Incognia Case

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Methodology

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- 2. Literature
- 3. Data
- 4. Exploration
- 5. Data Product
- 6. Naive Solution
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Problem

Chargeback Fraud Detection

Definition

Chargeback fraud occurs when consumers fraudulently attempt to secure a refund using the chargeback process. Instead of contacting the merchant directly for a refund, consumers dispute the transaction with their bank, thus initiating the chargeback process. Consumers will falsely complain that the product they ordered was delivered defective or not at all, that they did not authorize the transaction, or that they had requested the cancellation of a recurring transaction and were charged anyway. Whatever reason they give, chargeback fraud is when the real reason is something else entirely (https://www.cardinalcommerce.com/fraud/chargebacks/what-is-chargeback-fraud)

Example

A consumer uses the iFood's app for a delivery in a different location than where he/she lives at the moment. Lets say for an example that this consumer order 2 sushi combos (\$200) to be delivered to a friends house. However, a little after the delivery is concluded with success the same consumer ask its cardhoulder for a refund claiming that someone took over its account and ordered in his place. Depending on the consumers's cardholder, the \$200 will be refunded, eventhough it was a chargeback fraud. It would be really important for the iFood's app to known the consumer's location during its daily journey to be able to detect such chargeback frauds.

If such data was available at the time of the order and the same consumer have already been many times in the order locations then a detector (model or specialist) would be able to label dispute such chargeback fraud.

Literature

Rare Event Detection (Temporal)

Types:

- full length supervised time series classification: detect if a rare event occured in a fixed period of time.
- early supervised time-series classification: detect if a rare event will occur in the next period of time to be able
 to fixed or countermeasure.
- unbalanced non-temporal classification: in some rare event detection applications, instances are transformed without considering the time dimension (similar to anomaly detection).

Objective: classifify the rare class

Metric: AUC, Recall of the rare class

Techniques:

- Rare event logistic regression
- Kullback-Leibler divergence
- LSTMs
- SVMs

Anomaly Detection (Non-Temporal)

Types: highly unbalanced supervised classification

Objective: classifify the anomaly classes

Metric: AUC, recall of the rare class

Classification Techniques: popular supervised classifiers

Data

Payments

coluna	tipo	descrição
id	string	identificador da transação
account_id	string	identificador da conta
device_id	string	identificador do dispositivo
installation_id	string	identificador da instalação
timestamp	long	timestamp (unixtime - em ms)
device_age_ms	long	idade do dispositivo (em ms)
n_accounts_by_device_30d	int	número de contas acessadas pelo dispositivo nos últimos 30 dias
sum_values_by_device_30d	float	soma dos valores das transações realizadas pelo dispositivo nos últimos 30 dias
sum_values_by_device_3d	float	soma dos valores das transações realizadas pelo dispositivo nos últimos 3 dias
sum_values_by_installation_30d	float	soma dos valores das transações realizadas pela instalação nos últimos 30 dias
sum_values_by_installation_3d	float	soma dos valores das transações realizadas pela instalação nos últimos 3 dias
low_geo_chargeback_rate_30d	float	taxa de chargeback rate da região geográfica associada à transação (precisão geográfica baixa)
medium_geo_chargeback_rate_30d	float	taxa de chargeback rate da região geográfica associada à transação (precisão geográfica média)
high_geo_chargeback_rate_30d	float	taxa de chargeback rate da região geográfica associada à transação (precisão geográfica alta)
value	float	valor da transação
chargedback	float	se a transação ocasionou um chargeback



Exploration

Data Exploration

Resume

Observation's Begin

2021-04-30 21:01:20

Table Size

648173

Max Payments by Account

572

Number of Payments

648173

Max Account by Device

31281

Outlier?

