



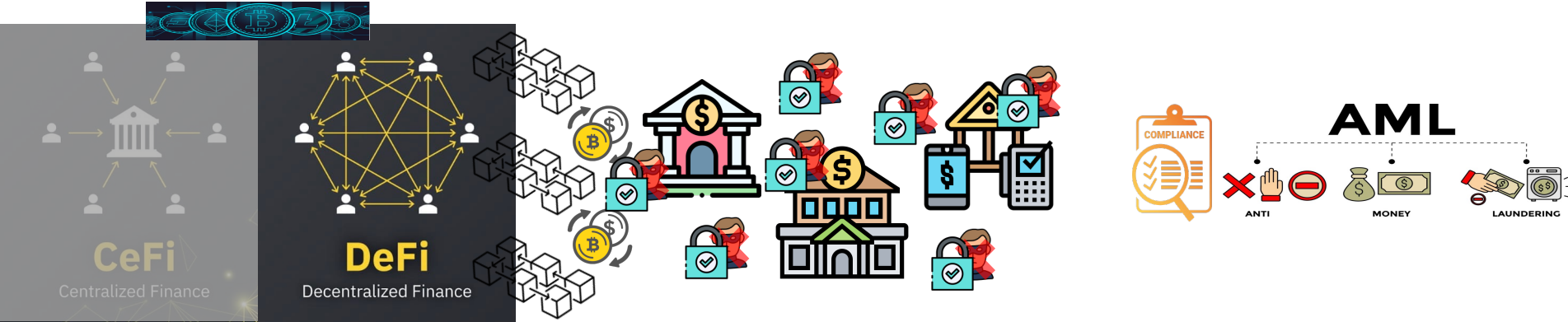
Demystifying Fraudulent Transactions and Illicit Nodes in the Bitcoin Network for Financial Forensics

Youssef Elmougy* and Ling Liu

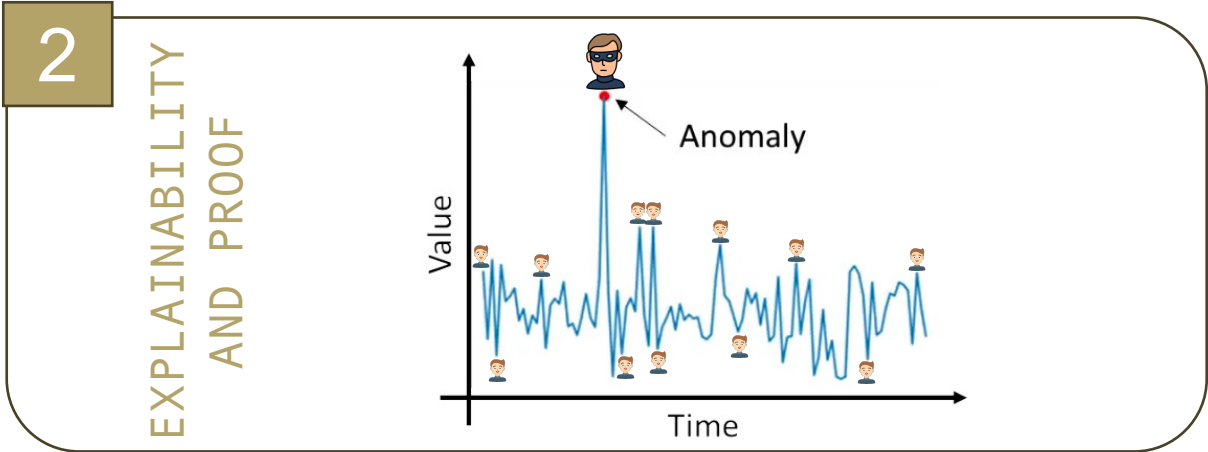
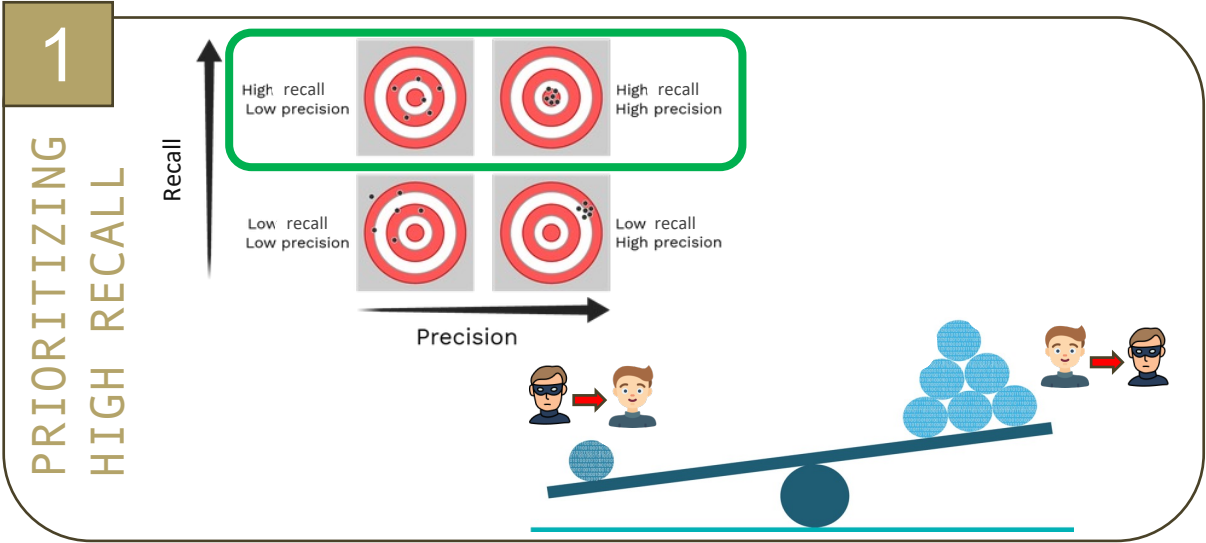
Georgia Institute of Technology, Atlanta GA USA

* Corresponding Author and Presenting Author

The Catalyst for Cryptocurrency Integration into Traditional Financial Institutions



Fraud detection models trained for financial forensics should:



Downfalls of Available Financial Forensics Datasets

Largest labelled Bitcoin transaction dataset currently publicly available:

Consists of over 203k transactions labelled illicit, licit, and unknown

Elliptic Data Set

Bitcoin Transaction Graph



Available at:

<https://www.kaggle.com/datasets/ellipticco/elliptic-data-set>

✘ Elliptic Data Set consists of only Bitcoin transactions, without features of the addresses involved and the different interactions between pairs of addresses

Prominent downfall: when a model predicts an illicit transaction, the addresses responsible cannot be clearly identified since a transaction may be associated with several input and output addresses

GIVEN THIS MOTIVATION, WE MAKE TWO CONTRIBUTIONS

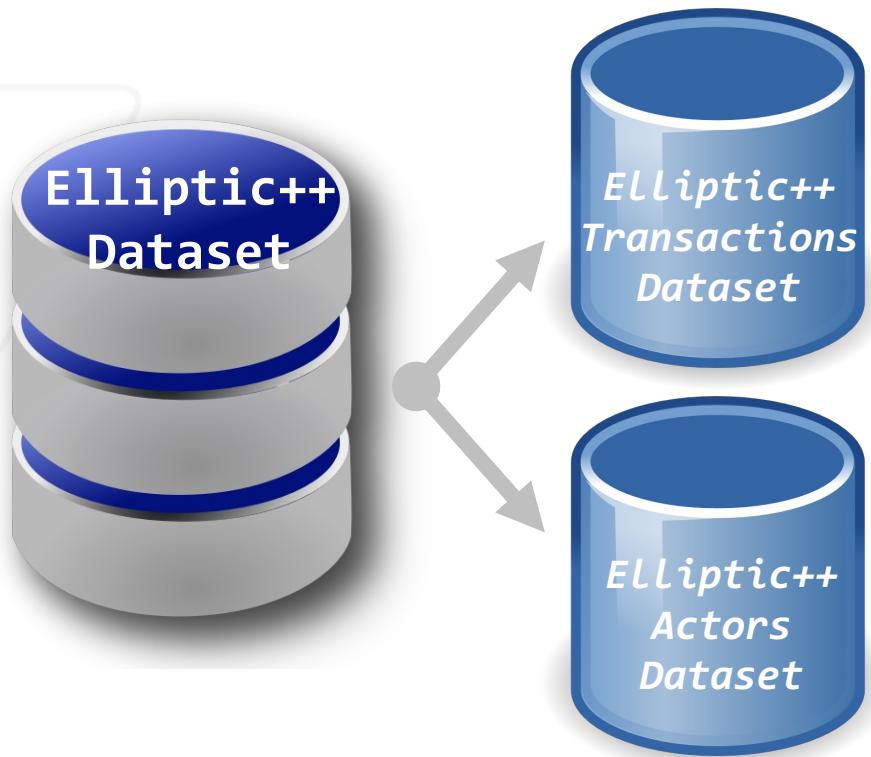
Dataset: <https://www.github.com/git-disl/EllipticPlusPlus>

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses

1st Contribution: The Elliptic++ Dataset



Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



- 203k transactions
- 234k payment flows
- 183 transaction features
- Labels: **illicit, licit, unknown**

- 822k wallet addresses
- 1.27M temporal occurrences
- 56 address features
- Labels: **illicit, licit, unknown**

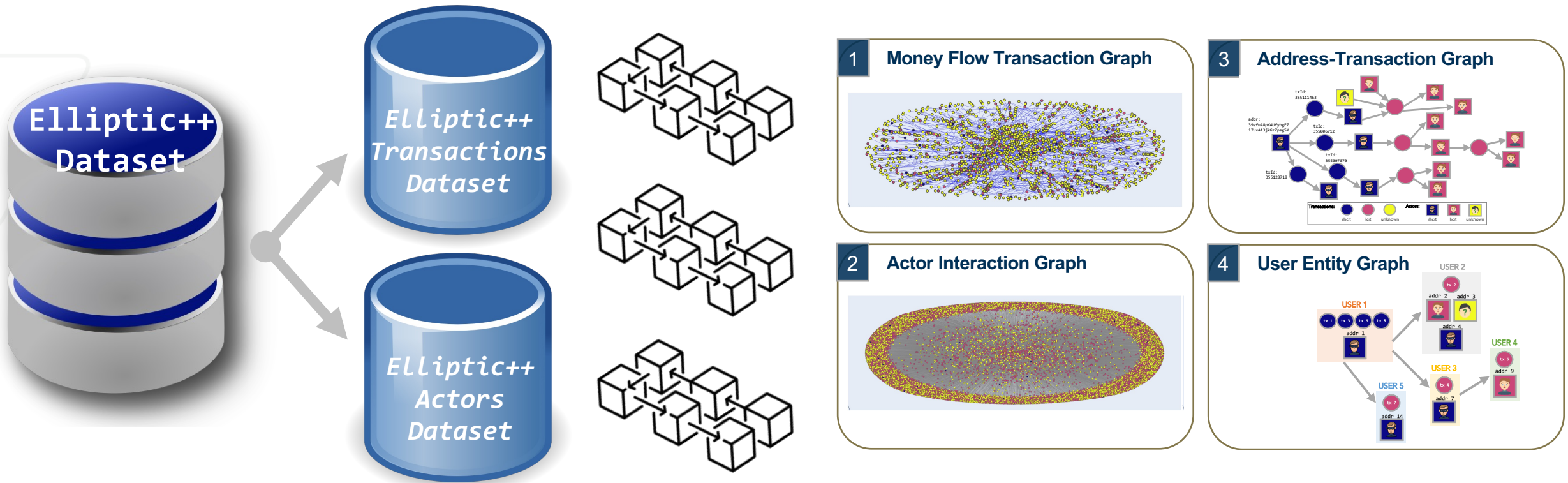
Forensics performed using The Elliptic++ Dataset:

- Identifies *fraudulent transactions* and *dishonest actors* in the Bitcoin network
- Allows *explainability* as to why a wallet address is *illusive*
- Characterizes *addresses as the center* of the risk detection model

1st Contribution: The Elliptic++ Dataset



Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



2nd Contribution: Fraud Detection using The Elliptic++ Dataset

Fraud detection by combining diverse ML algorithms and feature optimizations

Model	Precision	Recall	F1 Score	Micro-F1
RF ^{TX}	0.975	0.719	0.828	0.980
RF ^{TX} ψ	0.986	0.727	0.836	0.981
RF+XGB ^{TX}	0.977	0.706	0.820	0.979
RF+XGB ^{TX} ψ	0.987	0.717	0.826	0.980
RF+MLP+XGB ^{TX}	0.962	0.723	0.826	0.980
RF+MLP+XGB ^{TX} ψ	0.968	0.729	0.834	0.980

