



ONCFM

# **Algorithme de détection automatique de faux billets**

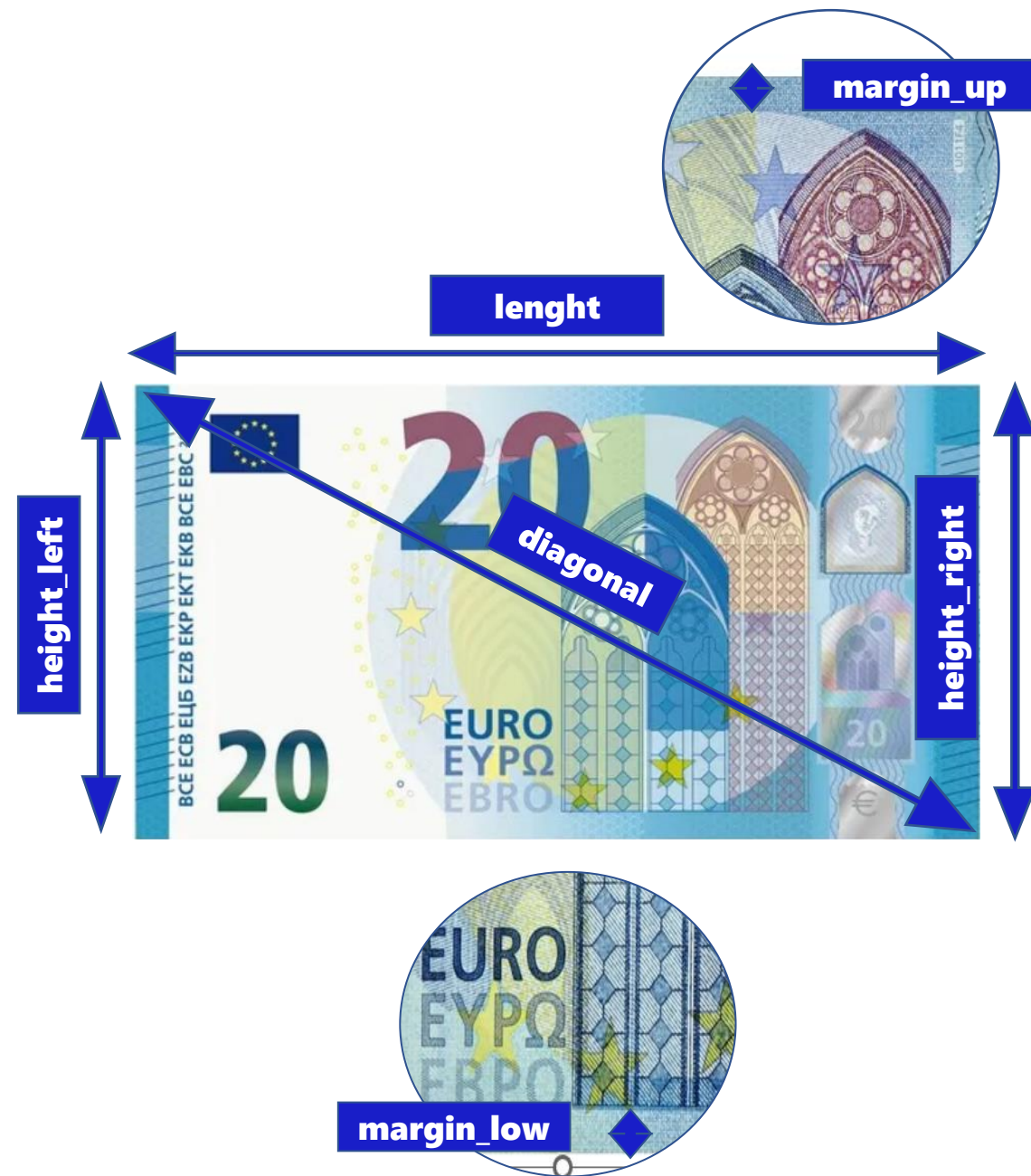
# **Analyse descriptive des données**

## Authenticité du billet

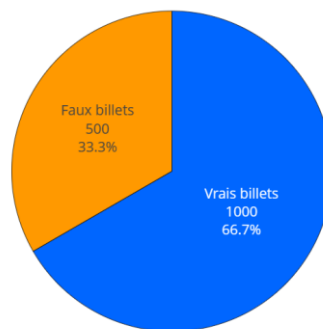
## Dimensions du billet

	is_genuine	diagonal	height_left	height_right	margin_low	margin_up	length
0	True	171.81	104.86	104.95	4.52	2.89	112.83
1	True	171.46	103.36	103.66	3.77	2.99	113.09
2	True	172.69	104.48	103.50	4.40	2.94	113.16
3	True	171.36	103.91	103.94	3.62	3.01	113.51
4	True	171.73	104.28	103.46	4.04	3.48	112.54
