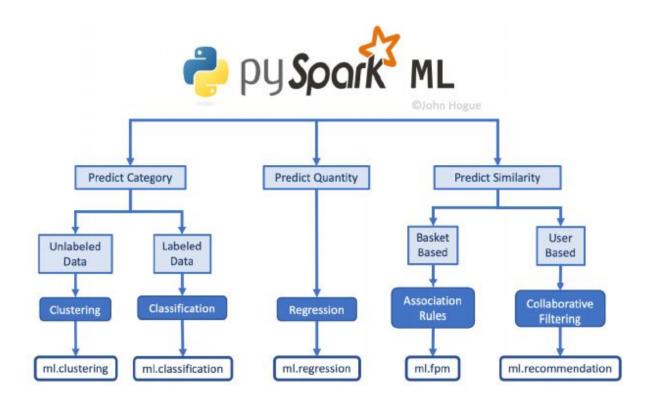
frauduleuses à l'aide de la bibliothèque MLLib d'Apache Spark

Détection des transactions bancaires

MLLib de Apache Spark



Données

Description des variables

- 'type' type de transaction : payment, transfer, cash out, cash in, debit.
- 'amount' montant de la transaction.
- 'nameOrig' identifiant de l'émetteur.
- 'oldbalanceOrg' solde de l'émetteur avant la transaction.
- 'newbalanceOrig' solde de l'émetteur après la transaction.
- 'nameDest' identifiant du destinataire.
- 'oldbalanceDest' solde du destinataire avant la transaction.
- 'newbalanceDest' solde du destinataire après la transaction.
- 'isFraud' 1 si la transaction est frauduleuse, sinon 0.
- 'isFraggedFraud' 1 si la transaction a été signalée frauduleuse, sinon 0.

EDA : Analyse exploratoire des données

```
type| amount| nameOrig|oldbalanceOrg|newbalanceOrig| nameDest|oldbalanceDest|newbalanceDest|isFraud|isFlaggedFraud|
step
   1 PAYMENT | 9839.64 | C1231006815 |
                                                160296.36 M1979787155
                                    170136.0
                                                                                0.0
                                                                                              0.0
   1 PAYMENT | 1864.28 | C1666544295 |
                                   21249.0 19384.72 M2044282225
                                                                                0.0
                                                                                               0.0
   1 TRANSFER | 181.0 C1305486145
                                        181.0
                                                        0.0 | C553264065 |
                                                                                0.0
                                                                                               0.0
   1 CASH OUT | 181.0 | C840083671
                                      181.0
                                                       0.0 C38997010
                                                                            21182.0
                                                                                               0.0
   1 PAYMENT | 11668.14 | C2048537720 |
                                                   29885.86 | M1230701703 |
                                     41554.0
                                                                                 0.0
                                                                                               0.0
```

```
root
```

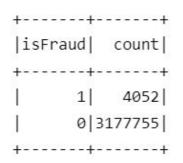
```
|-- step: string (nullable = true)
|-- type: string (nullable = true)
|-- amount: string (nullable = true)
|-- nameOrig: string (nullable = true)
|-- oldbalanceOrg: string (nullable = true)
|-- newbalanceOrig: string (nullable = true)
|-- nameDest: string (nullable = true)
|-- oldbalanceDest: string (nullable = true)
|-- newbalanceDest: string (nullable = true)
|-- isFraud: string (nullable = true)
|-- isFlaggedFraud: string (nullable = true)
```

```
# Get number of records
print("The data contain %d records." % df.count())
```

The data contain 6362620 records.

```
# Get number of columns
print("The data contain %d columns." % len(df.columns))
```

The data contain 11 columns.

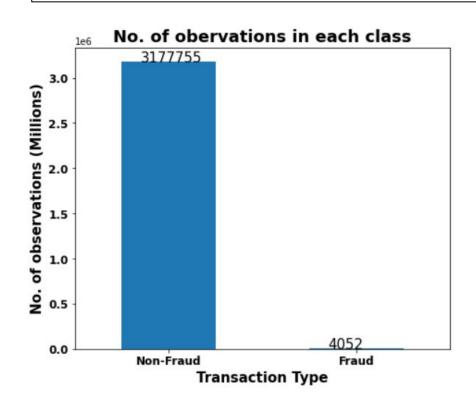


1. Statistique descriptive et visualisations

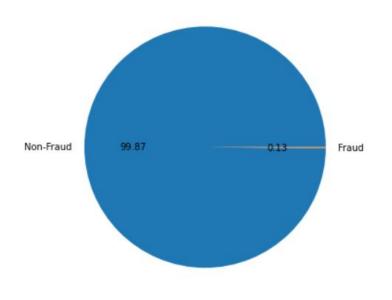
On peut observer les intervalles [min, max] des chacune des variables, les moyennes et les écart-types.

