_score= 0.9379498799117439

Matrice de confusion-NAIVE BAYES

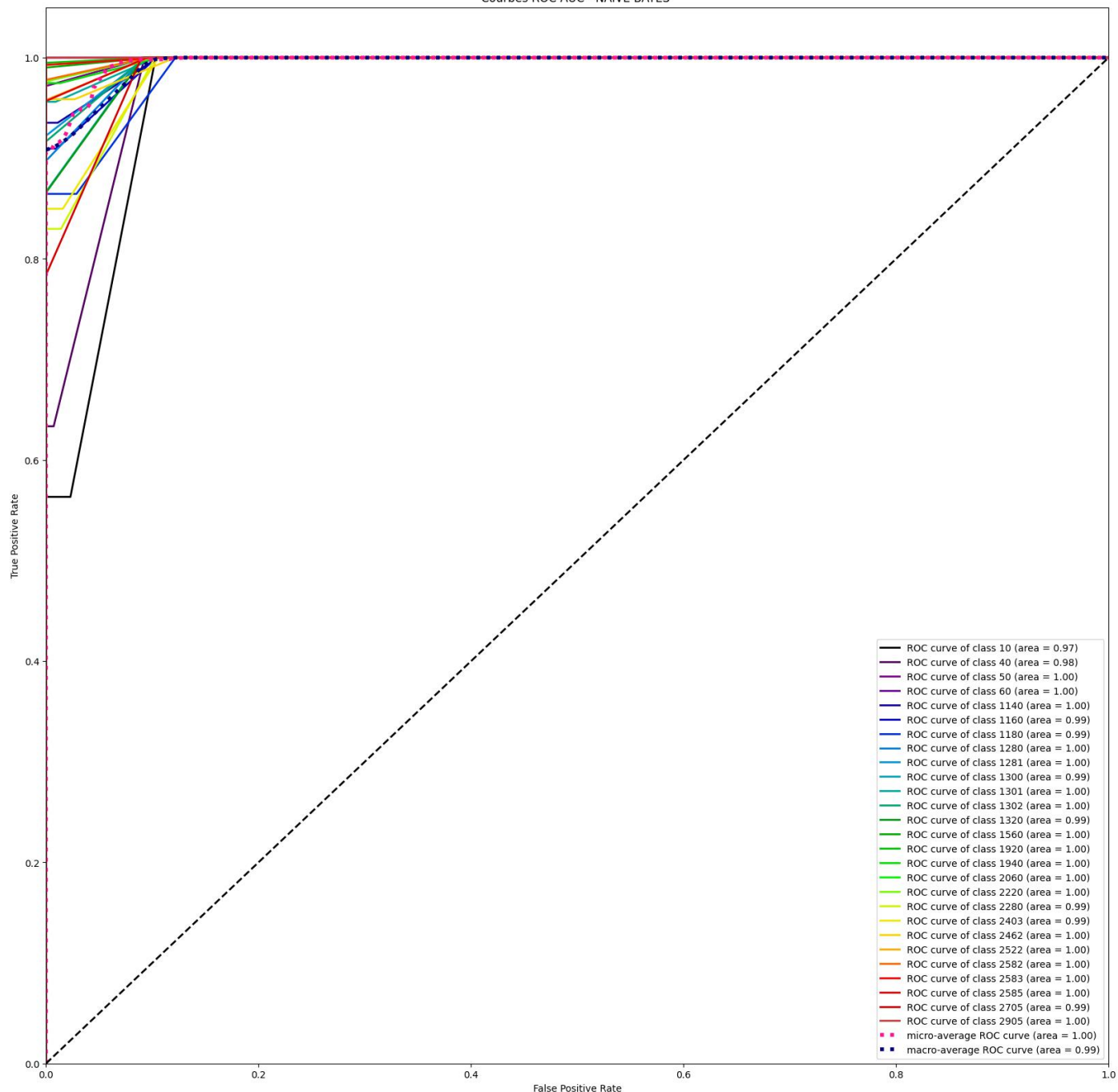
Valeurs prédites	10	347	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	269	0	0	0		
	40	0	299	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	173	0	0	0		
	50	0	0	310	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	9	0	0	0		
	60	0	0	0	150	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
	1140	0	0	0	0	491	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	34	0	0	0		
	1160	0	0	0	0	0	694	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	69	0	0	0		
	1180	0	0	0	0	0	0	134	0	0	0	0	0	0	0	0	0	0	0	0	0	0	21	0	0	0		
	1280	0	0	0	0	0	0	0	866	0	0	0	0	0	0	0	0	0	0	0	0	0	99	0	0	0		
	1281	0	0	0	0	0	0	0	0	345	0	0	0	0	0	0	0	0	0	0	0	0	29	0	0	0		
	1300	0	0	0	0	0	0	0	0	0	825	0	0	0	0	0	0	0	0	0	0	0	127	0	0	0		
	1301	0	0	0	0	0	0	0	0	0	0	109	0	0	0	0	0	0	0	0	0	0	5	0	0	0		
	1302	0	0	0	0	0	0	0	0	0	0	0	407	0	0	0	0	0	0	0	0	0	37	0	0	0		
	1320	0	0	0	0	0	0	0	0	0	0	0	0	568	0	0	0	0	0	0	0	0	88	0	0	0		
	1560	0	0	0	0	0	0	0	0	0	0	0	0	0	1003	0	0	0	0	0	0	0	10	0	0	0		
	1920	0	0	0	0	0	0	0	0	0	0	0	0	0	0	768	0	0	0	0	0	0	4	0	0	0		
	1940	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	154	0	0	0	0	0	4	0	0	0		
	2060	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	885	0	0	0	0	21	0	0	0		
	2220	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	168	0	0	0	4	0	0	0		
	2280	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	814	0	0	167	0	0	0		
	2403	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	837	0	148	0	0	0		
	2462	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	277	0	12	0	0	0		
	2522	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	999	44	0	0	0		
	2582	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	487	11	0	0	0	
	2583	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1924	0	0	0		
	2585	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	21	466	0	0		
	2705	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	118	0	432	0		
	2905	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	169		
Valeurs réelles		10	40	50	60	1140	1160	1180	1280	1281	1300	1301	1302	1320	1560	1920	1940	2060	2220	2280	2403	2462	2522	2582	2583	2585	2705	2905