...	...	...	...	...	...	...	...
1495	False	171.75	104.38	104.17	4.42	3.09	111.28
1496	False	172.19	104.63	104.44	5.27	3.37	110.97
1497	False	171.80	104.01	104.12	5.51	3.36	111.95
1498	False	172.06	104.28	104.06	5.17	3.46	112.25
1499	False	171.47	104.15	103.82	4.63	3.37	112.07

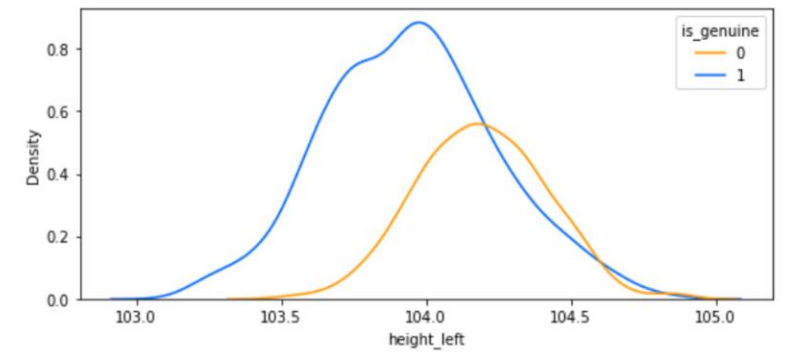
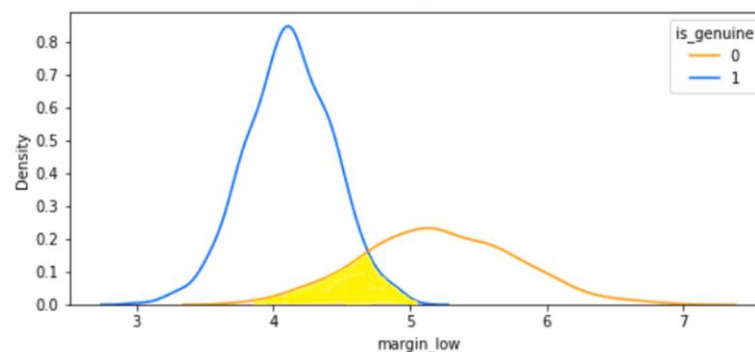
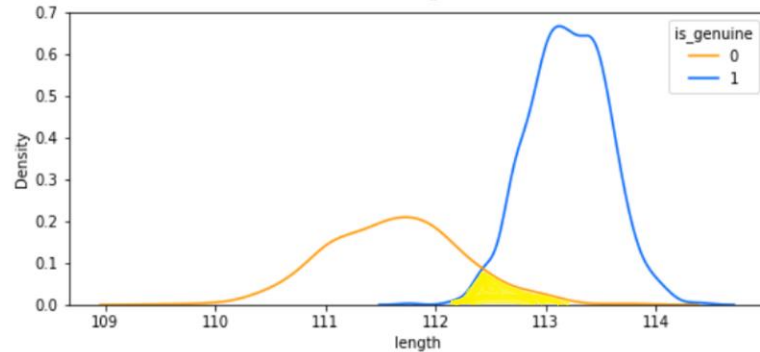
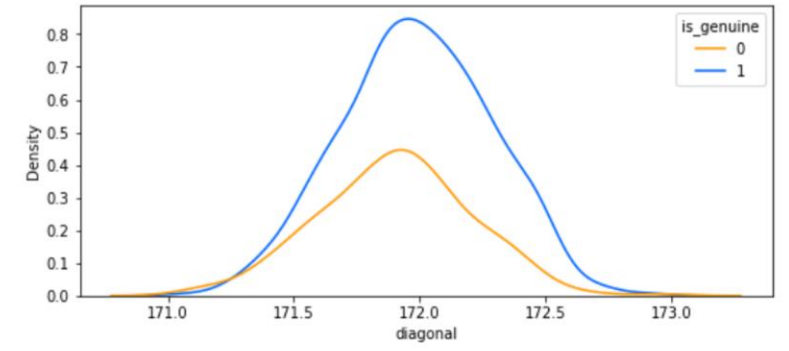
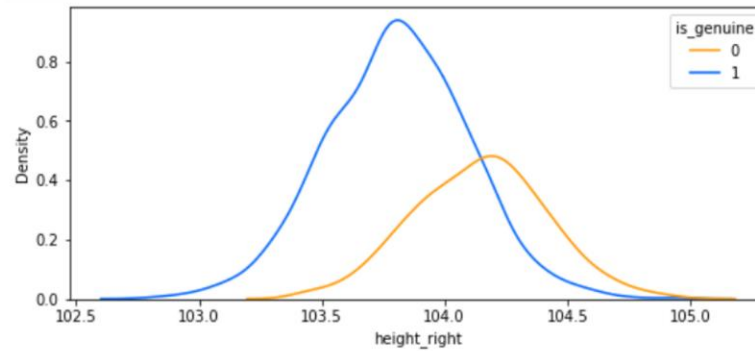
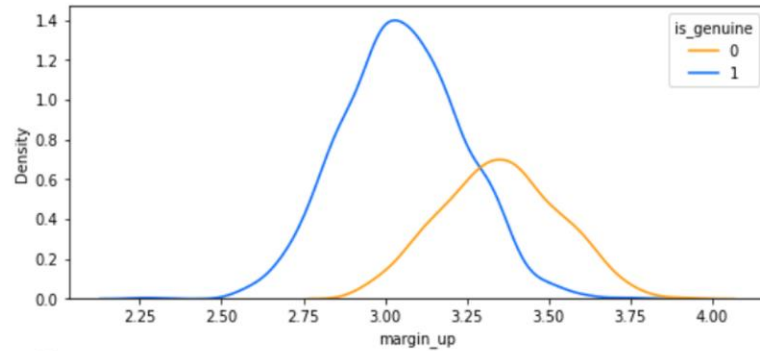
1500 rows × 7 columns