	summary	step	amount	oldbalanceOrg	newbalanceOrig	oldbalanceDest	newbalanceDest	isFraud
0	count	3181807	3181807	3181807	3181807	3181807	3181807	3181807
1	mean	243.33075419093615	180376.4924033053	834950.8743586149	856168.4637992503	1101763.094531916	1226409.6227083874	0.0012734901896940952
2	stddev	142.39545491604187	613321.2650007016	2891187.841932541	2927031.285909358	3369890.924303856	3654159.749004548	0.035663269790674586
3	min	1.0	0.0	0.0	0.0	0.0	0.0	0
4	max	743.0	6.98867313E7	5.958504037E7	4.958504037E7	3.5601588935E8	3.5617927892E8	1

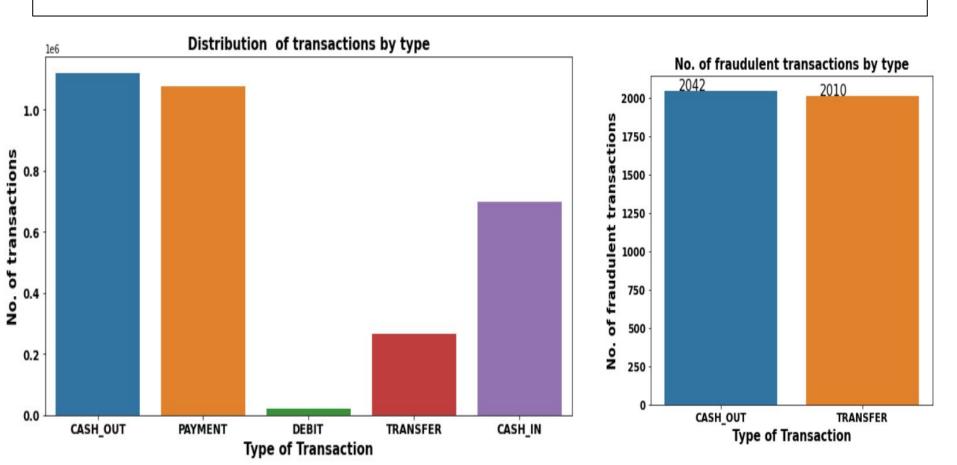
1. Statistique descriptive et visualisations



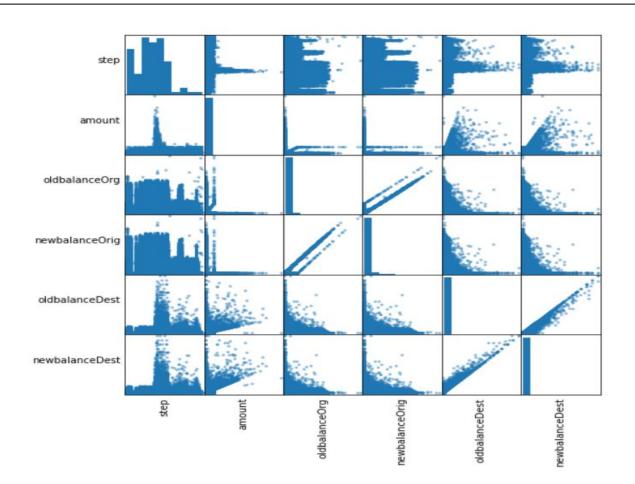
Percentage distribution of each class



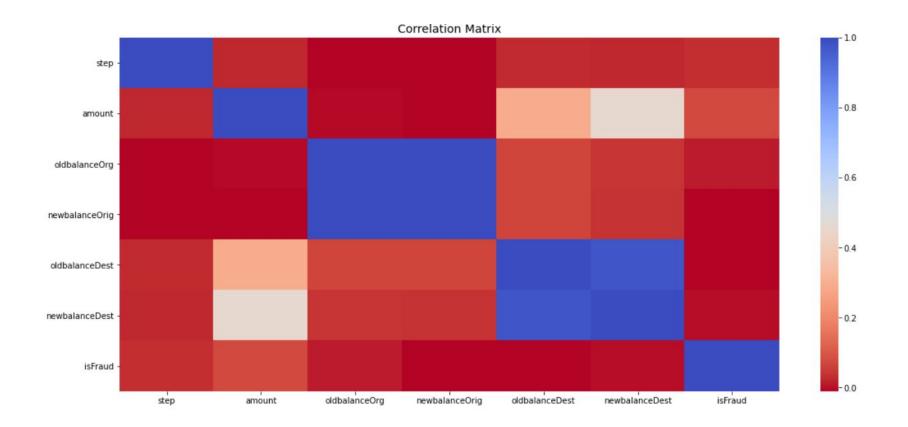
Statistique descriptive et visualisations



