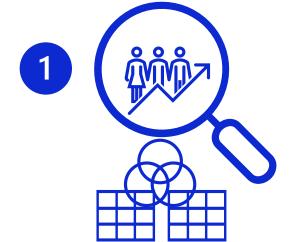
# Implémentez un modèle de scoring





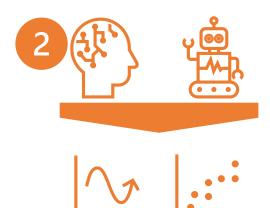








Construction DataFrame
Acquisition données
Nettoyage &
Feature Engineering
Analyse des distributions



## Dashboard & déploiement

Interprétation globale / locale Serveurs Flask / Streamlit Hébergement Cloud AWS Limites & préconisations

## Entrainement / Sélection modèle

Validation croisée + SMOTE

Fonction coût métier

Optimisation du seuil

Tests des modèles et sélection

Chargement données:

8 fichiers .csv

3 clés:



SK\_ID\_CURR (n° demande)

SK\_ID\_PREV

SK ID BUREAU



## Kernel Kaggle Dataset cible: https://www.kaggle.com/code/jsaguiar/lightgbm-with-simple-features/script

application\_{train|test}.csv

bureau.csv

bureau\_balance.csv

POS\_CASH\_balance.csv

credit\_card\_balance.csv

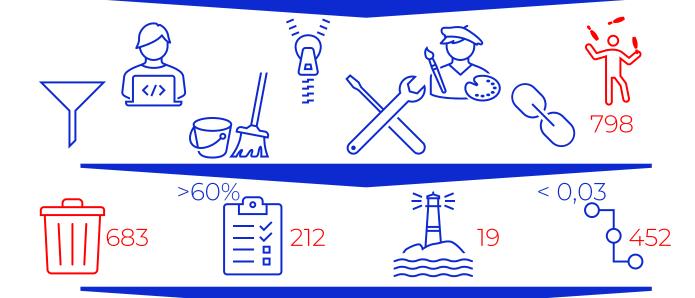
previous\_application.csv

installments\_payments.csv

HomeCredit\_columns\_description.csv

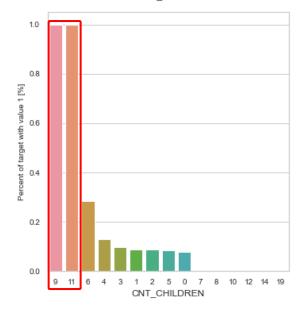
## Construction DataFrame

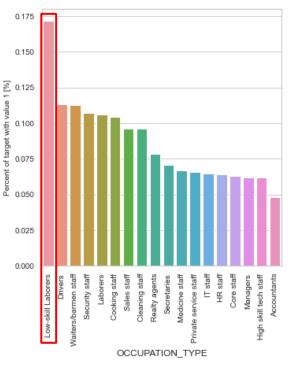
	Rows	Columns	%NaN	%Duplicate	Object _dtype	Float _dtype	Int _dtype	Bool _dtype	MB _Memory
./data\application_test.csv	48744	121	23.81	0.0	16	65	40	0	44.998
./data\application_train.csv	307511	122	24.40	0.0	16	65	41	0	286.227
./data\bureau.csv	1716428	17	13.50	0.0	3	8	6	0	222.620
./data\bureau_balance.csv	27299925	3	0.00	0.0	1	0	2	0	624.846
./data\credit_card_balance.csv	3840312	23	6.65	0.0	1	15	7	0	673.883
./data\HomeCredit_columns_description.csv	219	4	15.18	0.0	4	0	0	0	0.008
./data\installments_payments.csv	13605401	8	0.01	0.0	0	5	3	0	830.408
./data\POS_CASH_balance.csv	10001358	8	0.07	0.0	1	2	5	0	610.435
./data\previous_application.csv	1670214	37	17.98	0.0	16	15	6	0	471.481
./data\sample_submission.csv	48744	2	0.00	0.0	0	1	1	0	0.744

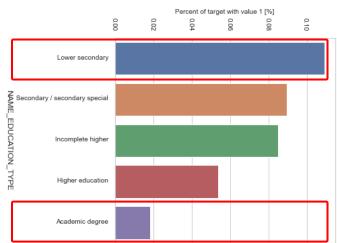




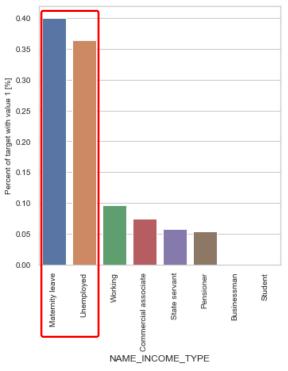
# 0.10 0.08 0.00 0.00 M F CODE\_GENDER XNA