- 1500 billets dont :
  - 1000 billets authentiques
  - 500 faux billets
- Aucune dimension en double
- 37 marges inférieures manquantes

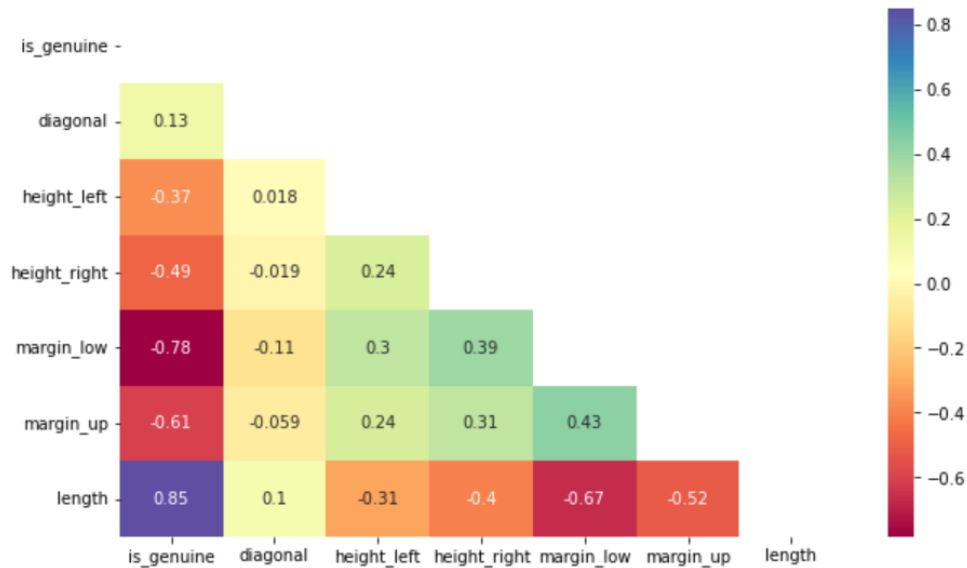


# Dimensions / Authenticité



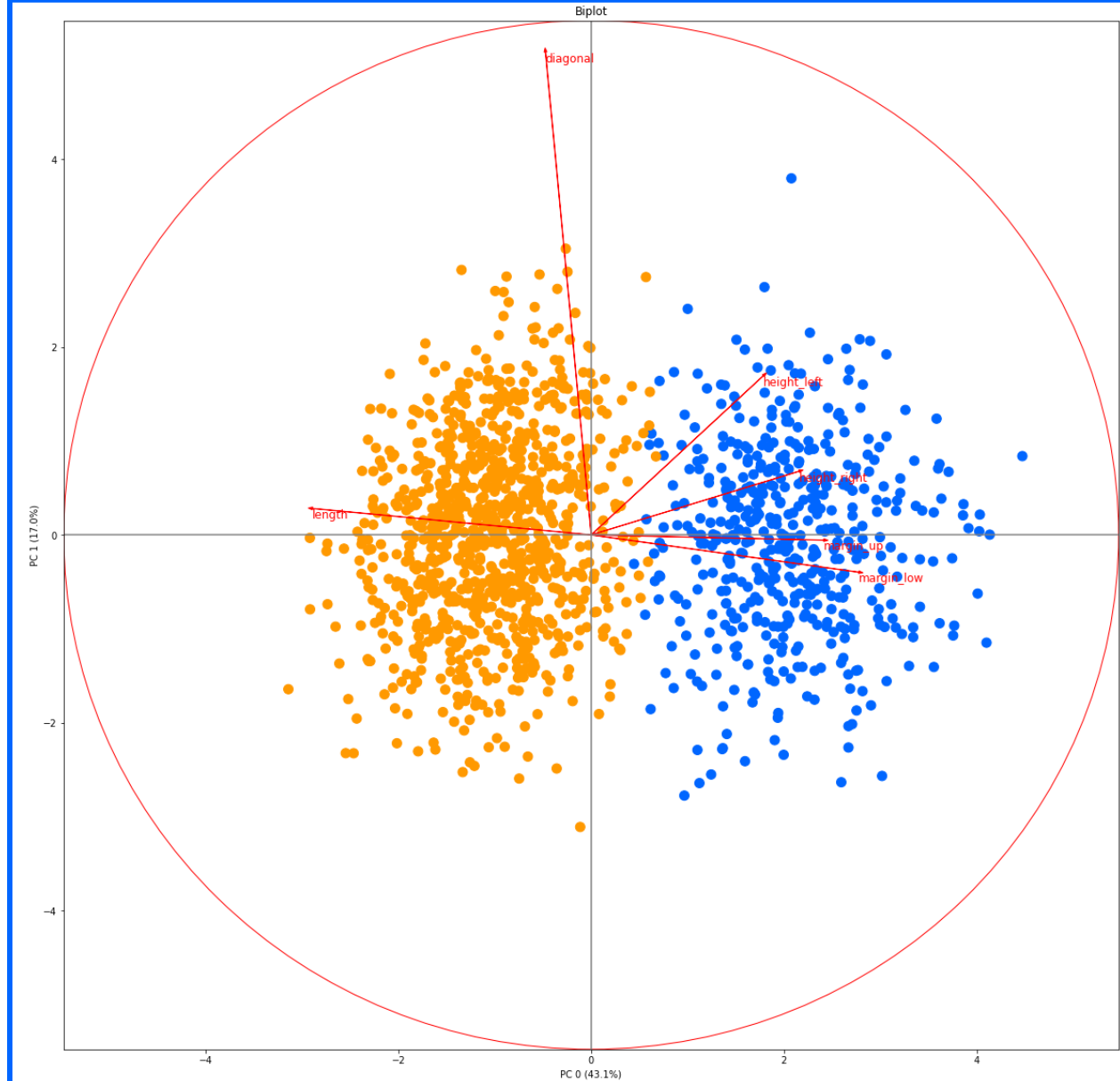
- Les billets authentiques sont plut longs que les faux.
- La marge entre le bord inférieur du billet et l'image de celui-ci (en mm) est plus grande dans les faux billets.