~5 months

Payments per Accounts

5.62

Max Installation by Device

31976

Observation's End

2021-09-21 02:50:43

Accounts per Devices

1.37

Number of Chargedbacks

3249

Installation per Device

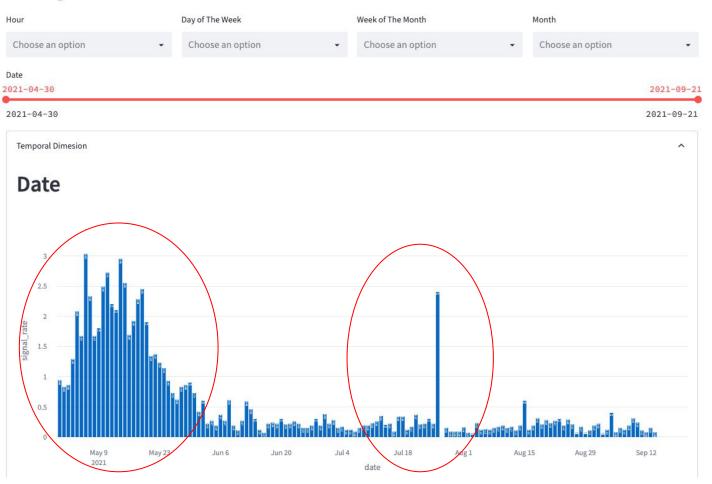
1.42

Chargedback Rate

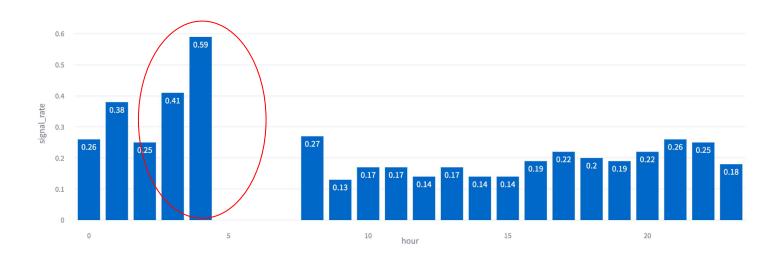
0.501

Highly unbalanced

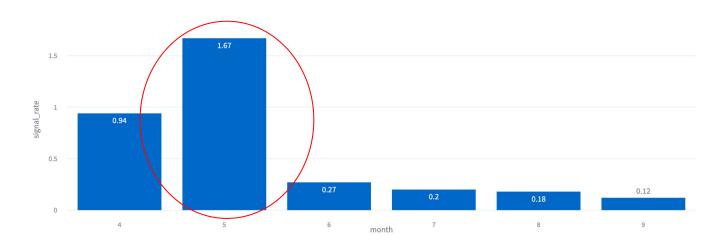
Temporal Dimesion



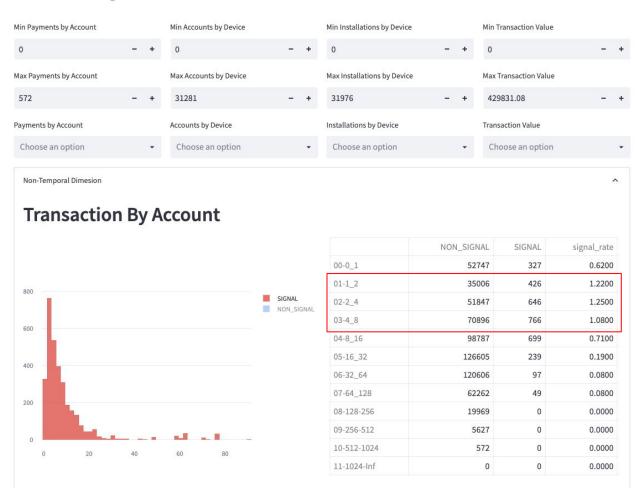
Hour



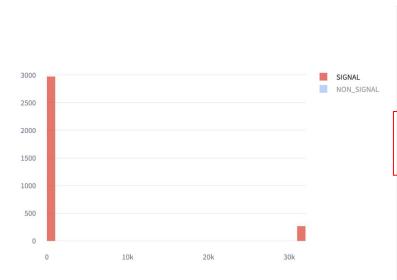
Month



Non-Temporal Dimesion

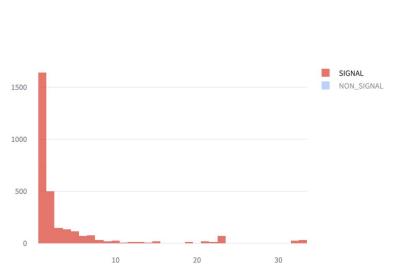


Accounts By Device



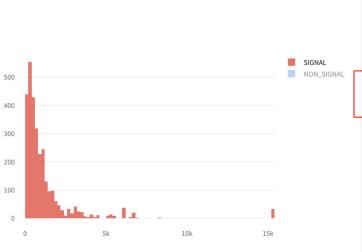
	NON_SIGNAL	SIGNAL	signal_rate
00-0_1	446138	1509	0.3400
01-1_2	29892	451	1.5100
02-2_4	13360	336	2.5100
03-4_8	4082	338	8.2800
04-8_16	2092	99	4.7300
05-16_32	768	76	9.9000
06-32_64	396	169	42.6800
07-64_128	0	0	0.0000
08-128-256	0	0	0.0000
09-256-512	0	0	0.0000
10-512-1024	0	0	0.0000
11-1024-Inf	148196	271	0.1800