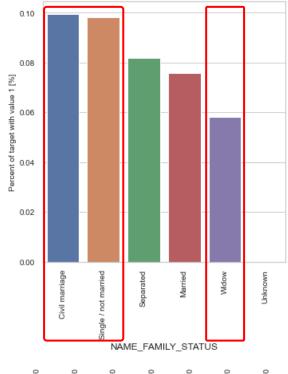


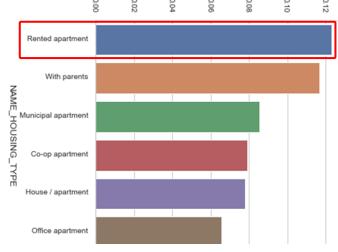


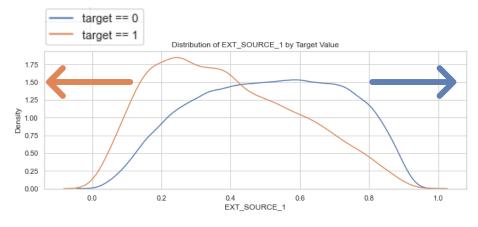


## Analyse des distributions

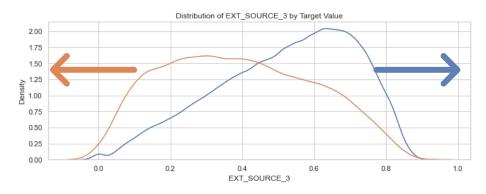






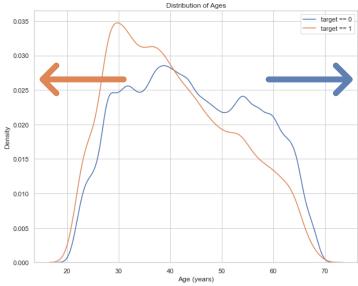






## Analyse des distributions





## Entrainement / Sélection modèle

## Validation croisée + SMOTE

### **Outliers**

Echantillonnage Stratifié

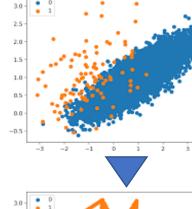
**GridSearchCV** 

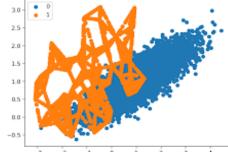
remplacer outliers par des Nan

respecter proportions« TARGET »

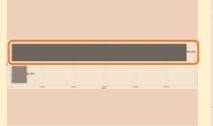


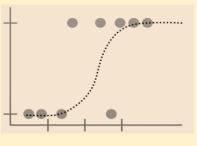
- Modèle : Pipeline [Imputer (moyenne)
  - +Smote + Scaler + Classifier]
- Scoring : Custom
   Score, ROC\_AUC,
   PR\_AUC, Time
- Splits : 3 KFolds

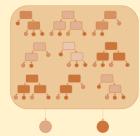




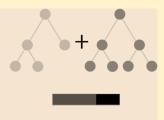
## 6 modèles testés:







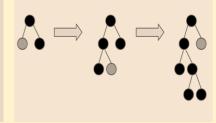




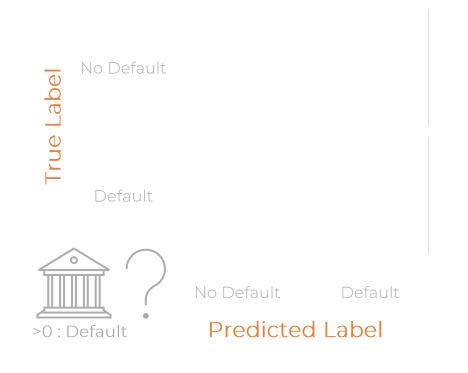


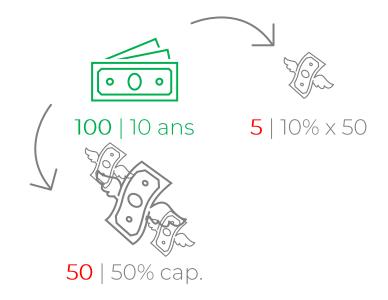
000000000

000000000



## Entrainement / Sélection modèle Fonction coût métier





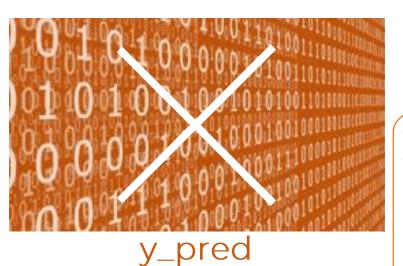


Minimiser fonction:

(entre 0 et 1)

## 2 Entrainement / Sélection modèle

## Optimisation du seuil



0,719	0,757	0,431	0,711
0,366	0,666	0,879	0,861
0,241	0,293	0,455	0,286
0,853	0,531	0,079	0,284
0,846	0,357	0,433	0,493
0,114	0,321	0,015	0,990
0,567	0,261	0,088	0,084

y\_pred\_proba

- best thresh = 0
- best score = 1

Step 1

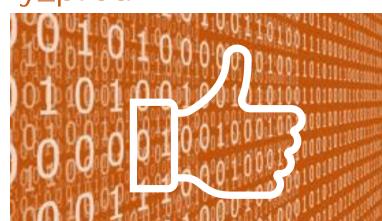
Loop : thresh € (0, 1, 100 pas)

- score = metric(y train, y\_pred\_proba > thresh)
- si score < best score
  - best thresh = thresh
  - best score = score

- best thresh = 0,101010
- best score = 0,534
- y\_pred = y\_pred\_proba > best\_thresh

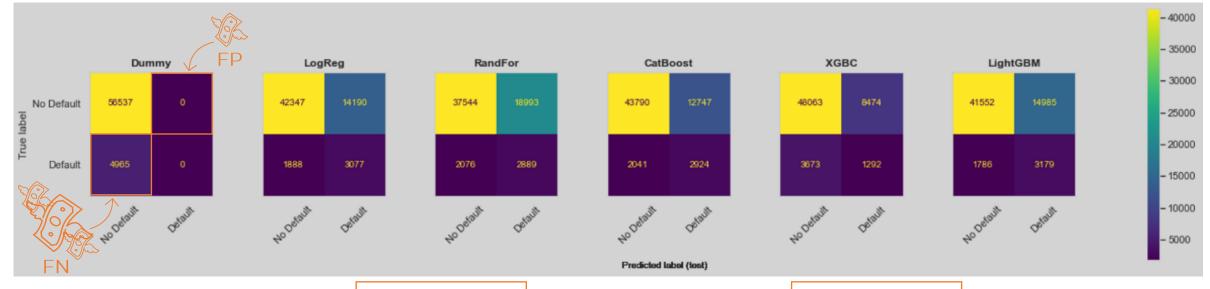
Final Step

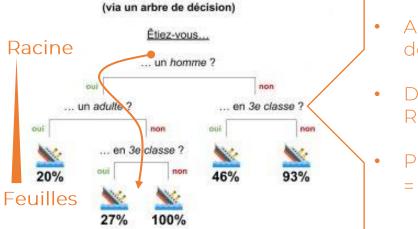
y\_pred





## 2 Entrainement / Sélection modèle Sélection des modèles



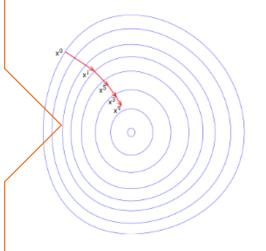


Quelles étaient vos chances de survie sur le Titanic ?

- Arbres de décision
- Départ = Racine
- Probabilités = Feuilles

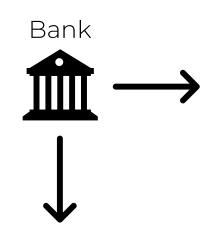
Model	Custom _score	Time
LightGBM	0.534	1752.54
LogReg	0.5377	382.99
CatBoost	0.5391	415.61
RandFor	0.6464	408.85
XGBC	0.735	6966.31
Dummy	0.8073	0.009

- Gradient Boosting Machine
- plusieurs apprenants « faibles »
- larbre = descente de gradient