## Matrice des corrélations



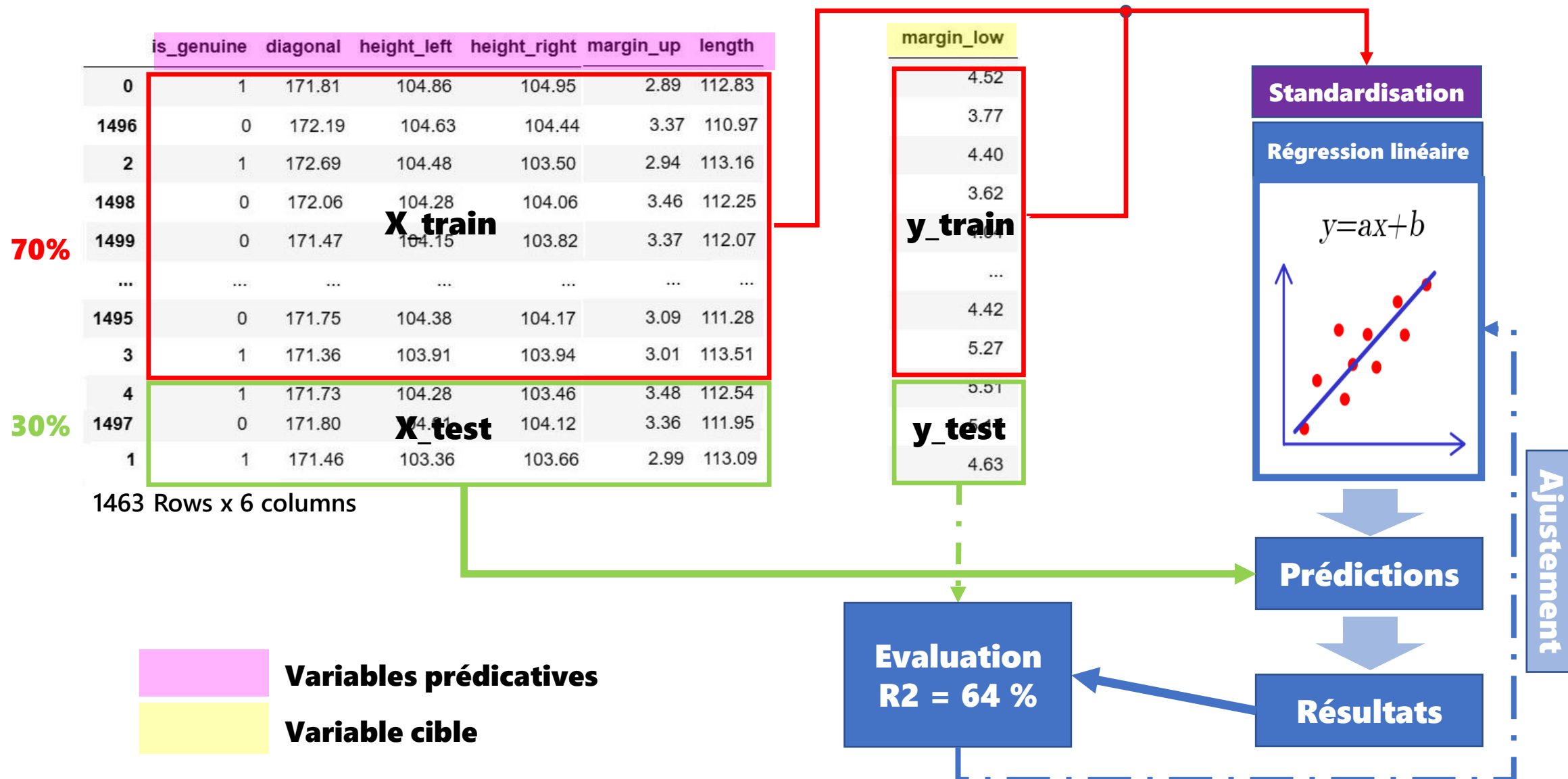
- La longueur et la marge basse sont respectivement corrélées positivement et négativement à l'authenticité du billet.
- La longueur est une dimension distinctive pour l'authenticité d'un billet.