Courbes des f1-score pour les ensembles d'entraînement et de test - NAIVE BAYES





Courbes ROC-AUC - NAIVE BAYES



## LREG (300 WORDS BY CODE)

ESTIMATOR LOGISTICREGRESSION()

PARAMS {'C': [50]}

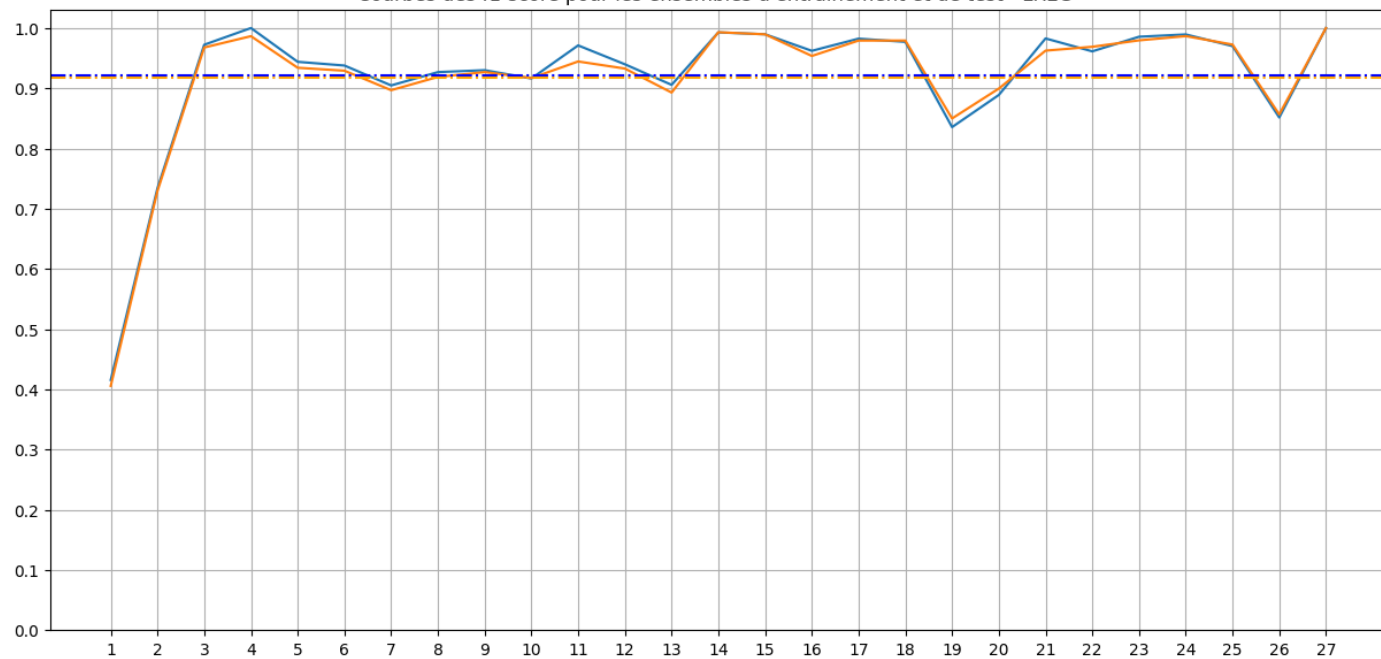
TRAIN\_R2\_SCORE = 0.8932109645657327

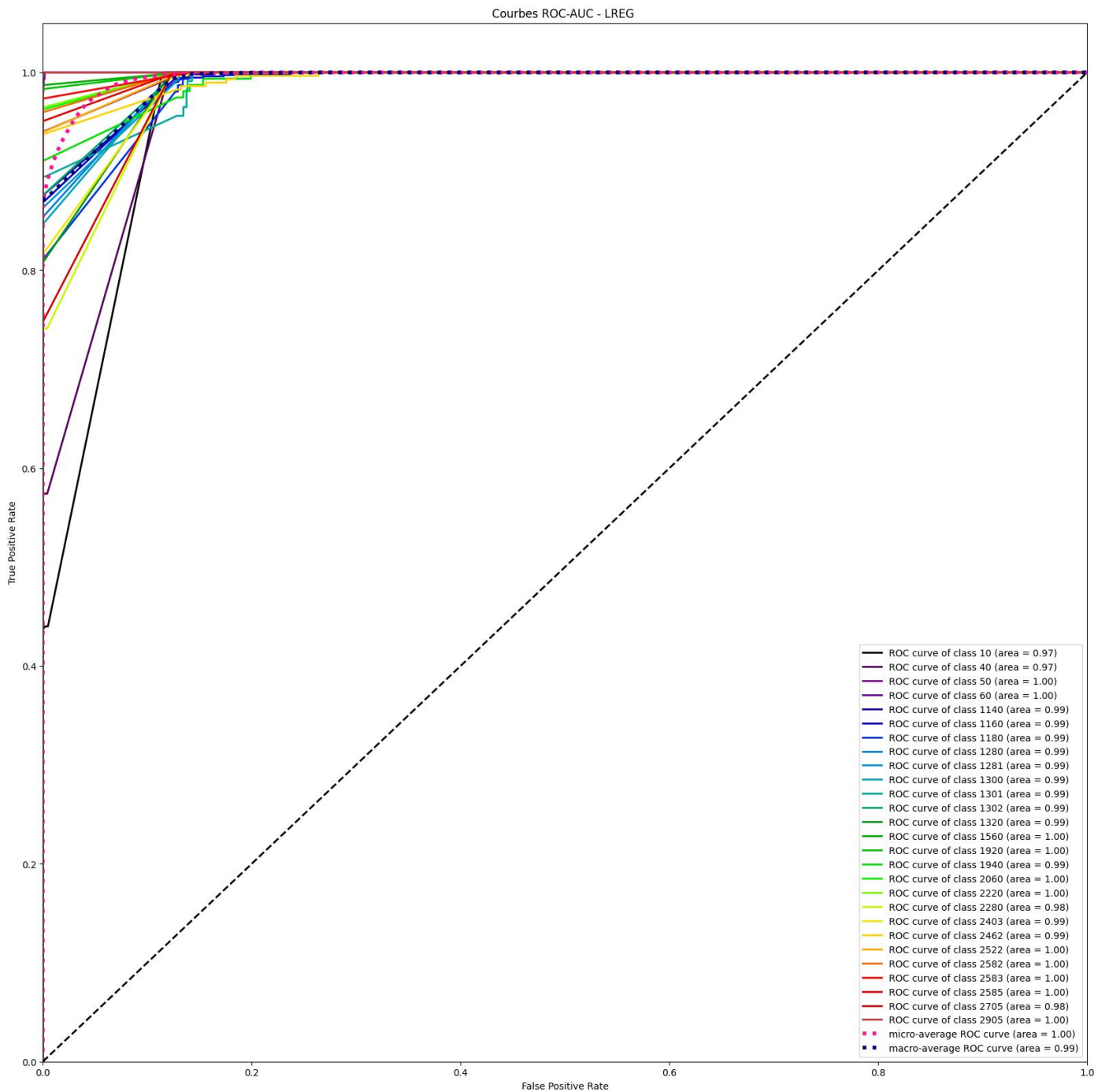
TEST\_R2\_SCORE = 0.8905974594298912

### Matrice de confusion-LREG

[illegible]

Courbes des f1-score pour les ensembles d'entraînement et de test - LREG





```
{'mean_fit_time': array([108.86818051]),
'std_fit_time': array([3.84594668]),
'mean_score_time': array([0.80805755]),
'std_score_time': array([0.09732477]),
'param_C': masked_array(data=[50],
mask=[False],
fill_value='?',
dtype=object),
'params': [{'C': 50}],
'split0_test_score': array([0.88809372]),
'split1_test_score': array([0.88872681]),
```

```
'split2_test_score': array([0.88412272]),  
'mean_test_score': array([0.88698108]),  
'std_test_score': array([0.00203763]),  
'rank_test_score': array([1])}
```