## 3 Dashboard & déploiement Interprétation globale / locale

http://15.188.141.153:8501/

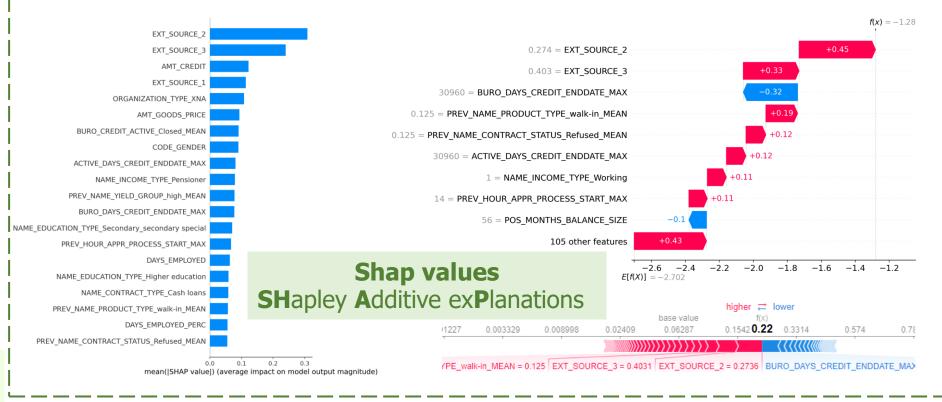


### Data Customer 345565

EXT SOURCE 2 0.2736 Val / Var Mean AMT CREDIT 619k Val / Var Mean ORGANIZATION TYPE XNA Val / Var Mean BURO\_CREDIT\_ACTIVE\_Closed\_MEAN 0.556 Val / Var Mean ACTIVE DAYS CREDIT ENDDATE MAX

Val / Var Mean

Val / Var Mean EXT SOURCE 3 0.403 Val / Var Mean EXT SOURCE 1 0.506 Val / Var Mean AMT GOODS PRICE 554k Val / Var Mean : CODE GENDER Val / Var Mean NAME\_INCOME\_TYPE\_Pensioner



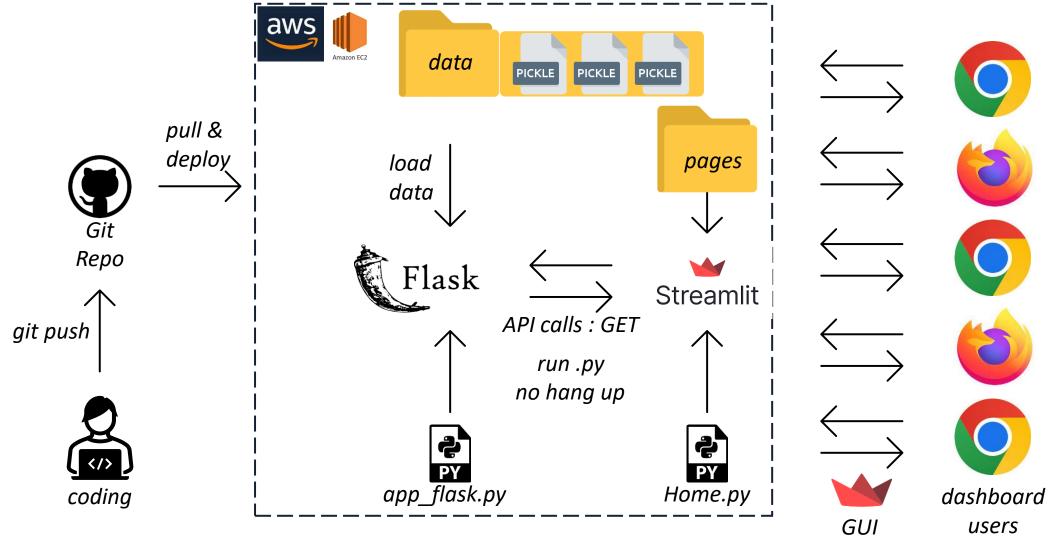
# Decision Model **BLACK BOX**



## Dashboard & déploiement

## Serveurs Flask / Streamlit

http://15.188.141.153:8501/



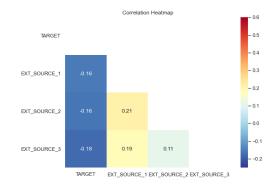
## Dashboard & déploiement

## Limites & préconisations

10 FN + FP Taille échantillon

1 | fonction coût métier

2 | détail 3 sources externes





3 | pré-traitement des données

## Questions?

## Merci!