## ACP – KMeans



# **Traitement des valeurs manquantes**

# Entraînement

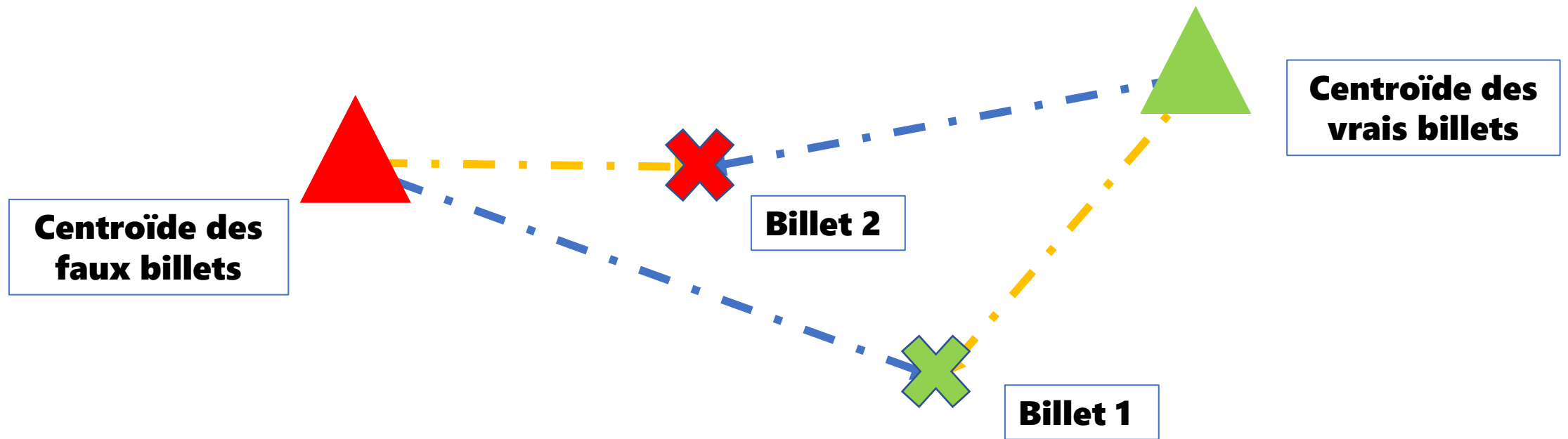


# **Algorithme de classification KMeans**



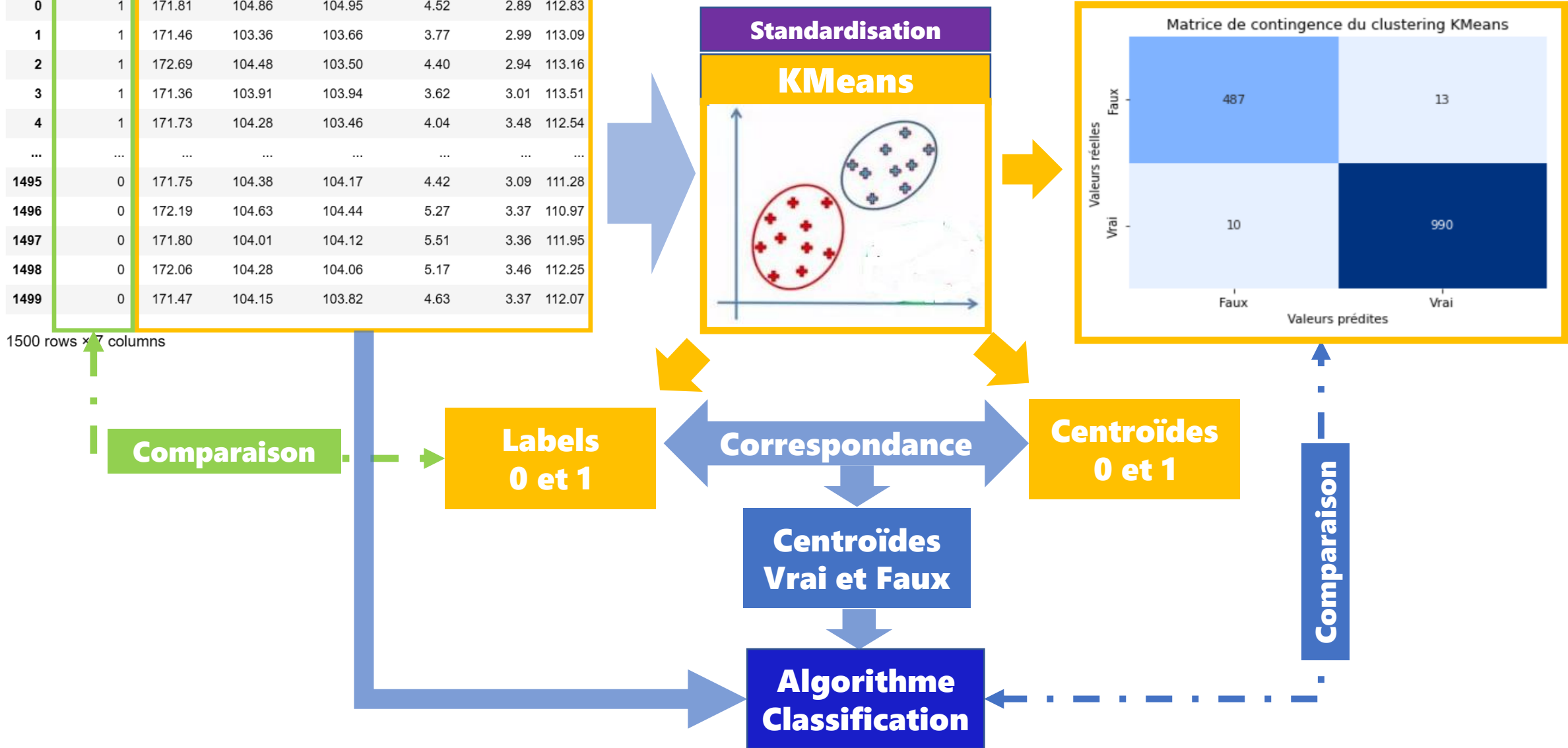
# Algorithme Classification KMeans

↔ **Courte distance**



	is_genuine	diagonal	height_left	height_right	margin_low	margin_up	length
0	1	171.81	104.86	104.95	4.52	2.89	112.83
1	1	171.46	103.36	103.66	3.77	2.99	113.09
2	1	172.69	104.48	103.50	4.40	2.94	113.16
3	1	171.36	103.91	103.94	3.62	3.01	113.51
4	1	171.73	104.28	103.46	4.04	3.48	112.54
...	...	...	...	...	...	...	...
1495	0	171.75	104.38	104.17	4.42	3.09	111.28
1496	0	172.19	104.63	104.44	5.27	3.37	110.97
1497	0	171.80	104.01	104.12	5.51	3.36	111.95