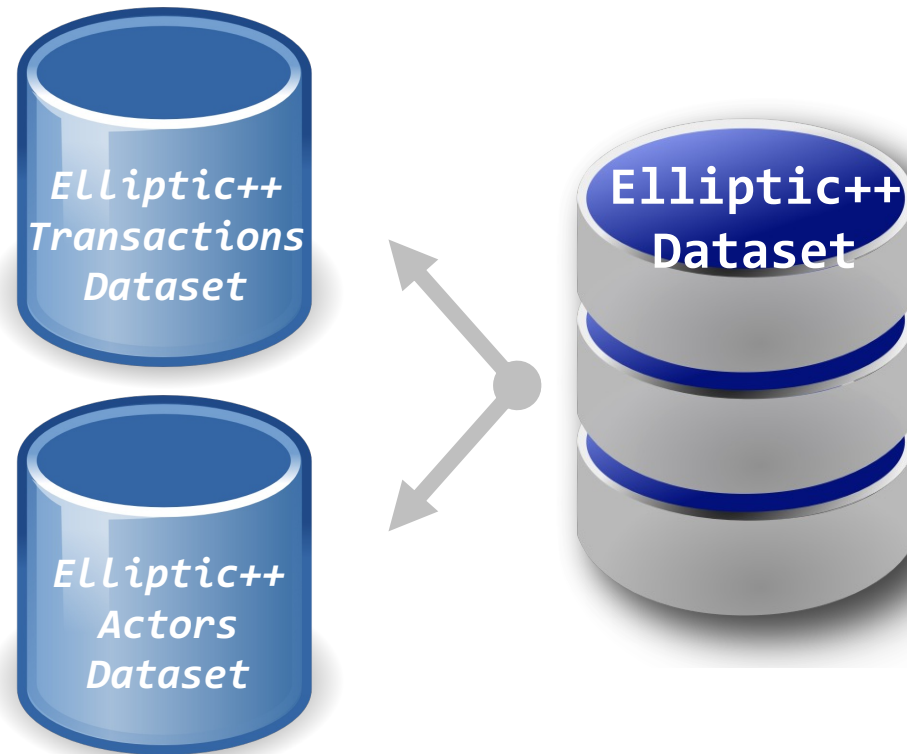
Model	Precision	Recall	F1 Score	Micro-F1
RF ^{AR}	0.911	0.789	0.845	0.990
RF ^{AR} ψ	0.921	0.802	0.858	0.990
RF+XGB ^{AR}	0.959	0.530	0.682	0.982
RF+XGB ^{AR} ψ	0.967	0.543	0.686	0.982
RF+MLP+XGB ^{AR}	0.933	0.572	0.709	0.983
RF+MLP+XGB ^{AR} ψ	0.945	0.601	0.718	0.984

Allows for in-depth understanding of the *root cause of fraudulent activities* in cryptocurrency transactions through *semantic and statistical explainability*, shining light on the strategies for fraud detection and prevention

The Elliptic++ Dataset

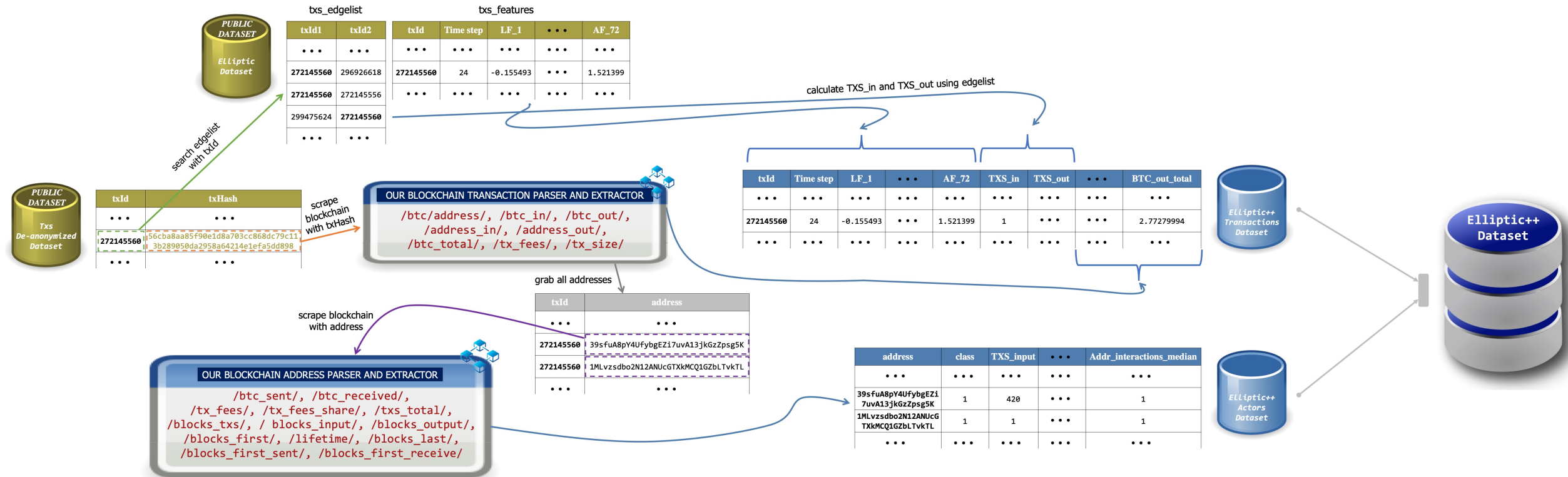
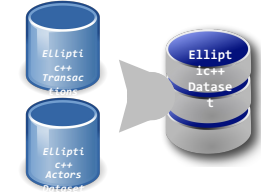


Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



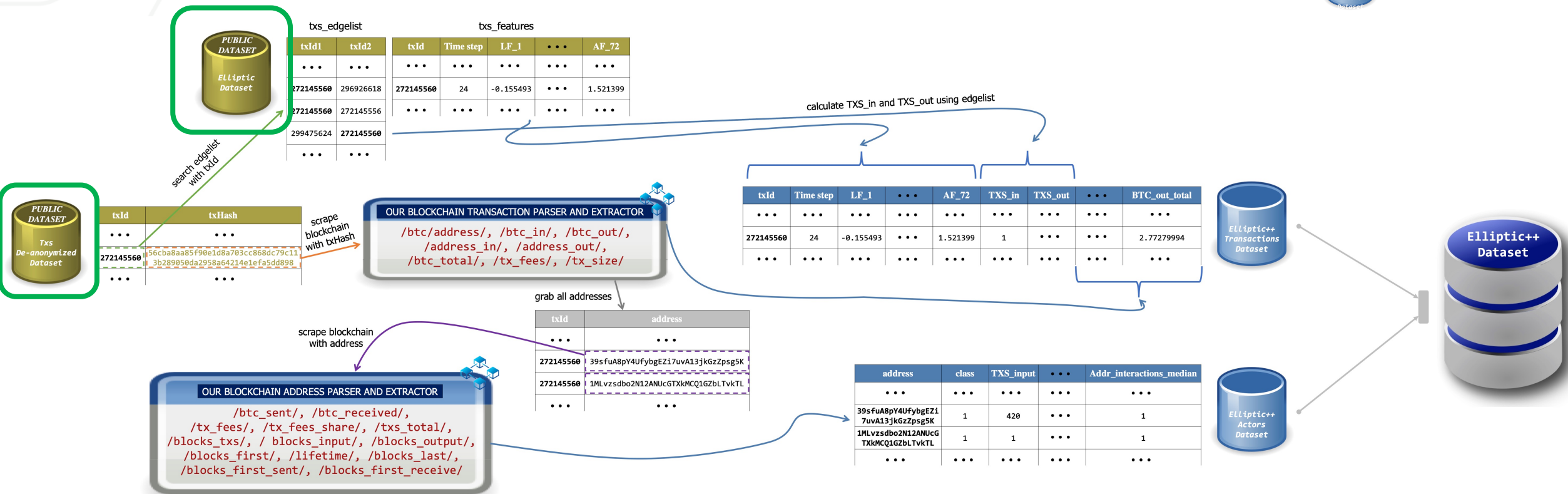
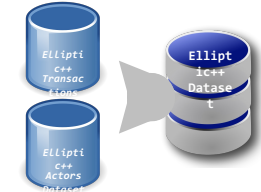
The Elliptic++ Dataset: Collection Pipeline

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



The Elliptic++ Dataset: Collection Pipeline

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



- Elliptic Dataset: www.kaggle.com/datasets/elliptico/elliptic-data-set
- Txs De-anonymized Dataset: www.kaggle.com/datasets/alexbenzik/deanonymized-995-pct-of-elliptic-transactions
- Bitcoin Blockchain: www.blockchain.com

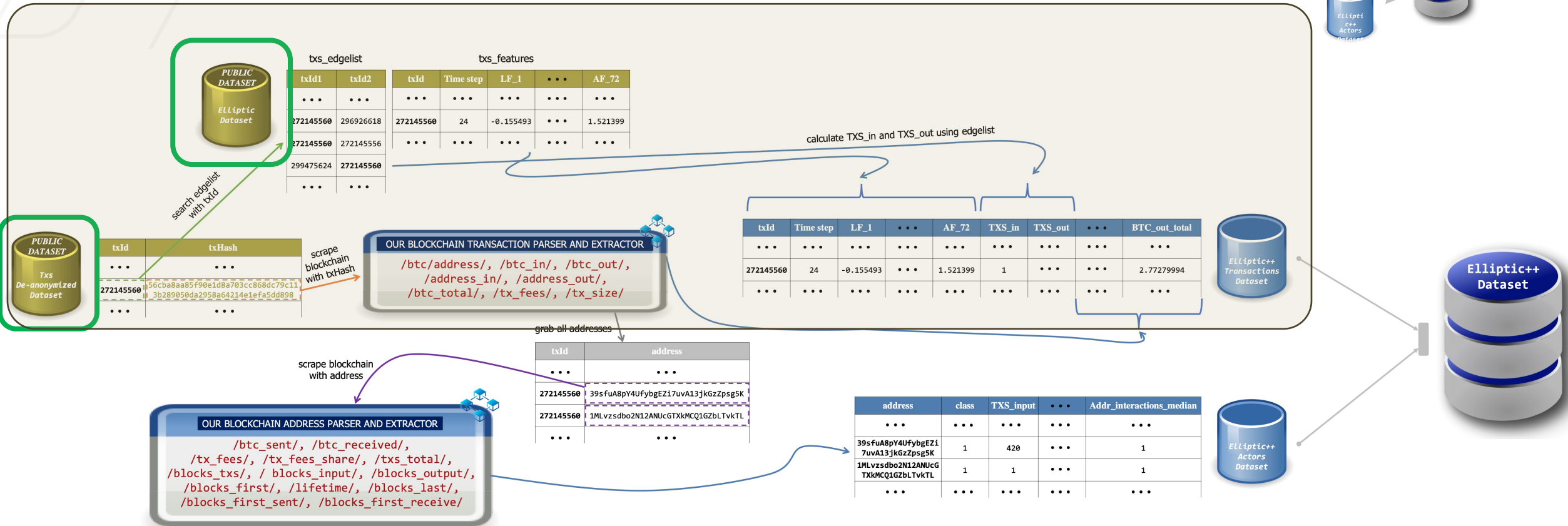
Dataset: <https://www.github.com/git-disl/EllipticPlusPlus>

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



The Elliptic++ Dataset: Collection Pipeline

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



- Elliptic Dataset: www.kaggle.com/datasets/elliptico/elliptic-data-set
- Txs De-anonymized Dataset: www.kaggle.com/datasets/alexbenzik/deanonymized-995-pct-of-elliptic-transactions
- Bitcoin Blockchain: www.blockchain.com

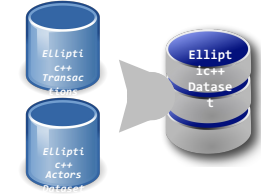
Dataset: <https://www.github.com/git-disl/EllipticPlusPlus>

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



The Elliptic++ Dataset: Collection Pipeline

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



txs_edgelist

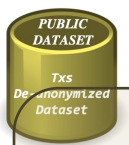
txId1	txId2
...	...
272145560	296926618
272145560	272145556
299475624	272145560
...	...

txs_features

txId	Time step	LF_1	...	AF_72
...
272145560	24	-0.155493	...	1.521399
...

calculate TXS_in and TXS_out using edgelist

txId	Time step	LF_1	...	AF_72	TXS_in	TXS_out	...	BTC_out_total
...
272145560	24	-0.155493	...	1.521399	1	2.77279994
...



txId	txHash
...	...
272145560	56cb88aa85f90e1d8a/05cc868dc/9c113b289050da2958a64214e1efa5dd899
...	...

search edgelist with txId

scrape blockchain with txHash

OUR BLOCKCHAIN TRANSACTION PARSER AND EXTRACTOR

```

/btc/address/, /btc_in/, /btc_out/,
/address_in/, /address_out/,
/btc_total/, /tx_fees/, /tx_size/
    
```

grab all addresses

txId	address
...	...
272145560	39sFuA8pY4UfybgEzi7uvA13jkGzZpsG5K
272145560	1MLvzsdb02N12ANUCGTxkMCQ1GZbLTvkTL
...	...

scrape blockchain with address

OUR BLOCKCHAIN ADDRESS PARSER AND EXTRACTOR

```

/btc_sent/, /btc_received/,
/tx_fees/, /tx_fees_share/, /txs_total/,
/blocks_txs/, /blocks_input/, /blocks_output/,
/blocks_first/, /lifetime/, /blocks_last/,
/blocks_first_sent/, /blocks_first_receive/
    
```

address	class	TXS_input	...	Addr_interactions_median
...
39sFuA8pY4UfybgEzi7uvA13jkGzZpsG5K	1	420	...	1
1MLvzsdb02N12ANUCGTxkMCQ1GZbLTvkTL	1	1	...	1
...