Installation By Device



	NON_SIGNAL	SIGNAL	signal_rate
00-0_1	439238	1643	0.3700
01-1_2	40158	502	1.2500
02-2_4	11500	279	2.4300
03-4_8	3601	292	8.1100
04-8_16	1255	98	7.8100
05-16_32	938	133	14.1800
06-32_64	38	31	81.5800
07-64_128	0	0	0.0000
08-128-256	0	0	0.0000
09-256-512	0	0	0.0000
10-512-1024	0	0	0.0000
11-1024-Inf	0	0	0.0000

Value By Account



	NON_SIGNAL	SIGNAL	signal_rate
00-0_50	24283	43	0.1800
01-50_100	22698	107	0.4700
02-100_250	42487	431	1.0100
03-250_500	43030	634	1.4700
04-500_1000	56330	756	1.3400
05-1000_2500	97674	707	0.7200
06-2500_5000	99859	176	0.1800
07-5000_10000	72969	91	0.1200
08-10000-25000	29931	33	0.1100
09-25000-50000	5534	0	0.0000
10-50000-100000	1695	0	0.0000
11-100000-Inf	160	0	0.0000

Data Product

Demo



Naive Approach

Supervised Classification

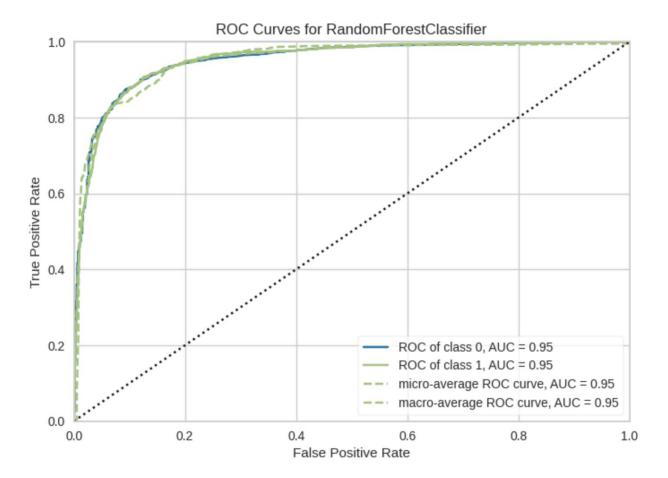
	Model	Accuracy	AUC	Recall	Prec.	F1	Kappa	мсс	TT (Sec)
qda	Quadratic Discriminant Analysis	0.9698	0.8976	0.5035	0.0852	0.1457	0.1381	0.1982	0.3000
lda	Linear Discriminant Analysis	0.9868	0.9216	0.3923	0.1659	0.2332	0.2276	0.2494	0.5290
dt	Decision Tree Classifier	0.9920	0.6260	0.2563	0.2363	0.2457	0.2417	0.2420	2.7500
et	Extra Trees Classifier	0.9949	0.8783	0.2142	0.5111	0.3017	0.2996	0.3287	3.2990
rf	Random Forest Classifier	0.9951	0.9100	0.1408	0.5686	0.2253	0.2237	0.2809	6.1780
xgboost	Extreme Gradient Boosting	0.9950	0.9532	0.1004	0.5427	0.1688	0.1674	0.2311	0.7860
lightgbm	Light Gradient Boosting Machine	0.9932	0.9375	0.0848	0.1703	0.1126	0.1096	0.1166	18.6980
ada	Ada Boost Classifier	0.9944	0.9446	0.0631	0.2811	0.1026	0.1008	0.1306	12.4760
catboost	CatBoost Classifier	0.9950	0.9519	0.0534	0.5729	0.0975	0.0967	0.1732	4.0950
gbc	Gradient Boosting Classifier	0.9948	0.9418	0.0275	0.4313	0.0513	0.0507	0.1056	58.6310
svm	SVM - Linear Kernel	0.9899	0.0000	0.0173	0.0175	0.0174	0.0123	0.0123	5.7750
knn	K Neighbors Classifier	0.9948	0.5450	0.0016	0.1333	0.0032	0.0031	0.0141	1.5990
lr	Logistic Regression	0.9949	0.8580	0.0000	0.0000	0.0000	0.0000	0.0000	0.5740
nb	Naive Bayes	0.9949	0.8584	0.0000	0.0000	0.0000	0.0000	0.0000	0.1500
ridge	Ridge Classifier	0.9949	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.1610
dummy	Dummy Classifier	0.9949	0.5000	0.0000	0.0000	0.0000	0.0000	0.0000	0.1300