1498	0	172.06	104.28	104.06	5.17	3.46	112.25
1499	0	171.47	104.15	103.82	4.63	3.37	112.07

1500 rows x 7 columns



# **Algorithme de classification**

## **Régression logistique**

## Entraînement

70%

**X\_train**

**y\_train**

30%

**X\_test**

**y\_test**

	diagonal	height_left	height_right	margin_low	margin_up	length
0	171.81	104.86	104.95	4.52	2.89	112.83
1	171.46	103.36	103.66	3.77	2.99	113.09
2	172.69	104.48	103.50	4.40	2.94	113.16
3	171.36	103.91	103.94	3.62	3.01	113.51
4	171.73	104.28	103.46	4.04	3.48	112.54
...	...	...	...	...	...	...
95	171.75	104.38	104.17	4.42	3.09	111.28
1496	172.19	104.63	104.44	5.27	3.37	110.97
497	171.80	104.01	104.12	5.51	3.36	111.95
1498	172.06	104.28	104.06	5.17	3.46	112.25
1499	171.47	104.15	103.82	4.63	3.37	112.07

1500 rows × 7 columns

is\_genuine

1

1

1

1

1

...

0

1

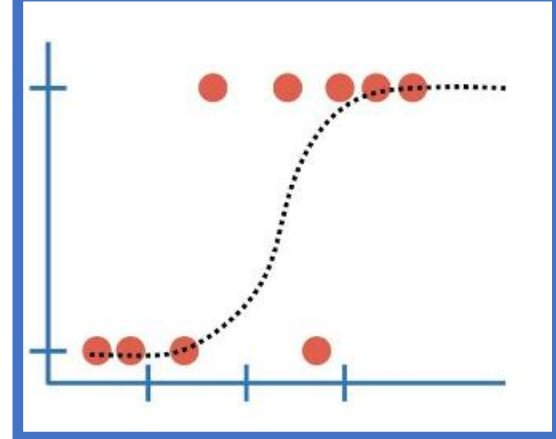
1

0

0

**Standardisation**

**Régression logistique**



**Prédictions**

**Résultats**

**Evaluation**

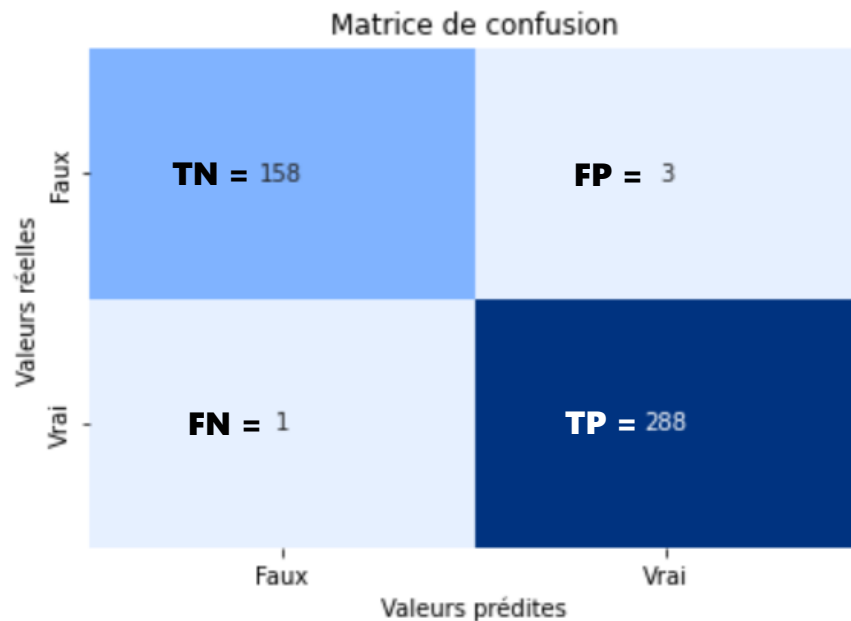
**Variables prédictives**

**Variable cible**

Ajustement / Amélioration

# Evaluation du modèle

## Matrice de confusion



## Rapport de classification

	precision	recall	f1-score	support
Faux billets	0.9937	0.9814	0.9875	161
Vrais billets	0.9897	0.9965	0.9931	289
accuracy			0.9911	450
macro avg	0.9917	0.9890	0.9903	450
weighted avg	0.9911	0.9911	0.9911	450