- Elliptic Dataset: www.kaggle.com/datasets/elliptico/elliptic-data-set
- Txs De-anonymized Dataset: www.kaggle.com/datasets/alexbenzik/deanonymized-995-pct-of-elliptic-transactions
- Bitcoin Blockchain: www.blockchain.com

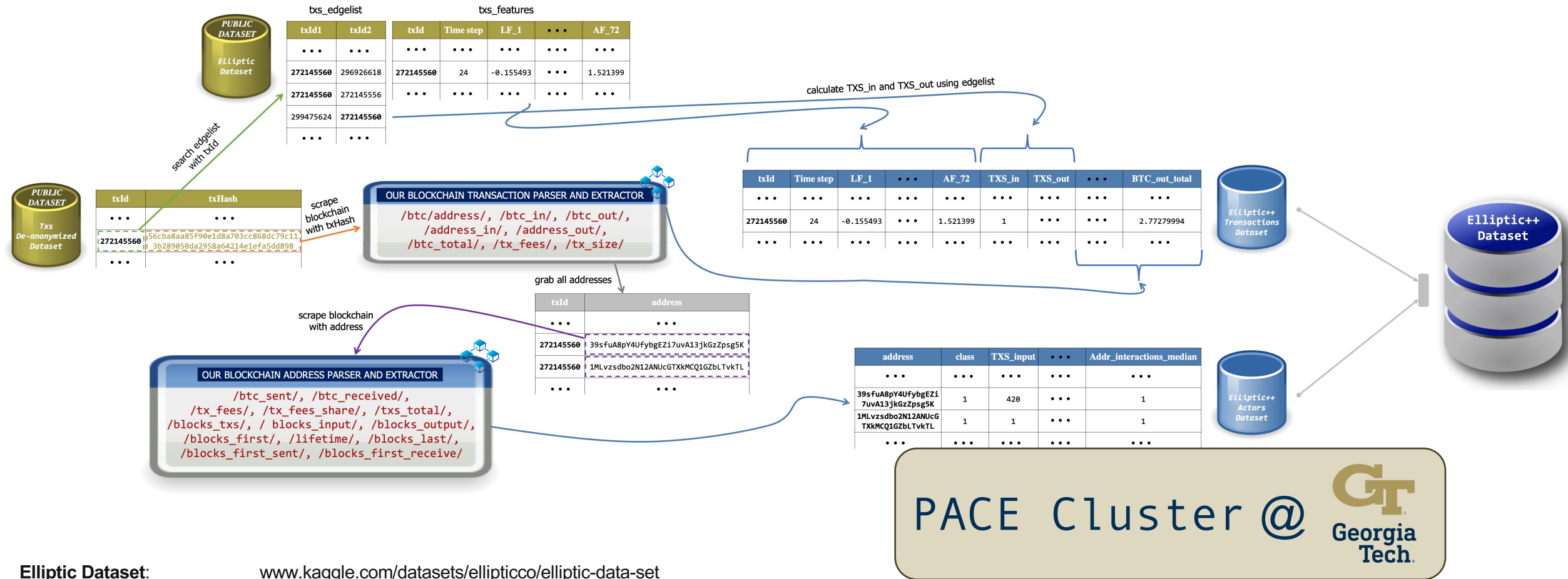
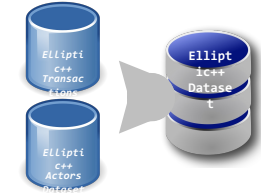
Dataset: <https://www.github.com/git-disl/EllipticPlusPlus>

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



The Elliptic++ Dataset: Collection Pipeline

Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



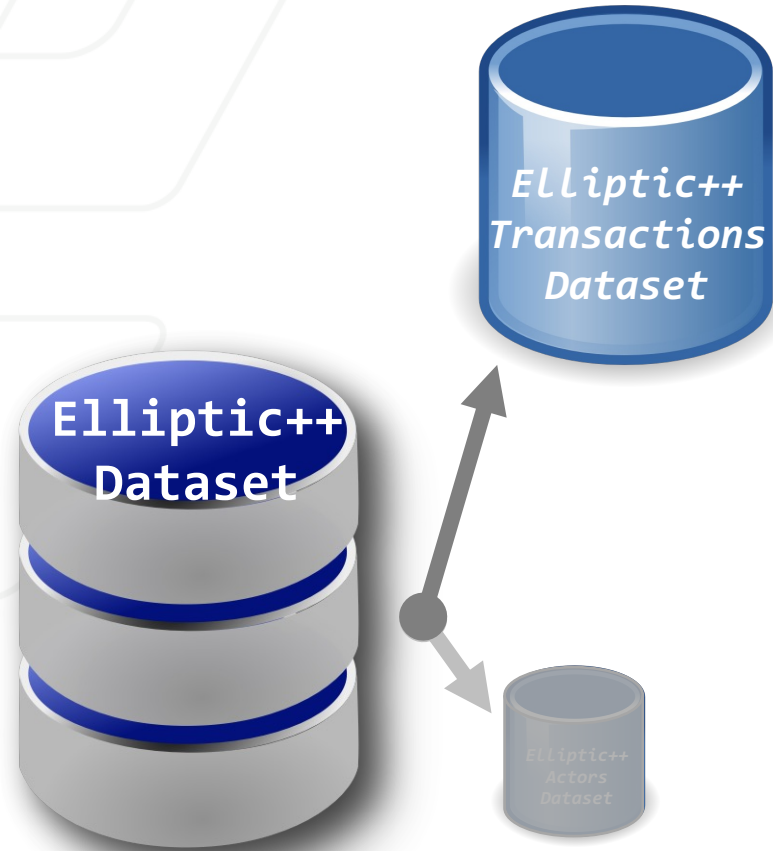
Elliptic Dataset: www.kaggle.com/datasets/elliptico/elliptic-data-set
 TxS De-anonymized Dataset: www.kaggle.com/datasets/alexbenzik/deanonymized-995-pct-of-elliptic-transactions
 Bitcoin Blockchain: www.blockchain.com

Dataset: <https://www.github.com/git-disl/EllipticPlusPlus>

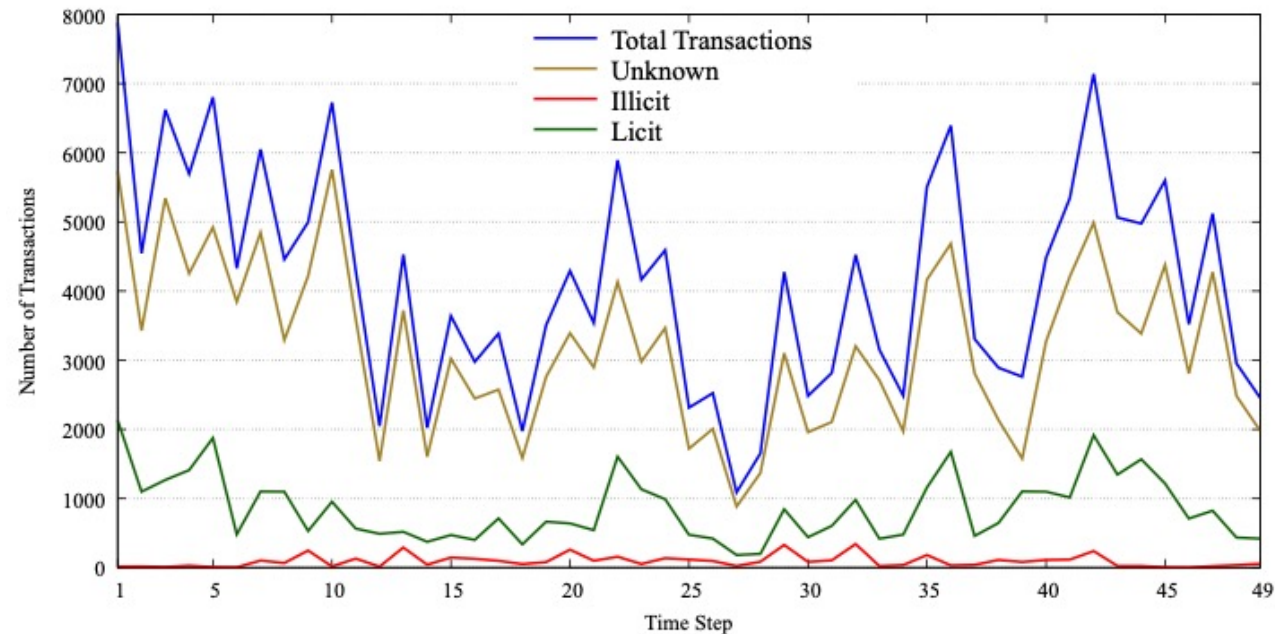
Elliptic++ Dataset: A Graph Network of Bitcoin Blockchain Transactions and Wallet Addresses



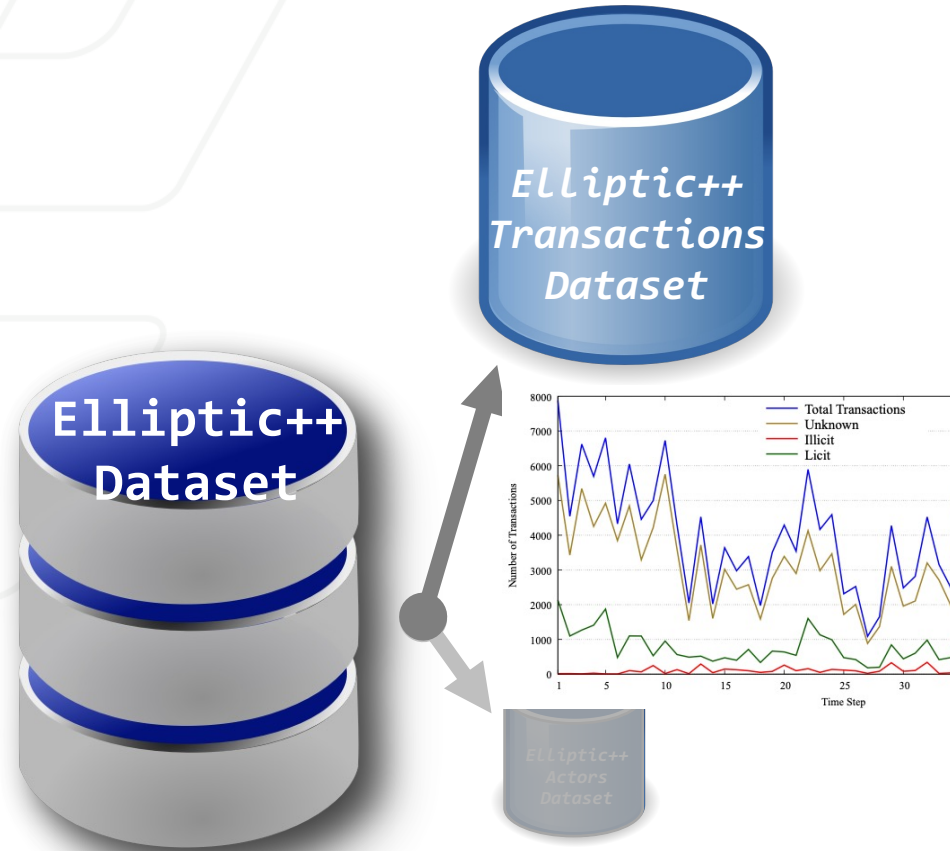
The Elliptic++ Dataset: Transactions Dataset



- 49 time steps
- 203,769 transactions
- 234,355 payment flows
- 183 transaction features
- Labels:
 - 2% **illicit**
 - 21% **licit**
 - 77% **unknown**



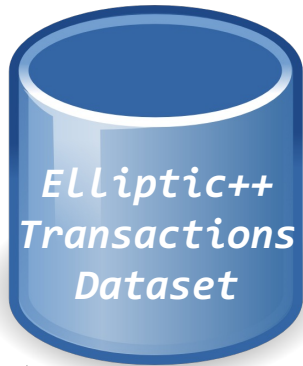
The Elliptic++ Dataset: Transactions Dataset



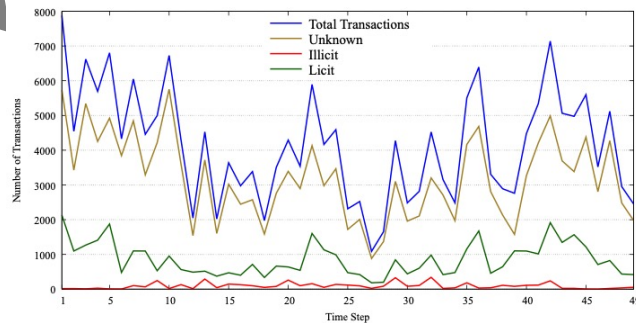
- 49 time steps
- 203,769 transactions
- 234,355 payment flows
- 183 transaction features
- Labels:
 - 2% illicit
 - 21% licit
 - 77% unknown

Feature	Description
BTC_{in}	Total BTC incoming
BTC_{out}	Total BTC outgoing
each has 5 values: total, min, max, mean, median	
Txs_{in}	Number of incoming transactions
Txs_{out}	Number of outgoing transactions
$Addr_{in}$	Number of input addresses
$Addr_{out}$	Number of output addresses
BTC_{total}	Total BTC transacted
$Fees$	Total fees in BTC
$Size$	Total transaction size
single value	

The Elliptic++ Dataset: Transactions Dataset



- 49 time steps
- 203,769 transactions
- 234,355 payment flows
- 183 transaction features
- Labels:
 - 2% illicit
 - 21% licit
 - 77% unknown



Feature	Description
<i>BTC_{in}</i>	Total BTC incoming
<i>BTC_{out}</i>	Total BTC outgoing
each has 5 values:	total, min, max, mean, median
<i>Txs_{in}</i>	Number of incoming transactions
<i>Txs_{out}</i>	Number of outgoing transactions
<i>Addr_{in}</i>	Number of input addresses
<i>Addr_{out}</i>	Number of output addresses
<i>BTC_{total}</i>	Total BTC transacted
<i>Fees</i>	Total fees in BTC
<i>Size</i>	Total transaction size
	single value