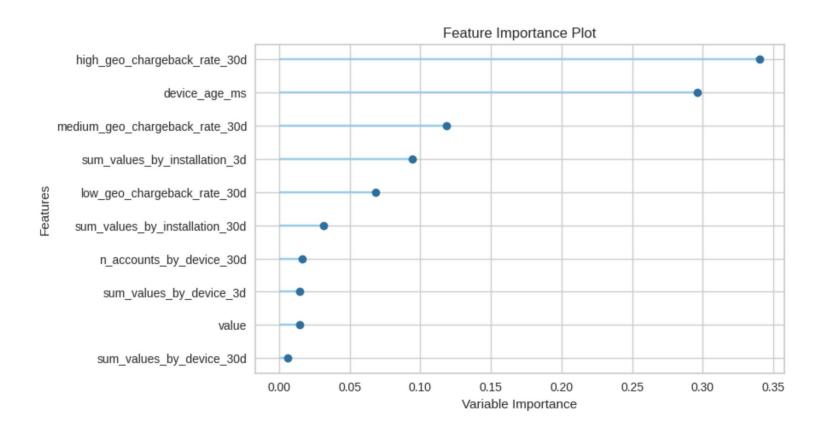
Random Forest (Naive)

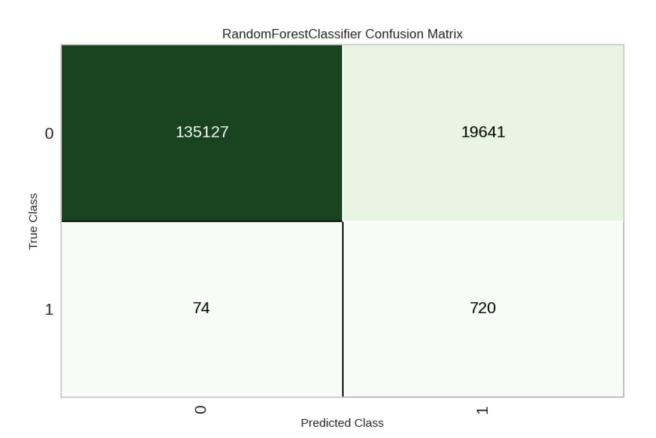
	Accuracy	AUC	Recall	Prec.	F1	Карра	мсс
Fold							
0	0.9953	0.9164	0.1676	0.6739	0.2684	0.2669	0.3346
1	0.9947	0.8979	0.1027	0.4130	0.1645	0.1628	0.2041
2	0.9948	0.8966	0.0973	0.4615	0.1607	0.1592	0.2102
3	0.9950	0.8976	0.1075	0.5556	0.1802	0.1788	0.2429
4	0.9951	0.9263	0.1667	0.5849	0.2594	0.2577	0.3105
5	0.9952	0.8877	0.1667	0.5962	0.2605	0.2588	0.3135
6	0.9951	0.9262	0.1297	0.5854	0.2124	0.2109	0.2740
7	0.9953	0.9188	0.1568	0.6444	0.2522	0.2507	0.3163
8	0.9952	0.9243	0.1622	0.5882	0.2542	0.2526	0.3072
9	0.9951	0.9082	0.1514	0.5833	0.2403	0.2387	0.2955
Mean	0.9951	0.9100	0.1408	0.5686	0.2253	0.2237	0.2809
Std	0.0002	0.0135	0.0273	0.0739	0.0401	0.0401	0.0440

Random Forest (Tuned for Recall)