2. Corrélation entre variables



2. Corrélation entre variables



Prétraitement des données

1. Pré-traitements et indexation des données

indexed data.show(5) |step| amount|newbalanceOrig|oldbalanceDest|isFraud| num orig| num dest|typeIndexed|nameOrigIndexed|nameDestIndexed| 1.0 181.0 0.0 21182.0 1 8.40083671E8 8.40083671E8 0.0 0.0 0.0 0|1.912850431E9|1.912850431E9| 1.0 7861.64 168225.59 0.0 1.0 0.0 1.0 1.0 9644.94 0.0 0|1.900366749E9|1.900366749E9| 4.0 10845.0 0.0 0.0 1.0 | 2560.74 | 2509.26 | 0.0 0 0 1.648232591E9 1.648232591E9 1.0 0.0 1.0 1.0 | 1563.82 | 0.0 | 0 | 7.61750706E8 | 7.61750706E8 | 0.01 1.0 0.0 1.0

only showing top 5 rows

2. Transformation des données

```
features | isFraud |
(10,[0,1,3,5,6],[...]
[7861.64,176087.2...]
[9644.94,4465.0,0...]
[2560.74,5070.0,2...]
[1563.82,450.0,0....]
                         0
only showing top 5 rows
```

Classification des transactions

Objectif:

- Effectuer une classification binaire des transactions bancaires
- -> comparaison entre les différents modèles de classification

Modèle de classification : Logistic regression

```
lr model = lr train(train)
  lr eval = lr eval test(lr model, test)
|isFraud|prediction| count|
+-----+
     1 0.0 415
     0| 0.0|634751|
     1 1.0 380
     0 1.0 46
Recall: 0.4779874213836478
Precision: 0.892018779342723
F1 Score: 0.6224406224406225
Area under ROC = 0.9909929508193294
Area under PR = 0.5130515281003208
```

Modèle de classification : Decision Tree Classifier

```
dt eval = Dt eval test(dt model, test)
           features | rawPrediction|probability|prediction|
[23.31,45360.0,45...|[1437028.0,0.0]| [1.0,0.0]|
|(10,[0,5,6,7,9],[...|[1437028.0,0.0]| [1.0,0.0]|
                                                   0.0
|[112.56,609035.85...|[1437028.0,0.0]| [1.0,0.0]|
                                                   0.0
|[154.87,9339.0,91...|[1437028.0,0.0]| [1.0,0.0]|
                                                   0.0
[339.82,12076.0,1...|[1437028.0,0.0]| [1.0,0.0]|
                                                   0.0
only showing top 5 rows
|isFraud|prediction| count|
      1 0.0 244
      0 | 0.0 | 634761 |
     1 1.0 551
          1.0
                     36
   ----+
Recall: 0.6930817610062893
Precision: 0.938671209540034
F1 Score: 0.7973950795947901
Area under ROC = 0.7798224050687704
Area under PR = 0.4142278715874976
```

Modèle de classification : Random Forest Classifier

```
rf model = rf train(train)
  rf eval = rf eval test(rf model, test)
|isFraud|prediction| count|
     1 0.0 479
     0 | 0.0 | 634796 |
     1 1.0 316
     0 1.0 1
  ----+
Recall: 0.39748427672955977
Precision: 0.9968454258675079
F1 Score: 0.5683453237410073
Area under ROC = 0.9692023953420934
Area under PR = 0.7132785942938621
```

Modèle de classification : Gradient-Boosted Tree Classifier

```
GBT model = GBT train(train)
   GBT eval = GBT eval test(GBT model, test)
|isFraud|prediction| count|
      1 0.0 237
      0 | 0.0 | 634783 |
             1.0 558
         1.0 14
Recall: 0.7018867924528301
Precision: 0.9755244755244755
F1 Score: 0.8163862472567667
Area under ROC = 0.9829058015208803
Area under PR = 0.7807439732858391
```

Modèle de classification : Naive Bayes

```
NB model = NB train(train)
  NB eval = NB eval test(NB model, test)
 -----+
|isFraud|prediction| count|
   1 0.0 355
     0| 0.0|596227|
    1 1.0 440
     0| 1.0| 38570|
Recall: 0.5534591194968553
Precision: 0.011279159189951295
F1 Score: 0.022107775405099866
Area under ROC = 0.49412195289727784
Area under PR = 0.0012358315240476004
```

Classification des transactions

Observation: Le GRadient Boost Classifier est plus performant que les autres modèles si on considère toutes les métriques de l'évaluation. Donc c'est celui qui sera utilisé pour classer les nouvelles transactions.

FIN