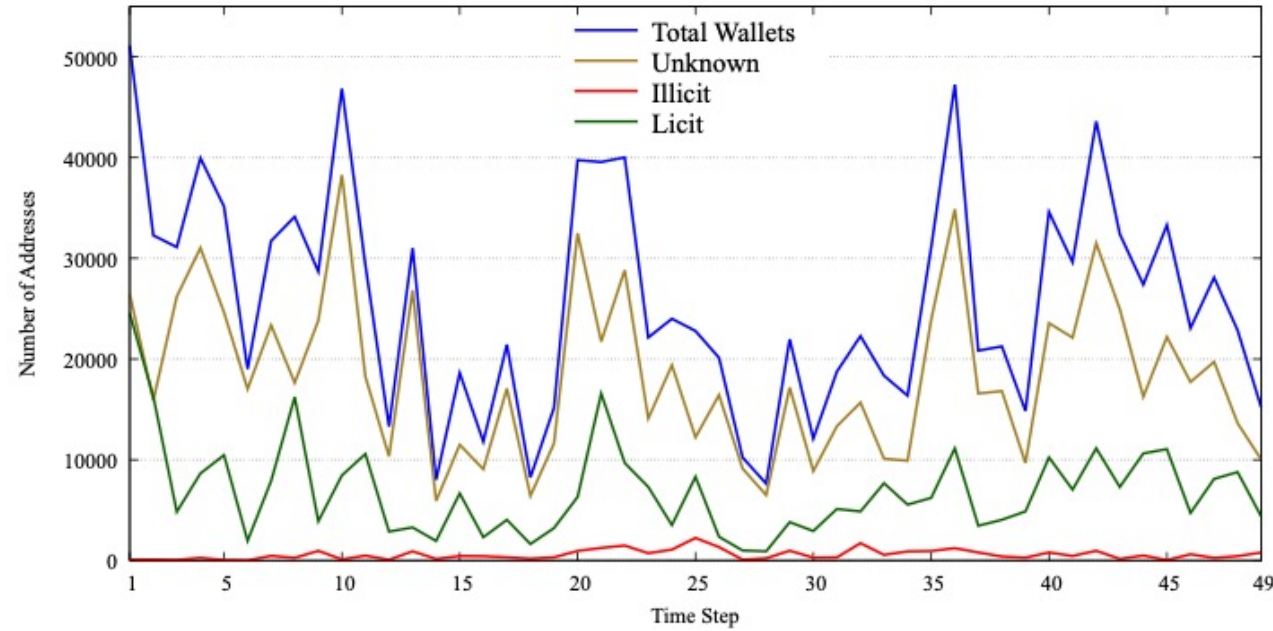
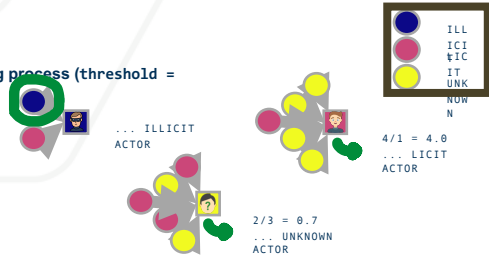
txId	Time step	LF_1	...	LF_93	AF_1	...	AF_72	TXS_in	...	BTC_out_total
272145560	24	-0.155493	...	1.135279	-0.159681	...	1.521399	1	...	2.77279994

txId1	txId2
272145560	296926618
272145560	272145556
299475624	272145560

txId	class
272145560	1

The Elliptic++ Dataset: Actors Dataset

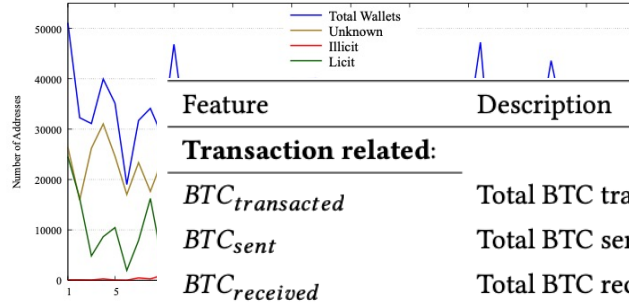
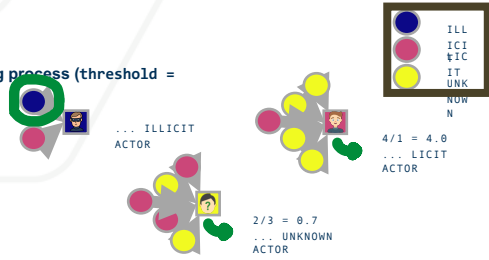
Labelling process (threshold = 3.7):



- 822,942 wallet addresses
- 1,268,260 temporal occurrences
- 56 address features
- Labels: 2% illicit, 31% licit, 67% unknown

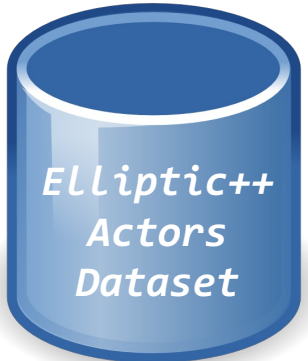
The Elliptic++ Dataset: Actors Dataset

Labelling process (threshold = 3.7):



Feature	Description
Transaction related:	
$BTC_{transacted}$	Total BTC transacted (sent+received)
BTC_{sent}	Total BTC sent
$BTC_{received}$	Total BTC received
$Fees$	Total fees in BTC
$Fees_{share}$	Total fees as share of BTC transacted
Time related:	
$Blocks_{stxs}$	Number of blocks between transactions
$Blocks_{input}$	Number of blocks between being an input address
$Blocks_{output}$	Number of blocks between being an output address
$Addr_{interactions}$	Number of interactions among addresses
each has 5 values: total, min, max, mean, median	

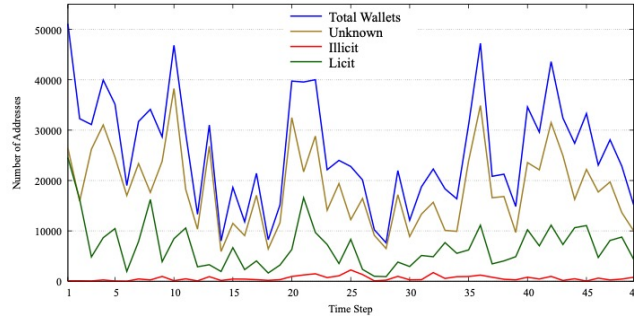
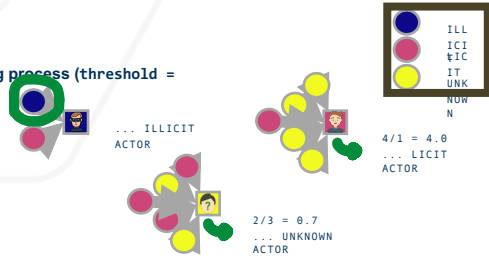
Class	class label: {illicit, licit, unknown}
Transaction related:	
Txs_{total}	Total number of blockchain transactions
Txs_{input}	Total number of dataset transactions as input address
Txs_{output}	Total number of dataset transactions as output address
Time related:	
$Timesteps$	Number of time steps transacting in
$Lifetime$	Lifetime in blocks
$Block_{first}$	Block height first transacted in
$Block_{last}$	Block height last transacted in
$Block_{first\ sent}$	Block height first sent in
$Block_{first\ receive}$	Block height first received in
$Repeat\ interactions$	Number of addresses transacted with multiple times
single value	



- 822,942 wallet addresses
- 1,268,260 temporal occurrences
- 56 address features
- Labels: 2% illicit, 31% licit, 67% unknown

The Elliptic++ Dataset: Actors Dataset

Labelling process (threshold = 3.7):



Feature	Description	Class	Description
Transaction related:		Transaction related:	
<i>BTC_{transacted}</i>	Total BTC transacted (sent+received)	<i>Txs_{total}</i>	Total number of blockchain transactions
<i>BTC_{sent}</i>	Total BTC sent	<i>Txs_{input}</i>	Total number of dataset transactions as input address
<i>BTC_{received}</i>	Total BTC received	<i>Txs_{output}</i>	Total number of dataset transactions as output address
<i>Fees</i>	Total fees in BTC	Time related:	
<i>Fees_{share}</i>	Total fees as share of BTC transacted	<i>Timesteps</i>	Number of time steps transacting in
Time related:		<i>Lifetime</i>	Lifetime in blocks
<i>Blocks_{txs}</i>	Number of blocks between transactions	<i>Block_{first}</i>	Block height first transacted in
<i>Blocks_{input}</i>	Number of blocks between being an input address	<i>Block_{last}</i>	Block height last transacted in
<i>Blocks_{output}</i>	Number of blocks between being an output address	<i>Block_{first sent}</i>	Block height first sent in
<i>Addr interactions</i>	Number of interactions among addresses	<i>Block_{first receive}</i>	Block height first received in
each has 5 values: total, min, max, mean, median		<i>Repeat interactions</i>	Number of addresses transacted with multiple times
		single value	

Elliptic++ Dataset



wallets_features.csv						
address	time step	txs_input	...	lifetime_blocks	...	Addr_interactions_median
39sfuA8pY4UfybgEzi7uvA13jkGzZpsg5K	23	420	...	18145	...	1

AddrAddr_edgelist.csv	
input_address	output_address
39sfuA8pY4UfybgEzi7uvA13jkGzZpsg5K	1ML...kTL

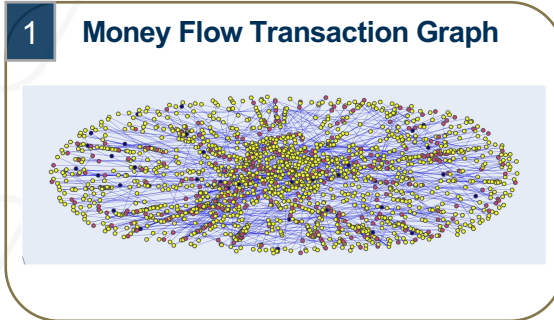
AddrTx_edgelist.csv	
input_address	txId
39sfuA8pY4UfybgEzi7uvA13jkGzZpsg5K	272145560

TxAddr_edgelist.csv	
txId	output_address
322554634	39sfuA8pY4UfybgEzi7uvA13jkGzZpsg5K

- 822,942 wallet addresses
- 1,268,260 temporal occurrences
- 56 address features
- Labels: 2% illicit, 31% licit, 67% unknown

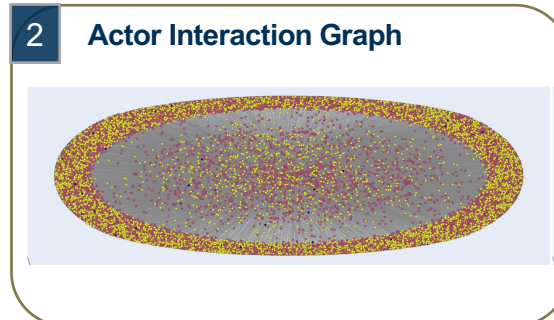
wallets_classes.csv	
address	class
39sf...sg5K	1

Graph Visualization of The Elliptic++ Dataset



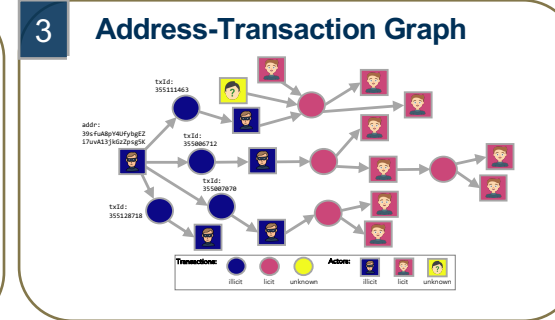
BTC flow from one transaction to the next

Utility: Exploration of spatial and temporal patterns surrounding a transaction



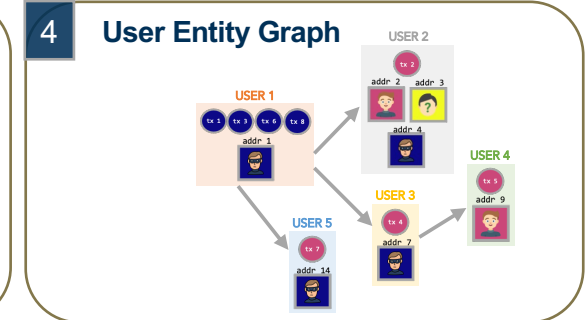
Pairwise interactions among input and output addresses of txs

Utility: Shows density of k-hop neighborhoods of addresses



BTC flow across txs and addresses

Utility: Evaluation of relationships among addresses of same transactions

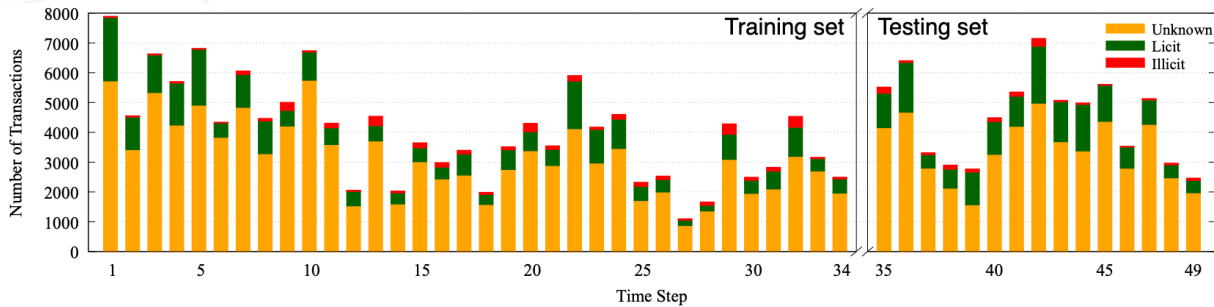


Address cluster graphs

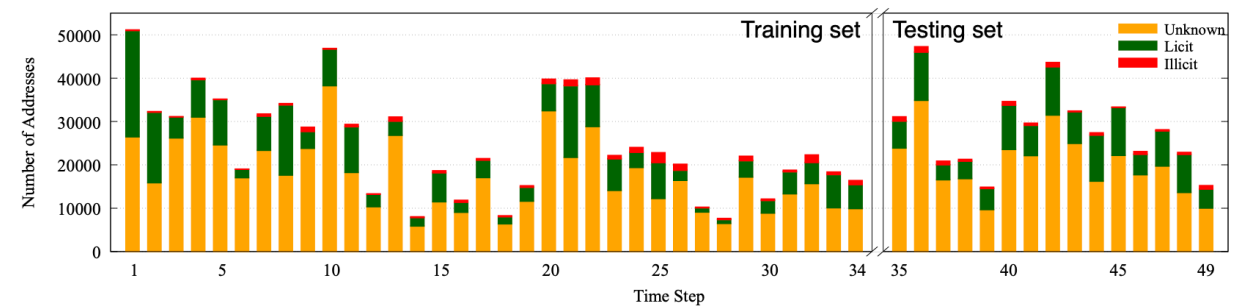
Utility: Linking of addresses controlled by a specific user (de-anonymization)