	Accuracy	AUC	Recall	Prec.	F1	Карра	мсс
Fold							
0	0.8705	0.9501	0.8865	0.0339	0.0652	0.0560	0.1585
1	0.8728	0.9393	0.8865	0.0345	0.0663	0.0571	0.1602
2	0.8769	0.9436	0.8757	0.0351	0.0676	0.0584	0.1609
3	0.8768	0.9516	0.8925	0.0359	0.0691	0.0598	0.1648
4	0.8739	0.9376	0.8656	0.0342	0.0657	0.0564	0.1571
5	0.8732	0.9463	0.8978	0.0351	0.0676	0.0584	0.1632
6	0.8777	0.9292	0.8541	0.0346	0.0665	0.0572	0.1571
7	0.8708	0.9411	0.8865	0.0339	0.0654	0.0561	0.1587
8	0.8784	0.9552	0.8865	0.0360	0.0692	0.0600	0.1644
9	0.8698	0.9538	0.8973	0.0341	0.0657	0.0564	0.1602
Mean	0.8741	0.9448	0.8829	0.0347	0.0668	0.0576	0.1605
Std	0.0030	0.0078	0.0133	0.0008	0.0014	0.0014	0.0027







Data-Driven Approach

Supervised Classification + More Features

Device

Installation

Account

sum_obs_values_by_device_30d
n_obs_transaction_by_device_30d
n_obs_installation_by_device_30d
n_obs_accounts_by_device_30d
sum_obs_values_by_device_15d
n_obs_transaction_by_device_15d
n_obs_installation_by_device_15d
sum_obs_values_by_device_15d
sum_obs_values_by_device_7d
n_obs_transaction_by_device_7d
n_obs_installation_by_device_7d
n_obs_accounts_by_device_7d
n_obs_accounts_by_device_7d

sum_obs_values_by_installation_30d n_obs_transaction_by_installation_30d sum_obs_values_by_installation_15d n_obs_transaction_by_installation_15d sum_obs_values_by_installation_7d n_obs_transaction_by_installation_7d

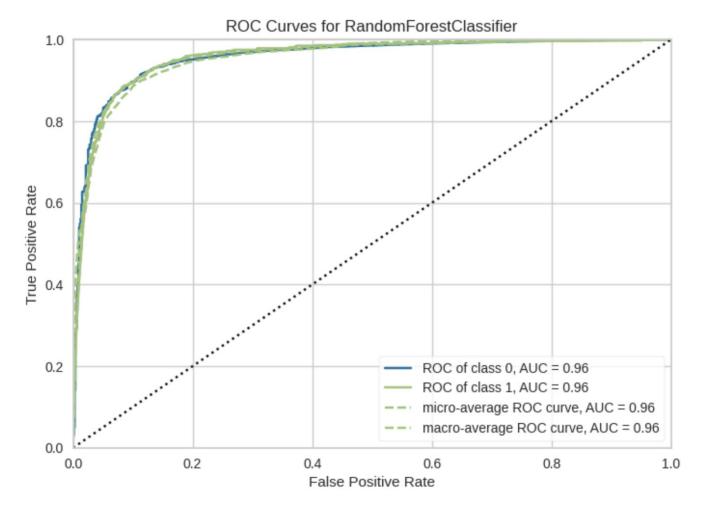
sum_obs_values_by_account_30d
n_obs_transaction_by_account_30d
n_obs_devices_by_account_30d
n_obs_installation_by_account_30d
sum_obs_values_by_account_15d
n_obs_transaction_by_account_15d
n_obs_devices_by_account_15d
sum_obs_values_by_account_15d
sum_obs_values_by_account_7d
n_obs_transaction_by_account_7d
n_obs_devices_by_account_7d
n_obs_installation_by_account_7d
n_obs_installation_by_account_7d

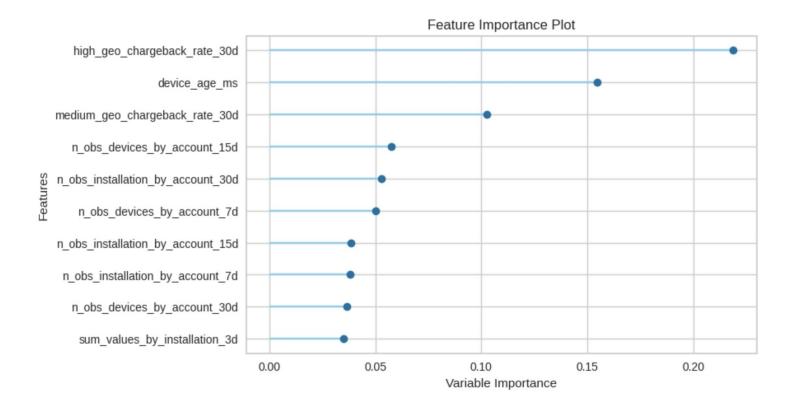
Random Forest (Tuned for Recall)