## Erreurs de classification :

- Trois faux billets classifiés comme étant authentiques (Faux Positifs).
- Un vrai billet classifié comme étant faux (Faux Négatifs).

## Amélioration du modèle



## Baisse des faux positifs



$$\uparrow \text{Precision} = \frac{TP}{TP + FP} \downarrow$$

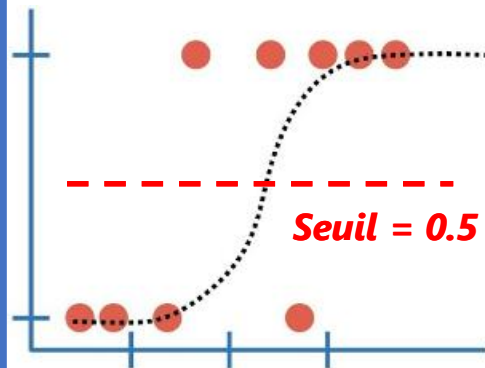
# Amélioration du modèle

## Changement du seuil

**GridSearchCV**  
**Optimisation des**  
**Hyperparamètres**

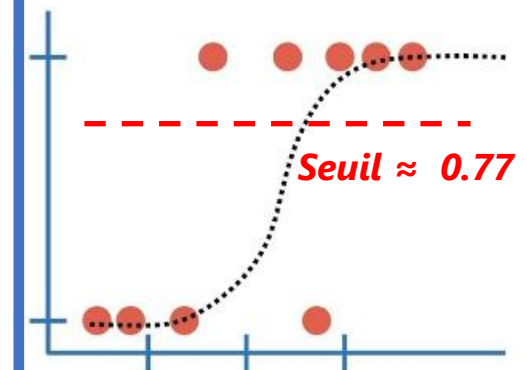
**Standardisation**

**Régression logistique**



**Standardisation**

**Régression logistique**  
**paramétrée**



## Matrice de confusion après amélioration

Valeurs réelles	Faux	Vrai
	Faux	Vrai
Faux	161	0
Vrai	4	285

## Rapport de classification

	precision	recall	f1-score	support
Faux billets	0.9758	1.0000	0.9877	161
Vrais billets	1.0000	0.9862	0.9930	289
accuracy			0.9911	450
macro avg	0.9879	0.9931	0.9904	450
weighted avg	0.9913	0.9911	0.9911	450

## Matrice de confusion initiale

Valeurs réelles	Faux	Vrai
	Faux	Vrai
Faux	158	3
Vrai	1	288

## Rapport de classification

	precision	recall	f1-score	support
Faux billets	0.9937	0.9814	0.9875	161
Vrais billets	0.9897	0.9965	0.9931	289
accuracy			0.9911	450
macro avg	0.9917	0.9890	0.9903	450
weighted avg	0.9911	0.9911	0.9911	450

## Test des algorithmes

	diagonal	height_left	height_right	margin_low	margin_up	length	id
0	171.76	104.01	103.54	5.21	3.30	111.42	A_1
1	171.87	104.17	104.13	6.00	3.31	112.09	A_2
2	172.00	104.58	104.29	4.99	3.39	111.57	A_3
3	172.49	104.55	104.34	4.44	3.03	113.20	A_4
4	171.65	103.63	103.56	3.77	3.16	113.33	A_5

**Classification  
KMeans**

**Classification  
Logistique**

id	diagonal	height_left	height_right	margin_low	margin_up	length	authenticité
A_1	171.76	104.01	103.54	5.21	3.30	111.42	Faux
A_2	171.87	104.17	104.13	6.00	3.31	112.09	Faux
A_3	172.00	104.58	104.29	4.99	3.39	111.57	Faux
A_4	172.49	104.55	104.34	4.44	3.03	113.20	Vrai
A_5	171.65	103.63	103.56	3.77	3.16	113.33	Vrai

## ACP – KMeans