Fraud Detection Methodology

Dataset Preprocessing



TRANSACTIONS DATASET



ACTORS DATASET

Machine Learning Models

Random Forest (RF) Scikit-learn: default parameters, 50 estimators

Multilayer Perceptrons (MLP) 1 hidden layer w/ 50 neurons, 500 epochs, Adam opt., 0.001 LR

Long Short-Term Memory (LSTM) TensorFlow: sigmoid act., Adam opt., 30 epochs, binary cross-entropy loss, 15 emb. output dims

Extreme Gradient Boosting (XGB) XGBoost: default parameters, multi:softmax obj., 2 classes

Logistic Regression (LR) Scikit-learn: default parameters

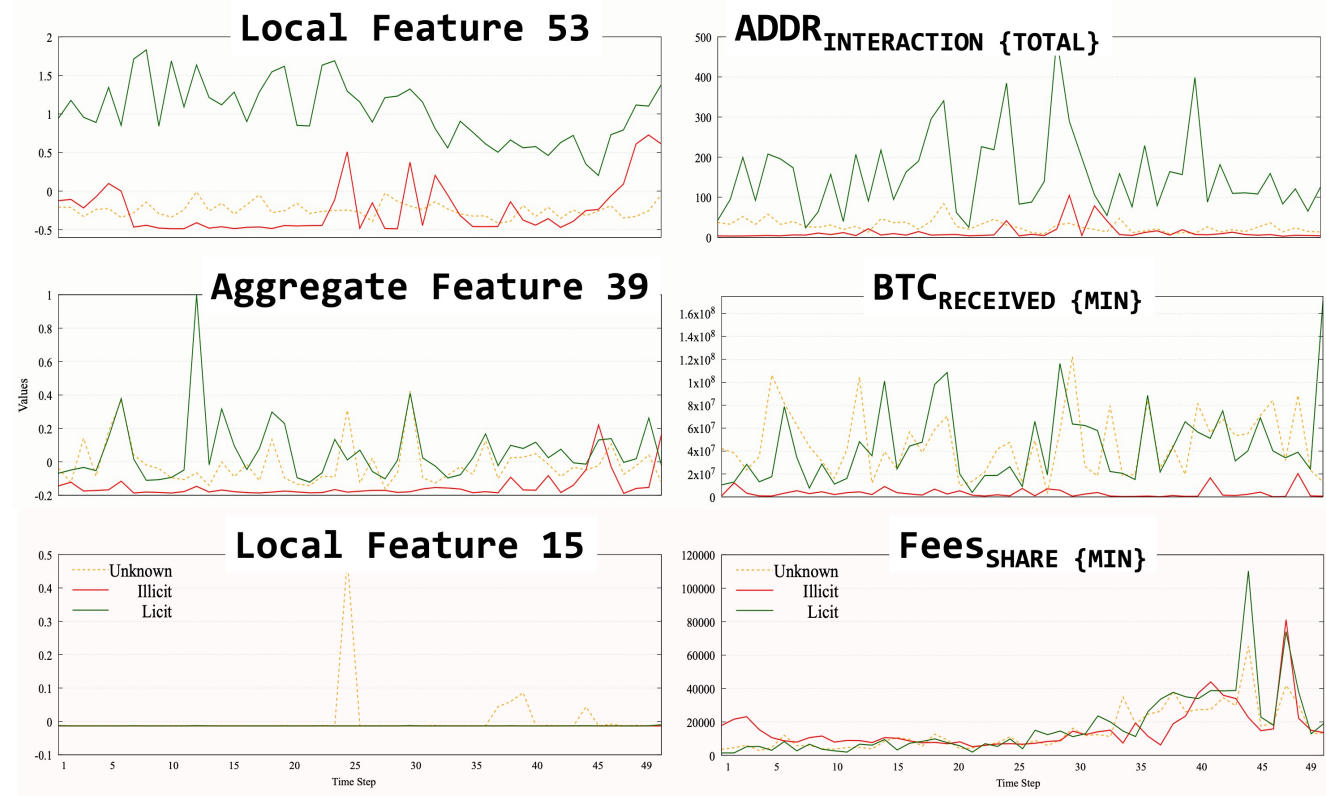
Evaluation Metrics

Precision, Recall, F1 Score, Micro-Avg F1 Score, Matthews Correlation Coefficient (MCC)

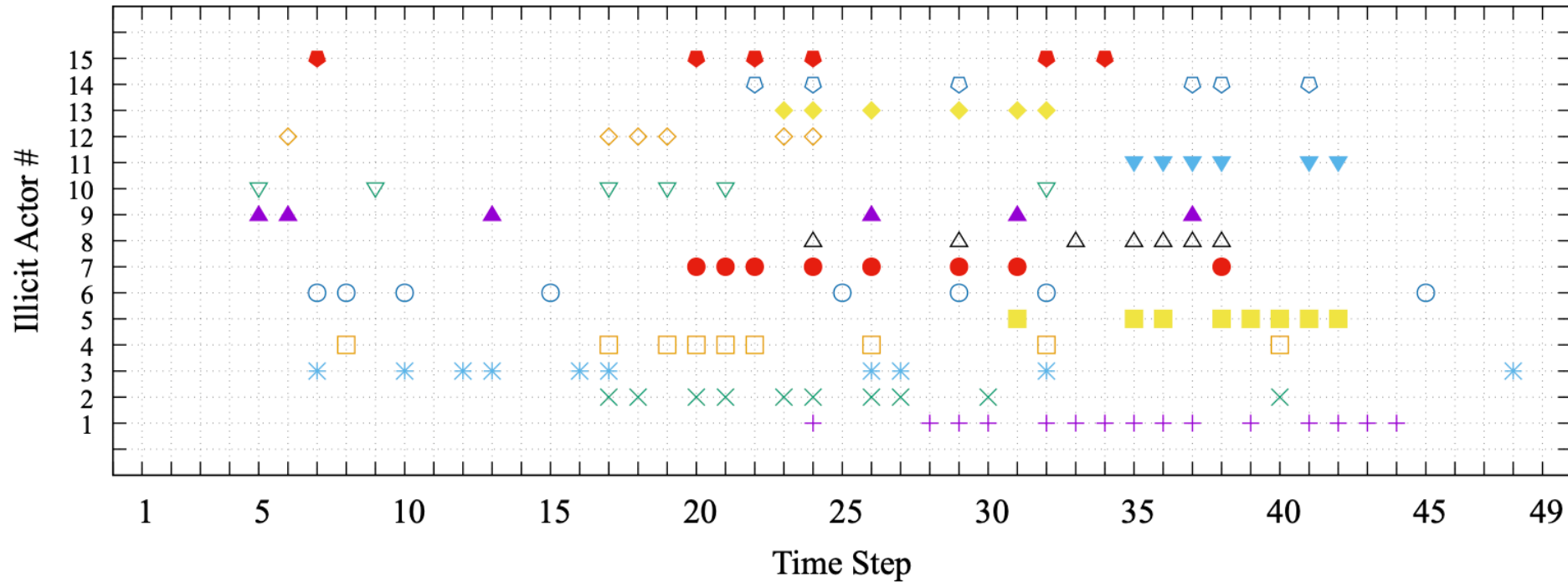
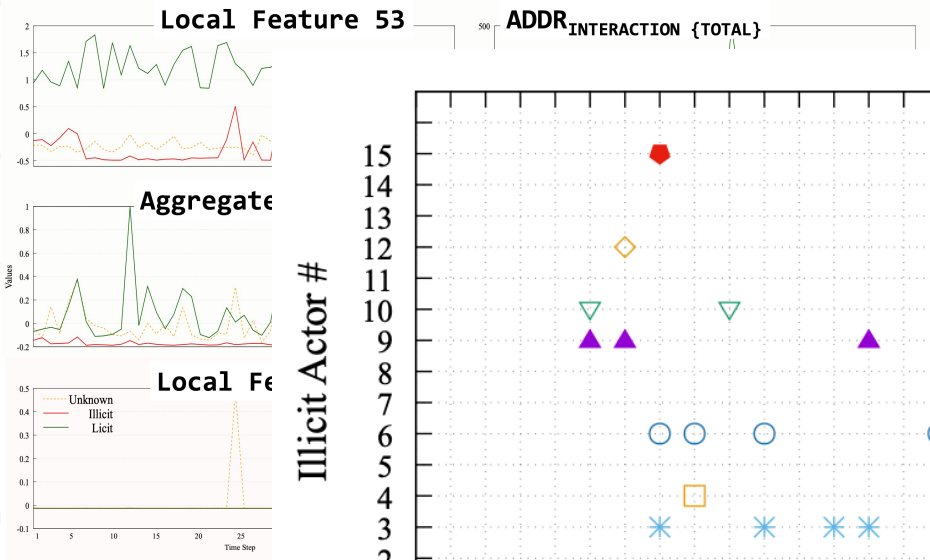
Case studies:

- (i) **EASY cases**: all models classify an illicit tx correctly
- (ii) **HARD cases**: all models classify an illicit tx incorrectly
- (iii) **AVERAGE cases**: some models failed to classify an illicit tx but ≥ 1 models classified correctly

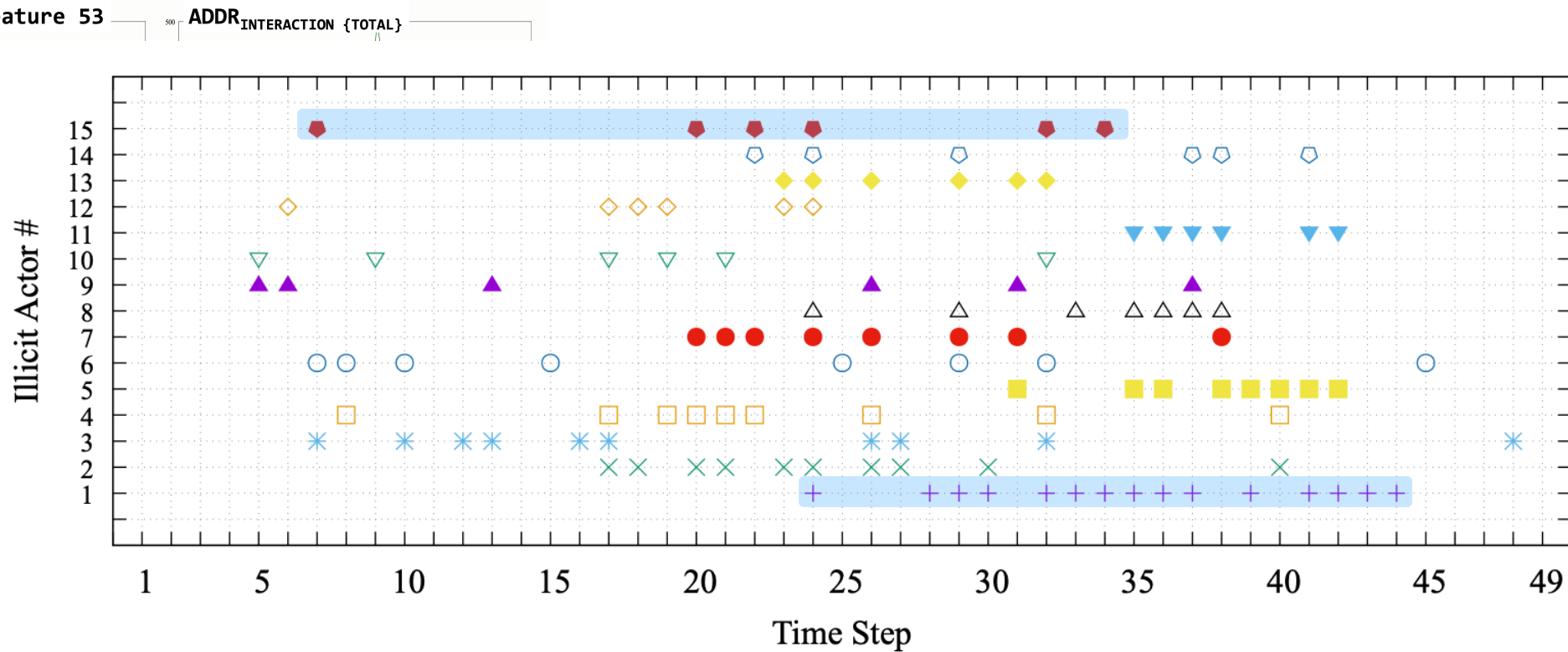
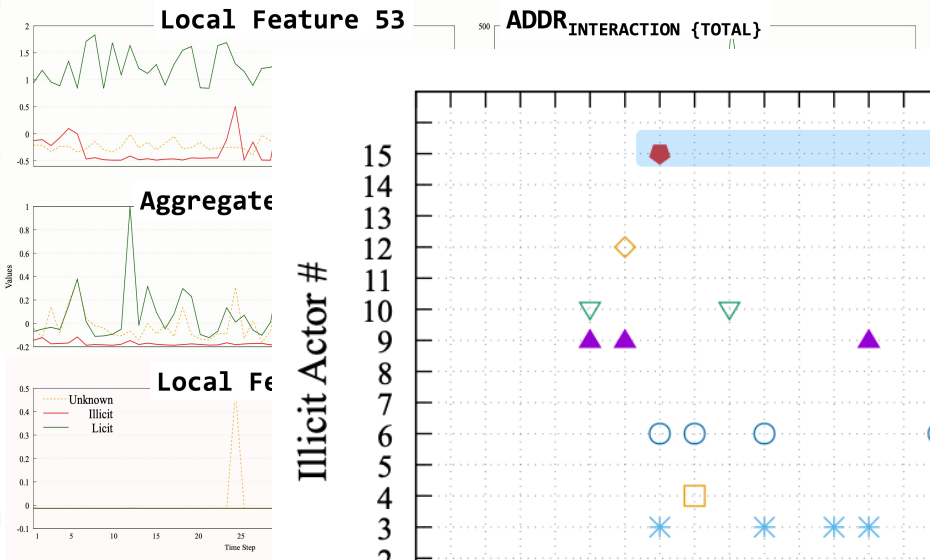
Results: Statistical Analysis of the Dataset