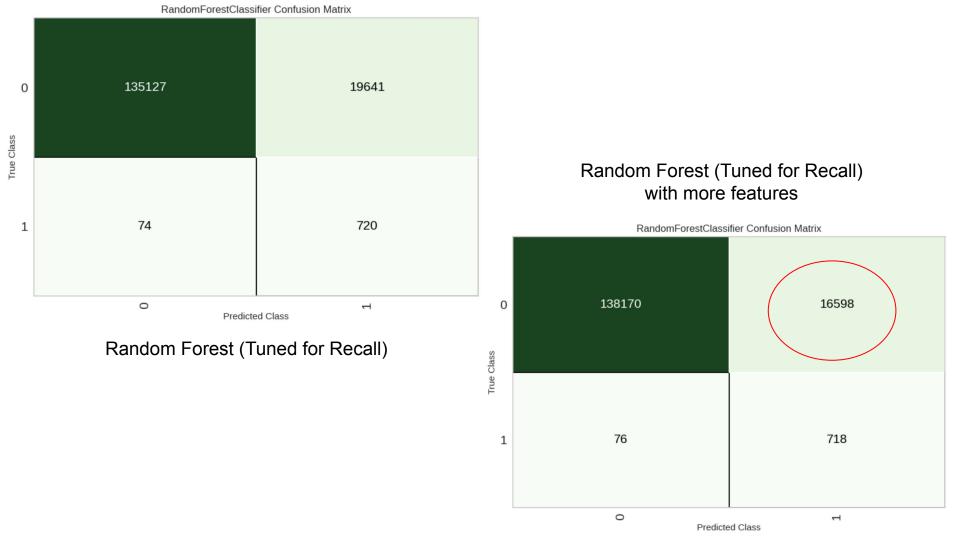
	Accuracy	AUC	Recall	Prec.	F1	Kappa	мсс
Fold							
0	0.8705	0.9501	0.8865	0.0339	0.0652	0.0560	0.1585
1	0.8728	0.9393	0.8865	0.0345	0.0663	0.0571	0.1602
2	0.8769	0.9436	0.8757	0.0351	0.0676	0.0584	0.1609
3	0.8768	0.9516	0.8925	0.0359	0.0691	0.0598	0.1648
4	0.8739	0.9376	0.8656	0.0342	0.0657	0.0564	0.1571
5	0.8732	0.9463	0.8978	0.0351	0.0676	0.0584	0.1632
6	0.8777	0.9292	0.8541	0.0346	0.0665	0.0572	0.1571
7	0.8708	0.9411	0.8865	0.0339	0.0654	0.0561	0.1587
8	0.8784	0.9552	0.8865	0.0360	0.0692	0.0600	0.1644
9	0.8698	0.9538	0.8973	0.0341	0.0657	0.0564	0.1602
Mean	0.8741	0.9448	0.8829	0.0347	0.0668	0.0576	0.1605
Std	0.0030	0.0078	0.0133	0.0008	0.0014	0.0014	0.0027

Random Forest (Tuned for Recall) with more features

	Accuracy	AUC	Recall	Prec.	F1	Kappa	мсс
Fold							
0	0.8916	0.9608	0.9081	0.0411	0.0787	0.0696	0.1802
1	0.8895	0.9524	0.9189	0.0408	0.0781	0.0691	0.1806
2	0.8967	0.9584	0.9027	0.0428	0.0818	0.0728	0.1839
3	0.8966	0.9594	0.9032	0.0431	0.0822	0.0731	0.1844
4	0.8832	0.9471	0.8656	0.0368	0.0706	0.0614	0.1642
5	0.8926	0.9505	0.8925	0.0410	0.0784	0.0693	0.1781
6	0.8920	0.9407	0.8595	0.0392	0.0750	0.0659	0.1699
7	0.8914	0.9510	0.8973	0.0406	0.0777	0.0686	0.1776
8	0.8915	0.9630	0.9459	0.0426	0.0816	0.0726	0.1882
9	0.8889	0.9568	0.8973	0.0397	0.0760	0.0669	0.1753
Mean	0.8914	0.9540	0.8991	0.0408	0.0780	0.0689	0.1782
Std	0.0037	0.0066	0.0233	0.0018	0.0034	0.0034	0.0067







Next Steps

Next Steps

- Play with more hyperparameters and data preparation
- Try more robust models using the sequence of transactions (temporal)
 - o RNN, LSTM (sequential models)
- Try custom models for anomaly detection
 - Isolation Forest
- Ask for the spatial dimension of the transactions