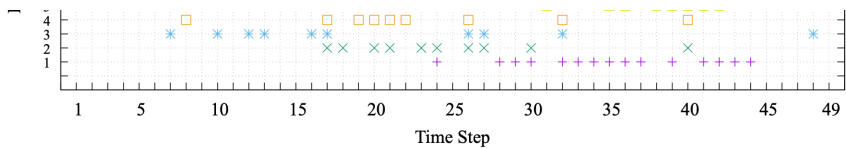
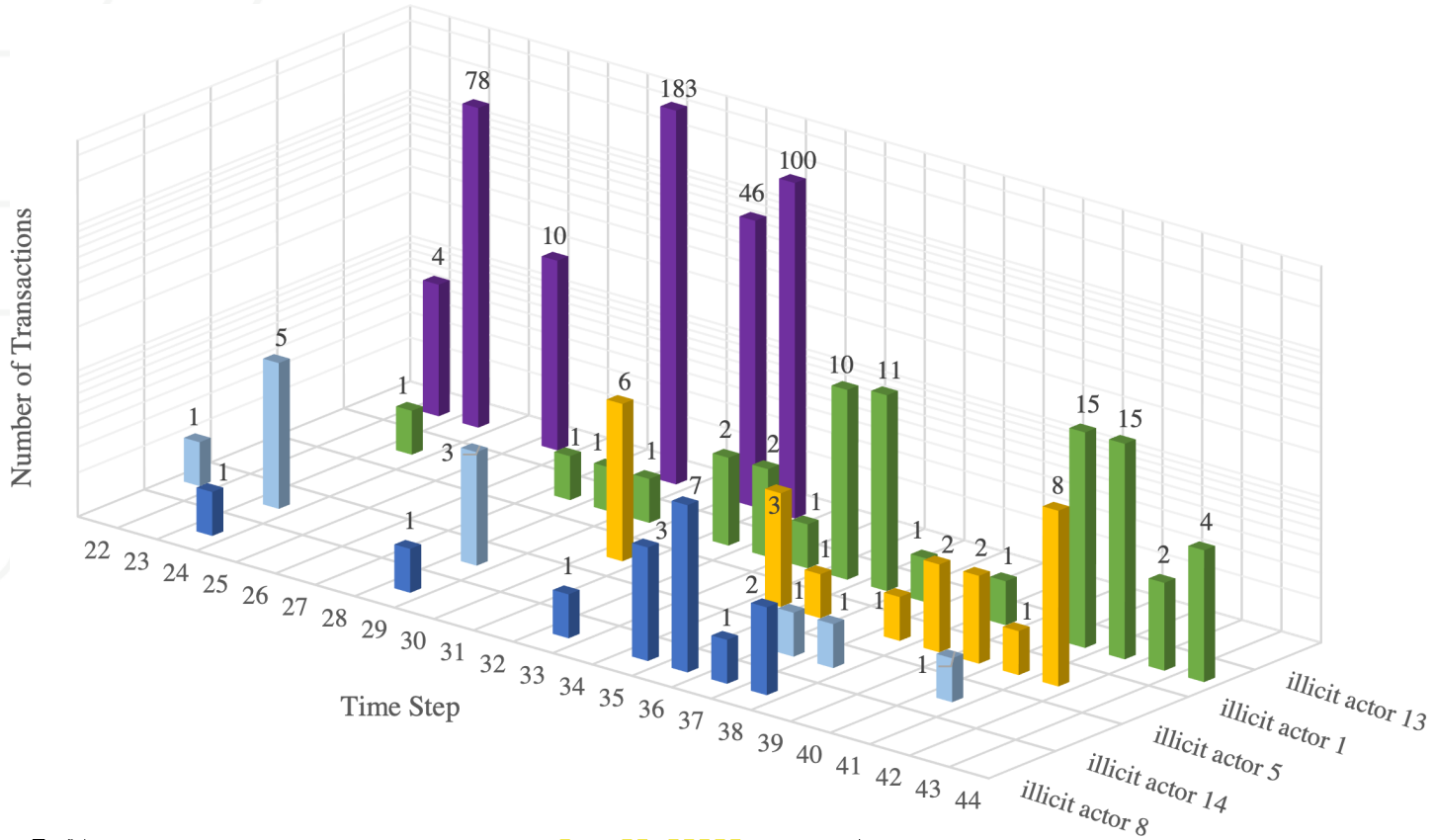
Results: Statistical Analysis of the Dataset



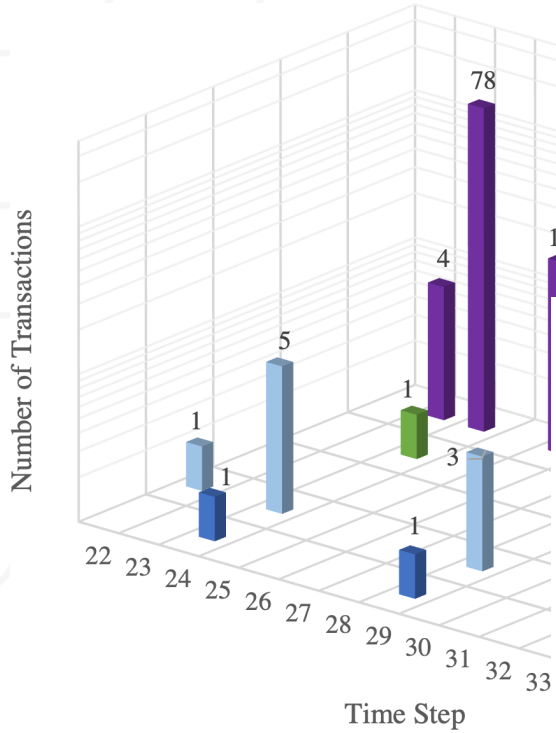
Results: Statistical Analysis of the Dataset



Results: Statistical Analysis of the Dataset



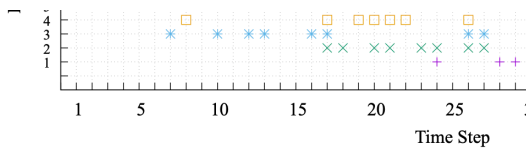
Results: Statistical Analysis of the Dataset



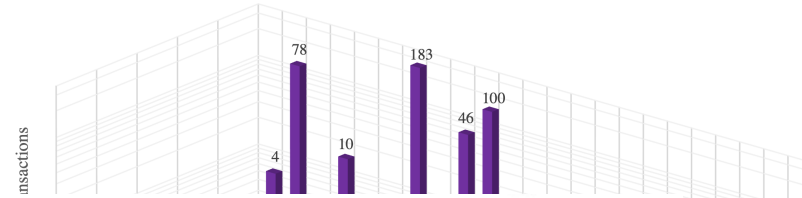
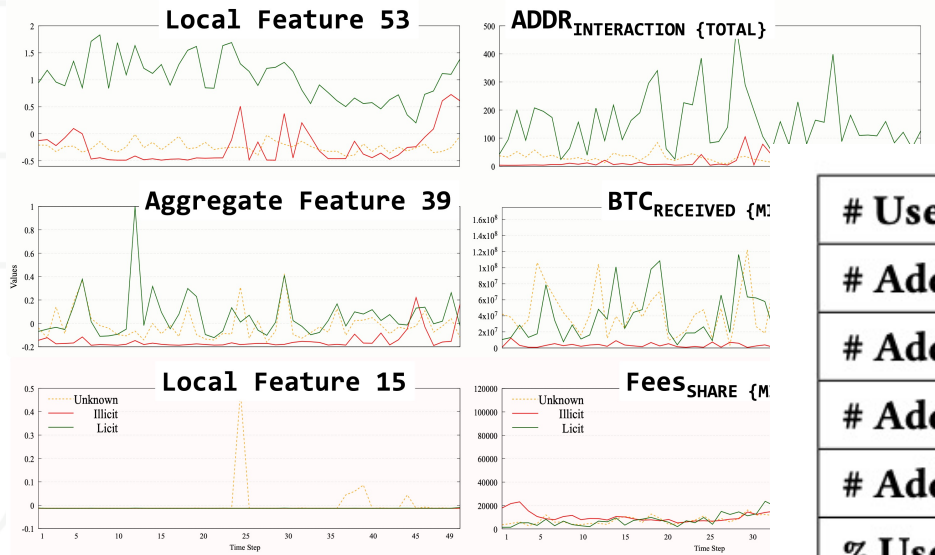
Time step	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17
illicit actor appears in 1 time step	68	69	42	186	36	8	233	116	384	26	185	47	342	74	193	174	118
illicit actor appears in 2 – 4 time steps	5	1	3	15	2	1	28	7	21	11	7	2	4	0	6	8	8
illicit actor appears in ≥ 5 time steps	0	0	0	0	2	2	5	4	3	3	2	1	2	0	2	1	6

Time step	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
illicit actor appears in 1 time step	77	122	405	612	692	376	532	1096	704	43	97	372	130	131	759	270	447
illicit actor appears in 2 – 4 time steps	4	5	20	16	18	17	40	9	21	1	7	52	11	24	23	7	11
illicit actor appears in ≥ 5 time steps	2	4	5	4	5	3	12	1	9	2	1	9	4	6	8	5	2

Time step	35	36	37	38	39	40	41	42	43	44	45	46	47	48	49
illicit actor appears in 1 time step	427	588	428	180	127	436	168	395	93	277	23	505	193	370	631
illicit actor appears in 2 – 4 time steps	9	5	5	9	0	14	8	12	11	6	9	4	11	17	20
illicit actor appears in ≥ 5 time steps	6	4	7	5	2	3	6	3	1	1	1	0	0	1	1



Results: Statistical Analysis of the Dataset

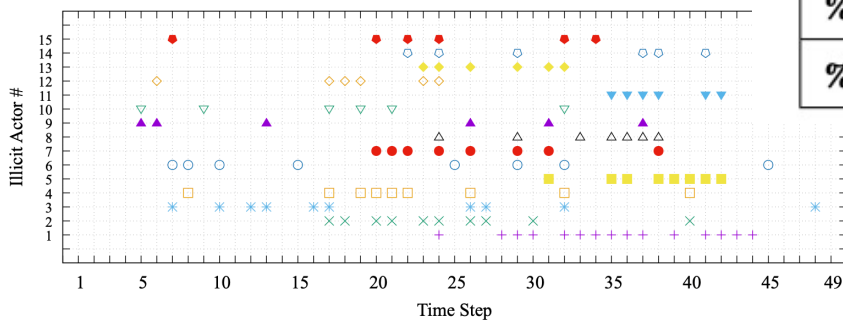


# Users	146,783
# Addresses per User: Min	1
# Addresses per User: Median	1
# Addresses per User: Mean	2.73
# Addresses per User: Max	14,885
% Users w/ 1 – 10 Addresses	98.72%
% Users w/ 11 – 1K Addresses	1.26%
% Users w/ 1K – max Addresses	0.02%

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17
68	69	42	186	36	8	233	116	384	26	185	47	342	74	193	174	118
5	1	3	15	2	1	28	7	21	11	7	2	4	0	6	8	8
0	0	0	0	2	2	5	4	3	3	2	1	2	0	2	1	6

8	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
7	122	405	612	692	376	532	1096	704	43	97	372	130	131	759	270	447
1	5	20	16	18	17	40	9	21	1	7	52	11	24	23	7	11
2	4	5	4	5	3	12	1	9	2	1	9	4	6	8	5	2

	35	36	37	38	39	40	41	42	43	44	45	46	47	48	49
	427	588	428	180	127	436	168	395	93	277	23	505	193	370	631
98	9	5	5	9	0	14	8	12	11	6	9	4	11	17	20
s	6	4	7	5	2	3	6	3	1	1	1	0	0	1	1



Results: Model Evaluation and Analysis

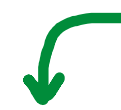
TX - Elliptic++ Transactions Dataset
 EC - Elliptic Dataset
 AR - Elliptic++ Actors Dataset

TRANSACTIONS
 DATASET



Model	Precision	Recall	F1 Score	Micro-F1
LR ^{EC}	0.326	0.707	0.446	0.886
LR ^{TX}	0.328	0.707	0.448	0.884
RF ^{EC}	0.940	0.724	0.818	0.979
RF ^{TX}	0.975	0.719	0.828	0.980
MLP ^{EC}	0.476	0.673	0.558	0.931
MLP ^{TX}	0.611	0.613	0.612	0.949
LSTM ^{EC}	0.665	0.350	0.459	0.946
LSTM ^{TX}	0.709	0.223	0.339	0.942
XGB ^{EC}	0.812	0.717	0.761	0.971
XGB ^{TX}	0.793	0.718	0.754	0.969
<i>2 classifiers ensemble, selecting top 3 classifiers</i>				
RF+MLP ^{EC}	0.987	0.624	0.765	0.975
RF+MLP ^{TX}	0.989	0.635	0.773	0.975
RF+XGB ^{EC}	0.960	0.704	0.812	0.979
RF+XGB ^{TX}	0.977	0.706	0.820	0.979
MLP+XGB ^{EC}	0.457	0.737	0.564	0.926
MLP+XGB ^{TX}	0.974	0.596	0.739	0.972
<i>3 classifiers ensemble, selecting top 3 classifiers</i>				
RF+MLP+XGB ^{EC}	0.947	0.719	0.817	0.979
RF+MLP+XGB ^{TX}	0.962	0.723	0.826	0.980

ACTORS DATASET



Model	Precision	Recall	F1 Score	Micro-F1
LR ^{AR}	0.477	0.046	0.083	0.964
RF ^{AR}	0.911	0.789	0.845	0.990
MLP ^{AR}	0.708	0.502	0.587	0.974
LSTM ^{AR}	0.922	0.033	0.064	0.965
XGB ^{AR}	0.869	0.534	0.662	0.980
<i>2 classifiers ensemble, selecting top 3 classifiers</i>				
RF+MLP ^{AR}	0.967	0.403	0.568	0.978
RF+XGB ^{AR}	0.959	0.530	0.682	0.982
MLP+XGB ^{AR}	0.929	0.324	0.481	0.975
<i>3 classifiers ensemble, selecting top 3 classifiers</i>				
RF+MLP+XGB ^{AR}	0.933	0.572	0.709	0.983

Results: EASY, HARD, AVERAGE Cases Analysis

Time Step		35	36	37	38	39	40	41	42	43	44	45	46	47	48	49	TOTAL
EASY		32	0	2	5	5	1	1	3	0	0	0	0	0	0	0	49
HARD		4	0	10	7	4	28	6	36	22	20	4	1	21	27	53	243
AVERAGE	LR	0	0	3	0	2	3	0	6	2	3	1	0	1	9	2	791
	RF	0	0	1	0	0	0	0	1	0	0	0	0	0	0	0	
	MLP	0	0	1	1	0	2	0	2	0	0	0	0	0	0	0	
	LSTM	1	0	0	1	0	1	0	0	0	0	0	0	0	0	0	
	XGB	0	0	0	0	0	0	1	2	0	0	0	0	0	0	0	
	RF,XGB	4	0	0	1	2	1	17	2	0	0	0	0	0	0	0	
	LR,MLP	1	0	0	1	0	2	0	2	0	0	0	0	0	0	0	
	RF,MLP,XGB	5	6	0	8	3	4	1	0	0	0	0	0	0	0	0	
	LR,RF,XGB	6	1	10	27	18	10	5	21	0	0	0	0	0	0	0	
RF,MLP,XGB,LR	124	24	12	57	45	55	81	159	0	1	0	1	0	0	0	71%	

This motivates us to focus on optimization of the RF model with feature refinement

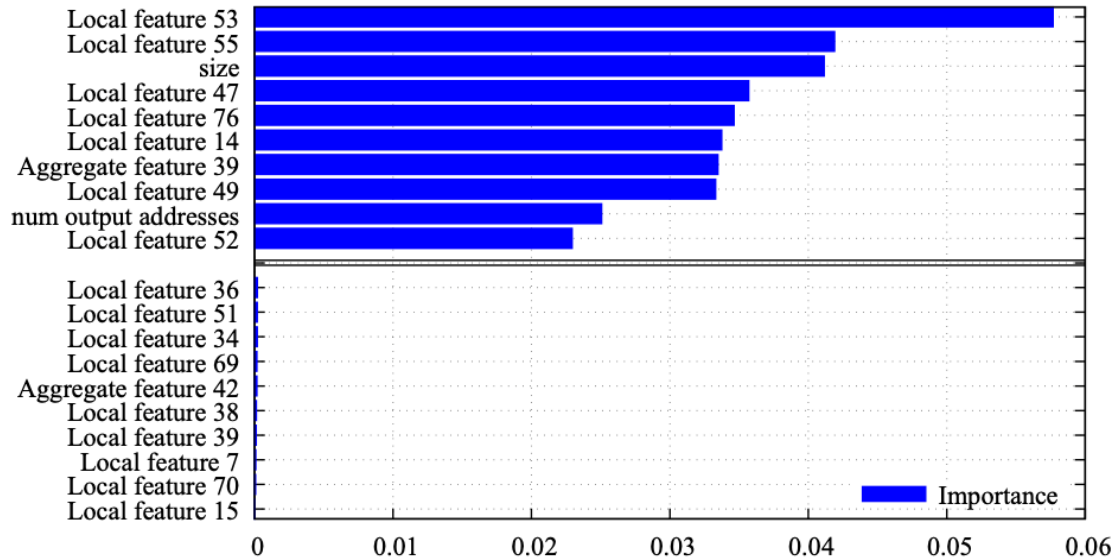
Model Optimization by Feature Refinement

TRANSACTIONS DATASET

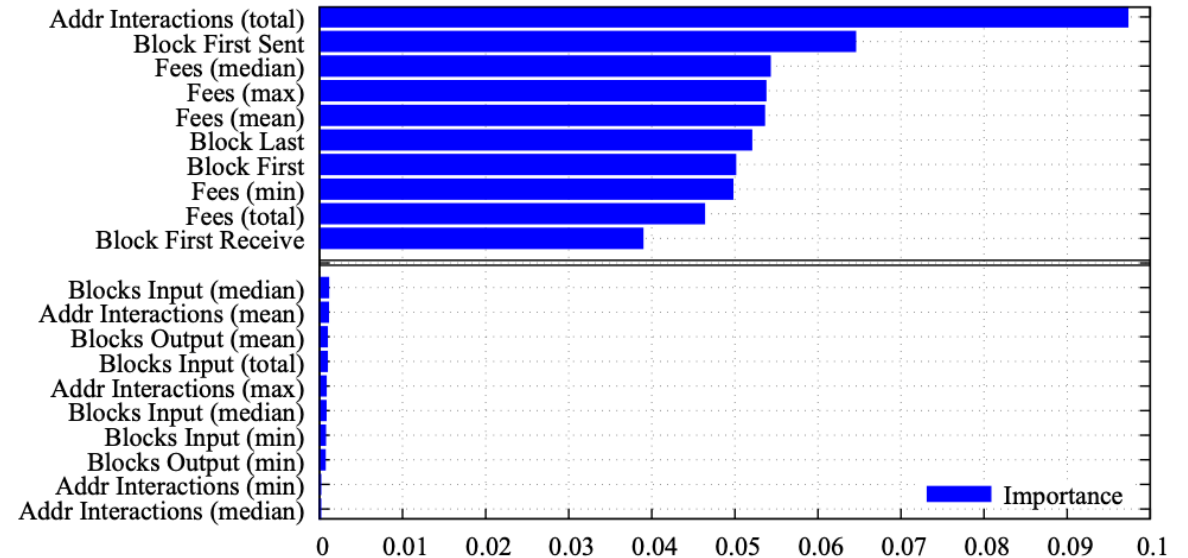
ACTORS DATASET

Model Optimization by Feature Refinement

TRANSACTIONS DATASET

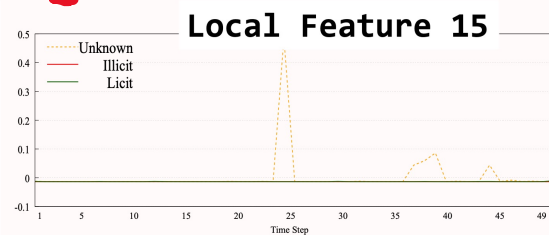
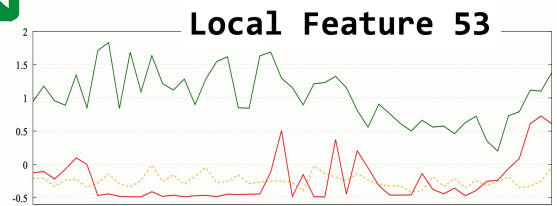
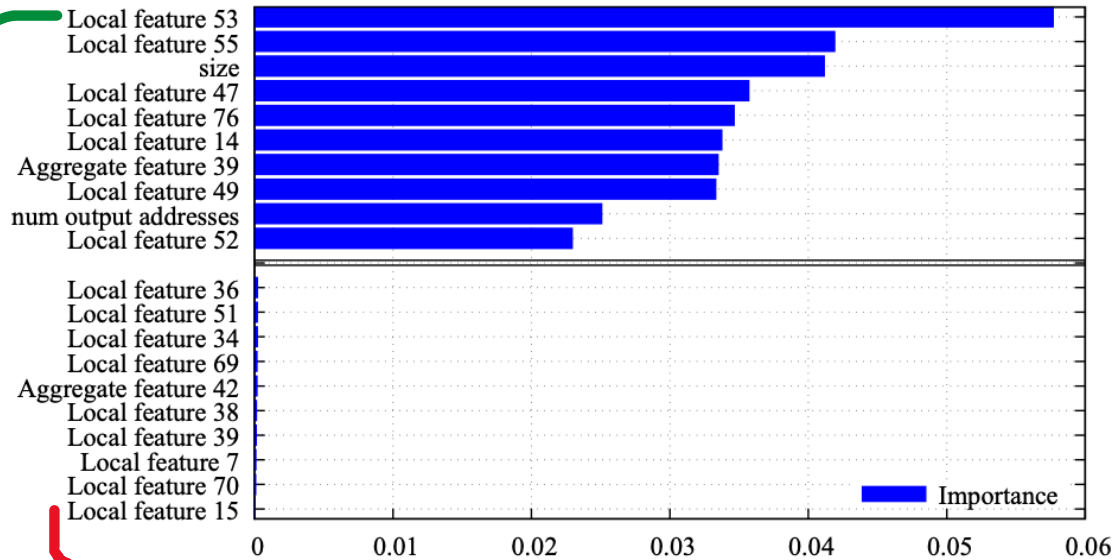


ACTORS DATASET

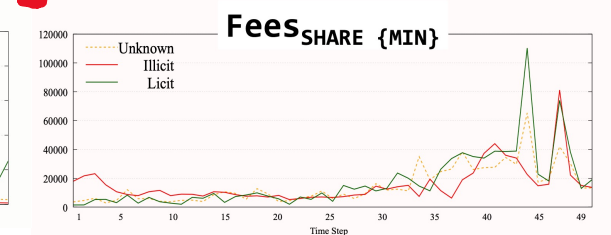
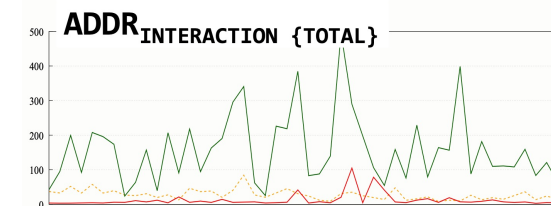
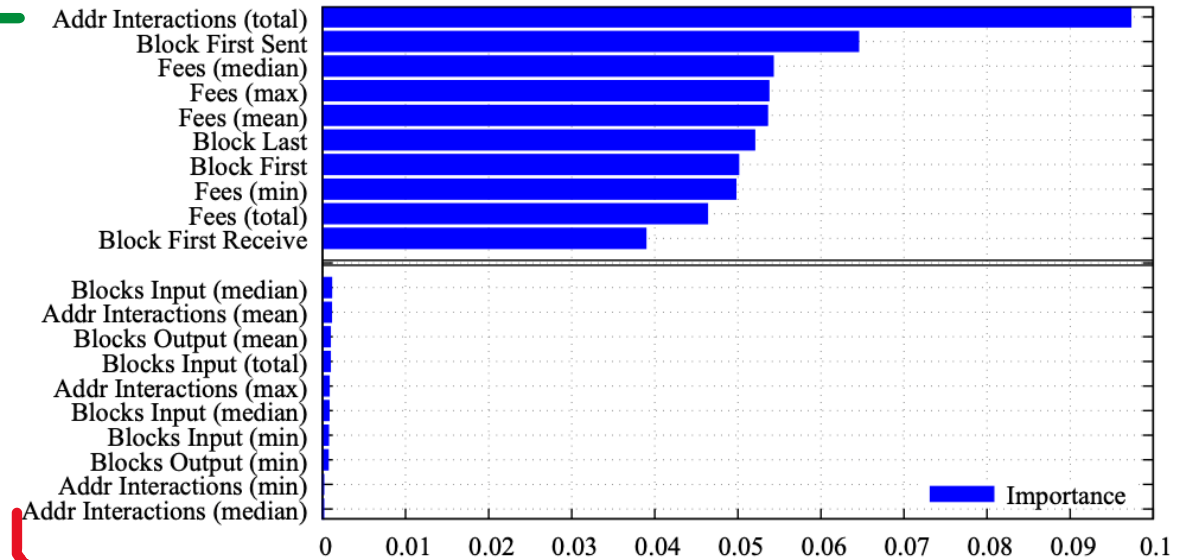


Model Optimization by Feature Refinement

TRANSACTIONS DATASET

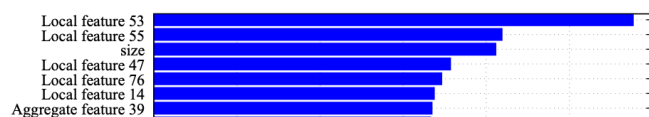


ACTORS DATASET



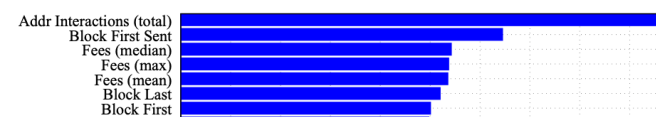
Model Optimization by Feature Refinement

TRANSACTIONS DATASET



Model	Precision	Recall	F1 Score	Micro-F1
RF ^{TX}	0.975	0.719	0.828	0.980
RF ^{TX} ^ψ	0.986	0.727	0.836	0.981
RF+XGB ^{TX}	0.977	0.706	0.820	0.979
RF+XGB ^{TX} ^ψ	0.987	0.717	0.826	0.980
RF+MLP+XGB ^{TX}	0.962	0.723	0.826	0.980
RF+MLP+XGB ^{TX} ^ψ	0.968	0.729	0.834	0.980

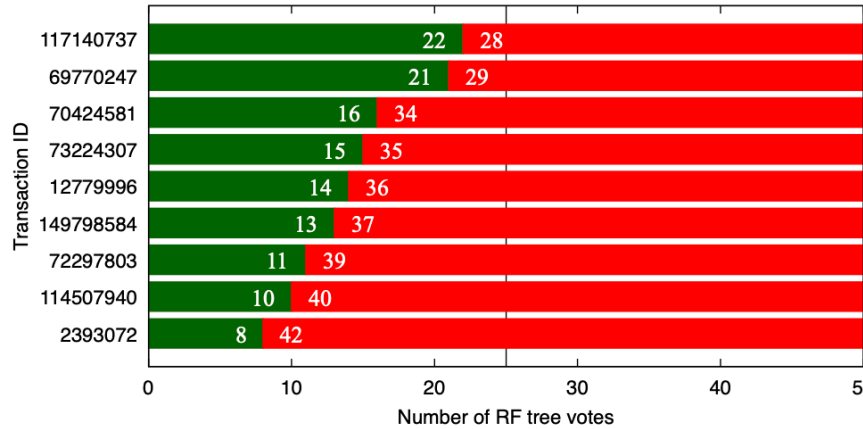
ACTORS DATASET



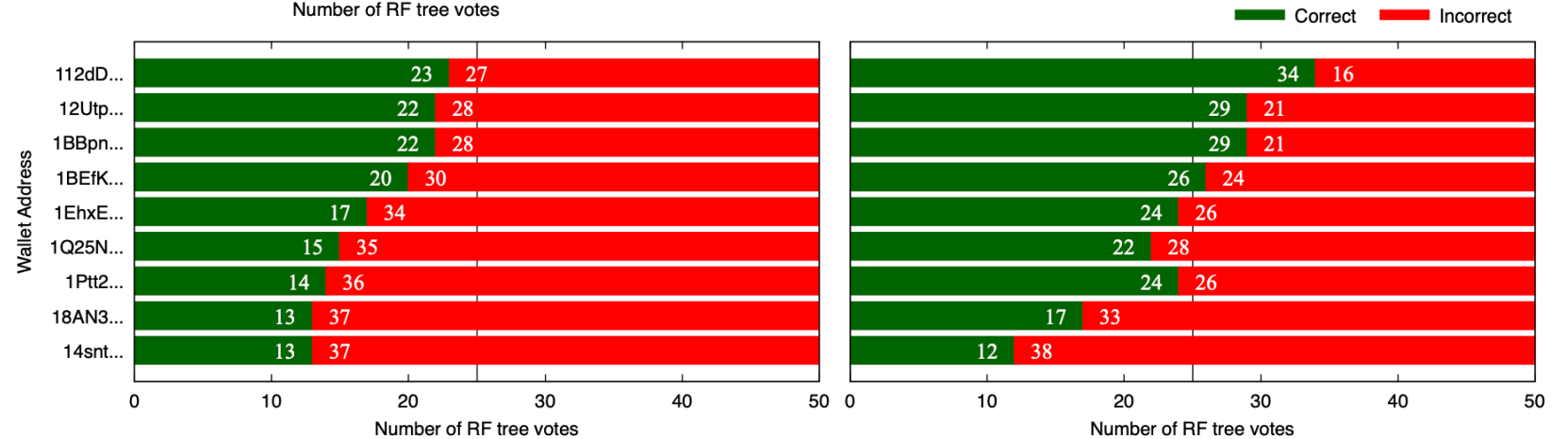
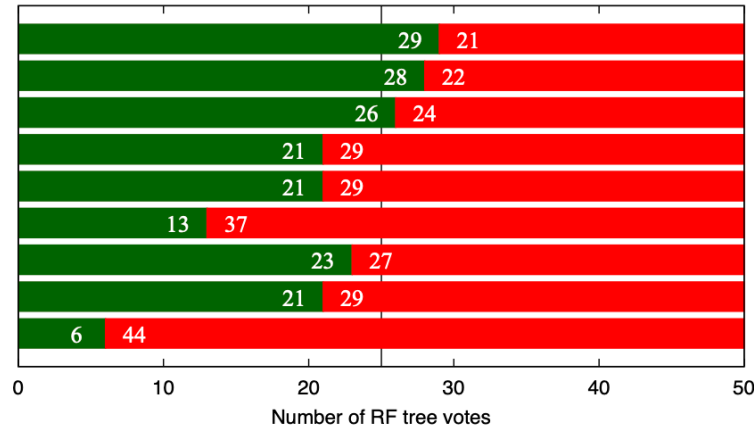
Model	Precision	Recall	F1 Score	Micro-F1
RF ^{AR}	0.911	0.789	0.845	0.990
RF ^{AR} ^ψ	0.921	0.802	0.858	0.990
RF+XGB ^{AR}	0.959	0.530	0.682	0.982
RF+XGB ^{AR} ^ψ	0.967	0.543	0.686	0.982
RF+MLP+XGB ^{AR}	0.933	0.572	0.709	0.983
RF+MLP+XGB ^{AR} ^ψ	0.945	0.601	0.718	0.984

Model Optimization by Feature Refinement

TRANSACTIONS DATASET

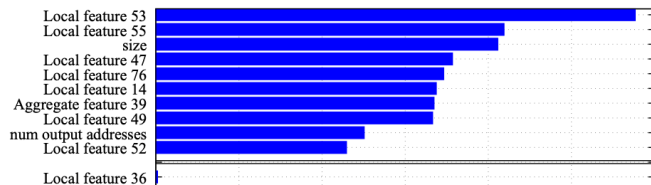


ACTORS DATASET

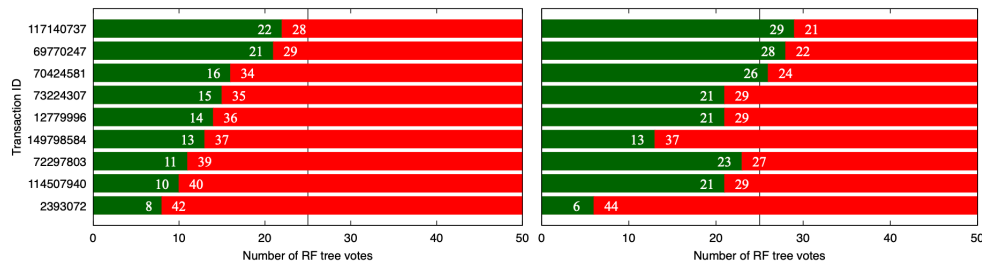


Model Optimization by Feature Refinement

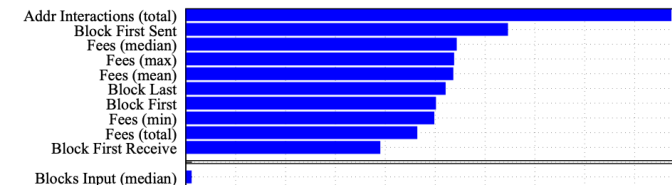
TRANSACTIONS DATASET



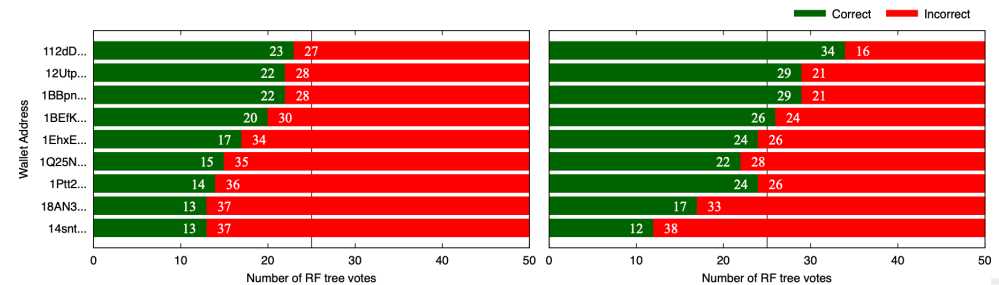
Model	Precision	Recall	F1 Score	Micro-F1
RF ^{TX}	0.975	0.719	0.828	0.980
RF ^{TX} ^ψ	0.986	0.727	0.836	0.981
RF+XGB ^{TX}	0.977	0.706	0.820	0.979
RF+XGB ^{TX} ^ψ	0.987	0.717	0.826	0.980
RF+MLP+XGB ^{TX}	0.962	0.723	0.826	0.980
RF+MLP+XGB ^{TX} ^ψ	0.968	0.729	0.834	0.980



ACTORS DATASET



Model	Precision	Recall	F1 Score	Micro-F1
RF ^{AR}	0.911	0.789	0.845	0.990
RF ^{AR} ^ψ	0.921	0.802	0.858	0.990
RF+XGB ^{AR}	0.959	0.530	0.682	0.982
RF+XGB ^{AR} ^ψ	0.967	0.543	0.686	0.982
RF+MLP+XGB ^{AR}	0.933	0.572	0.709	0.983
RF+MLP+XGB ^{AR} ^ψ	0.945	0.601	0.718	0.984



Concluding Remarks

Our Contributions to financial forensics on blockchain networks:

1

We release The Elliptic++ Dataset, combining over 203k transactions and 822k addresses, and provide four unique graph representations.

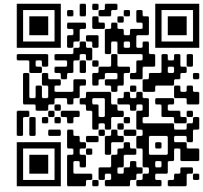
2

We demonstrate the utility of the dataset for detecting fraudulent transactions and illicit accounts using representative ML approaches.

3

We provide detailed analysis on improving the generalization performance of individual and ensemble ML algorithms on the dataset.

The Elliptic++ Dataset and Tutorials



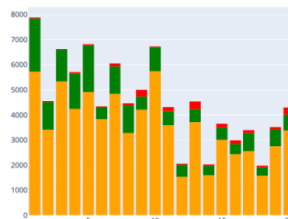
Publicly available at
github.com/git-disl/EllipticPlusPlus

Along with several
tutorials!!!

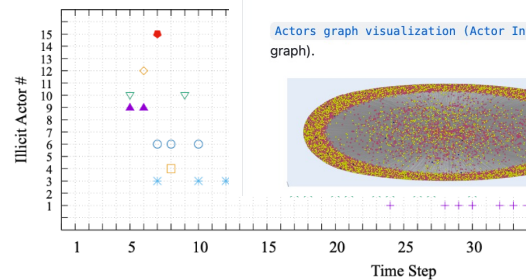
Dataset Tutorials

We are sharing tutorial notebooks for users and researchers to explore, study, and learn from. Notebooks are available for both datasets and cover dataset statistics, graph visualization, classification, case analysis, and feature refi

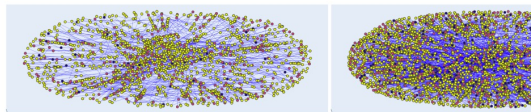
Transactions dataset statistics : overall



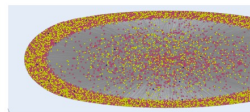
Actors dataset statistics : overall



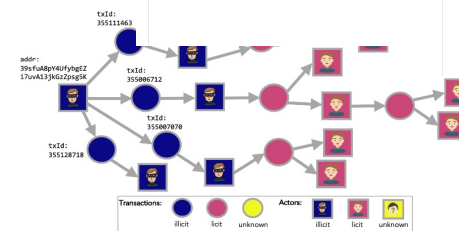
Transactions graph visualization : visualizations of the Money Flow Transaction graph



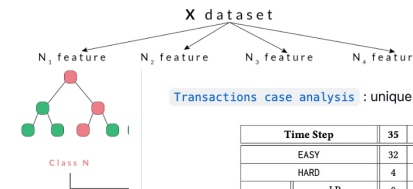
Actors graph visualization (Actor Int graph)



Actors graph visualization (Address tx-addr graph)



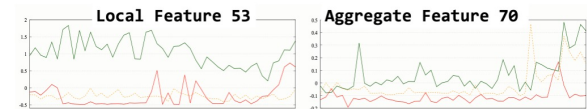
Transactions classification : model training and classification on the transactions data.



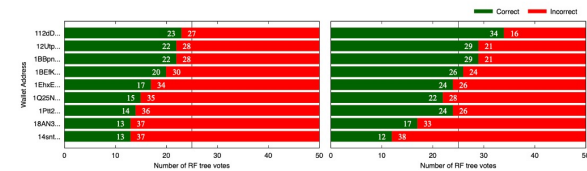
Transactions case analysis : unique case (EASY, HARD, AVERAGE) analysis using the transactions data.

Time Step	35	36	37	38	39	40	41	42	43	44	45	46	47	48	49	TOTAL
EASY	32	0	2	5	5	1	1	3	0	0	0	0	0	0	0	49
HARD	4	0	10	7	4	28	6	36	22	20	4	1	21	27	53	243
LR	0	0	3	0	2	3	0	6	2	3	1	0	1	9	2	

Transactions feature analysis : feature importance analysis of the transactions data.



Actors feature analysis : feature importance analysis of the actors data.





Demystifying Fraudulent Transactions and Illicit Nodes in the Bitcoin Network for Financial Forensics

Youssef Elmougy and Ling Liu

Georgia Institute of Technology, Atlanta GA USA

Dataset: github.com/git-disl/EllipticPlusPlus



Thank you for your attention!