

An isometric illustration of a city with various buildings, streets, and trees. A network of blue and red lines connects different points across the city, representing data flows. A prominent red square with a white exclamation mark inside is positioned on the right side, indicating a warning or alert. A large white rounded rectangle is overlaid on the center of the image, containing the title and author information.

Topology-Agnostic Detection of Temporal Money Laundering Flows in Billion-Scale Transactions

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TU/e EINDHOVEN
UNIVERSITY OF
TECHNOLOGY

TMNL
Transactie Monitoring Nederland

A framework for detecting money laundering networks

- Background and problem formulation
- Challenges and key contributions
 - Scalable
 - Topology (and typology) agnostic
 - Minimum assumptions (filtering, grouping, etc.)
 - Applicable to a multi-bank setting
- Experimental evaluation on real data
- Conclusion

Money laundering is a threat to society

- An estimated **16 billion Euros annually** are laundered just in the Netherlands
- Laundering money is of key importance to the financing of **criminal activity**
- Therefore, causing **human suffering** and large damage to society

Money laundering is a threat to society



Criminals
obtain “dirty money”
from illicit activities

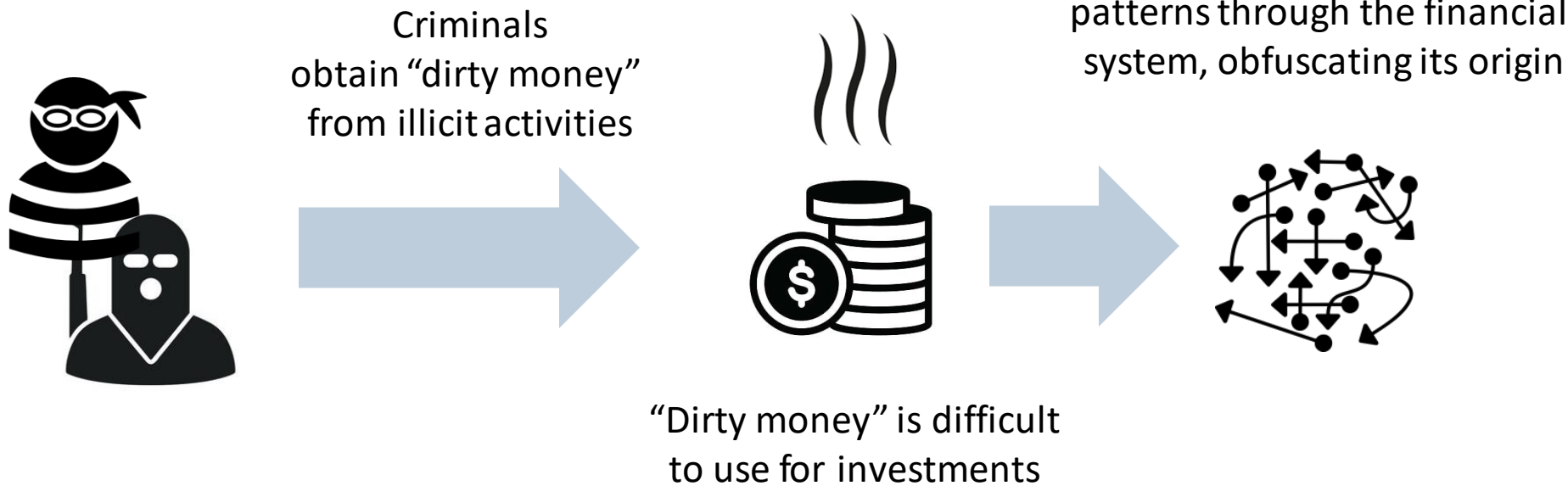


- Human trafficking
- Corruption
- Drug trafficking
- Terrorism

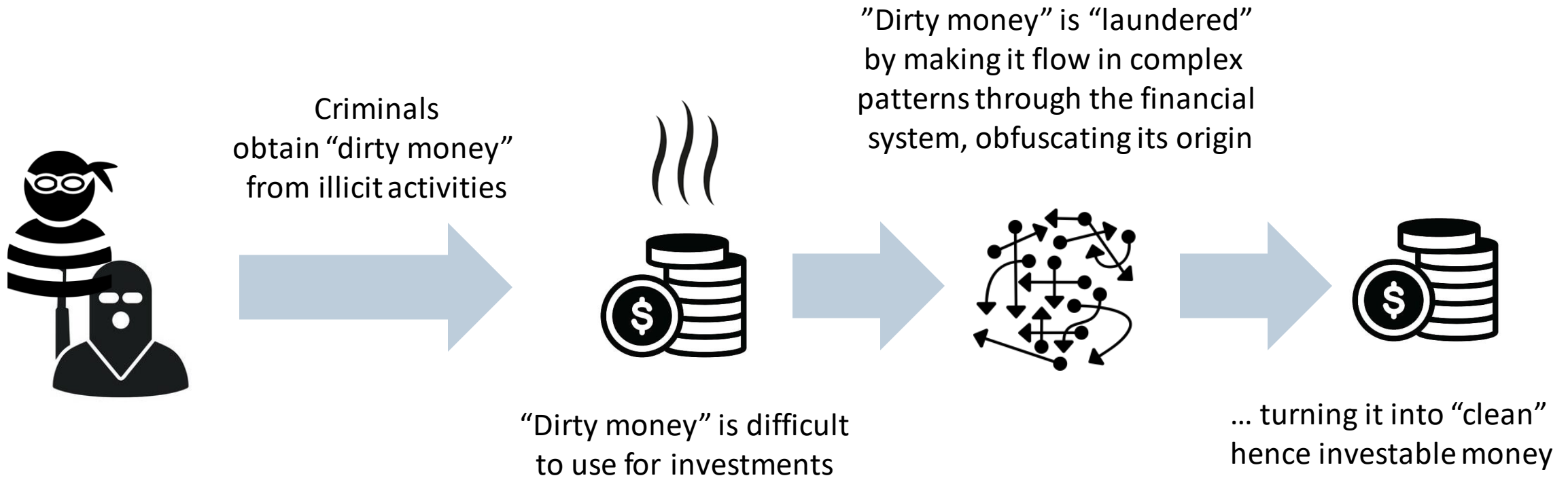
Money laundering is a threat to society



Money laundering is a threat to society








Money laundering is a threat to society



Transaction Monitoring Netherlands (TMNL)

We're fighting money laundering at an unprecedented scale

- Joint venture of 5 Dutch banks:
 





- Pooling pseudonymized transaction data** (of businesses) at TMNL
- The larger the **transaction graph**, the better we can detect money laundering
 - ... consequently, the more complex the problem becomes
- We build **models** that detect unusual patterns on the inter-bank transaction graph that might indicate money laundering



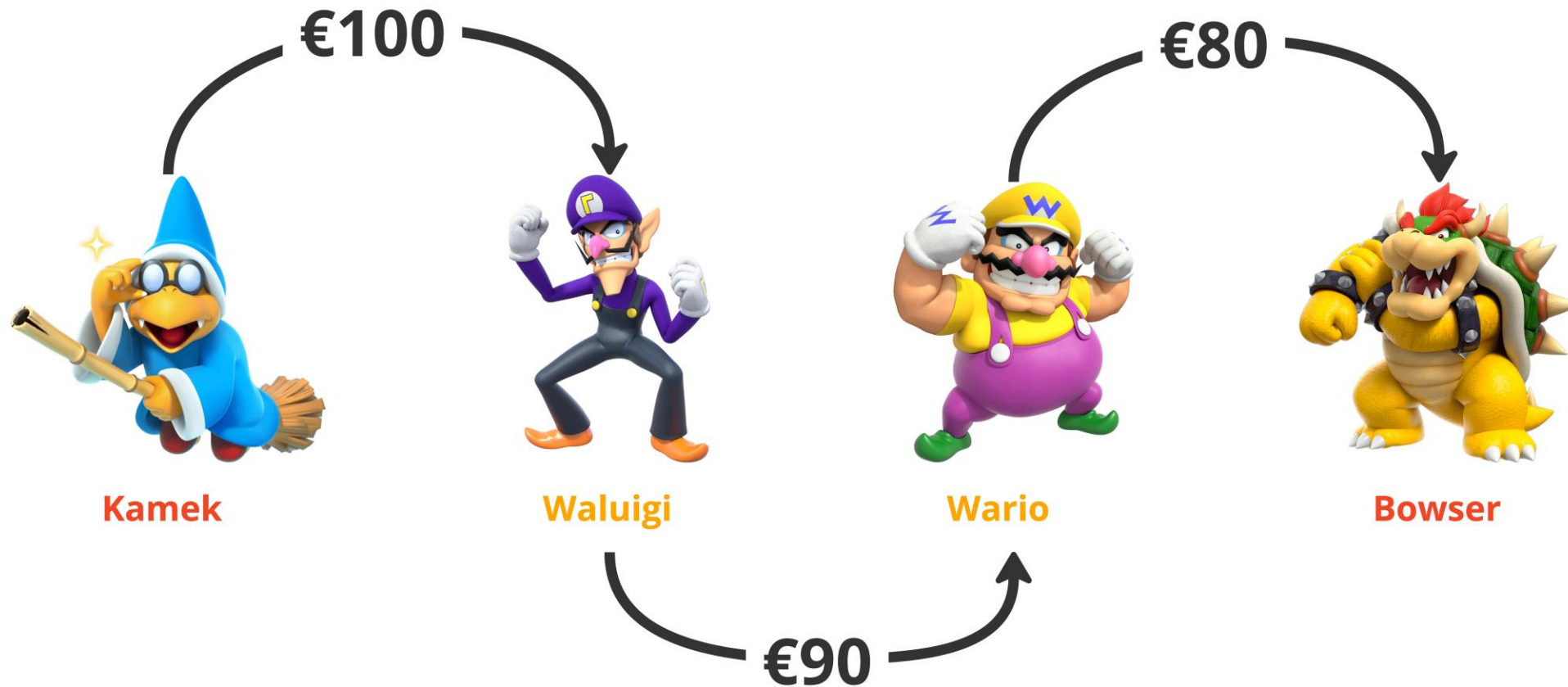
Challenges

Anti Money Laundering (AML) Modeling

- Needle-in-a-haystack problem
- Complex and ever evolving money-laundering patterns
- Computationally expensive
- Lack of data features (due to privacy, bias, etc.)

Background

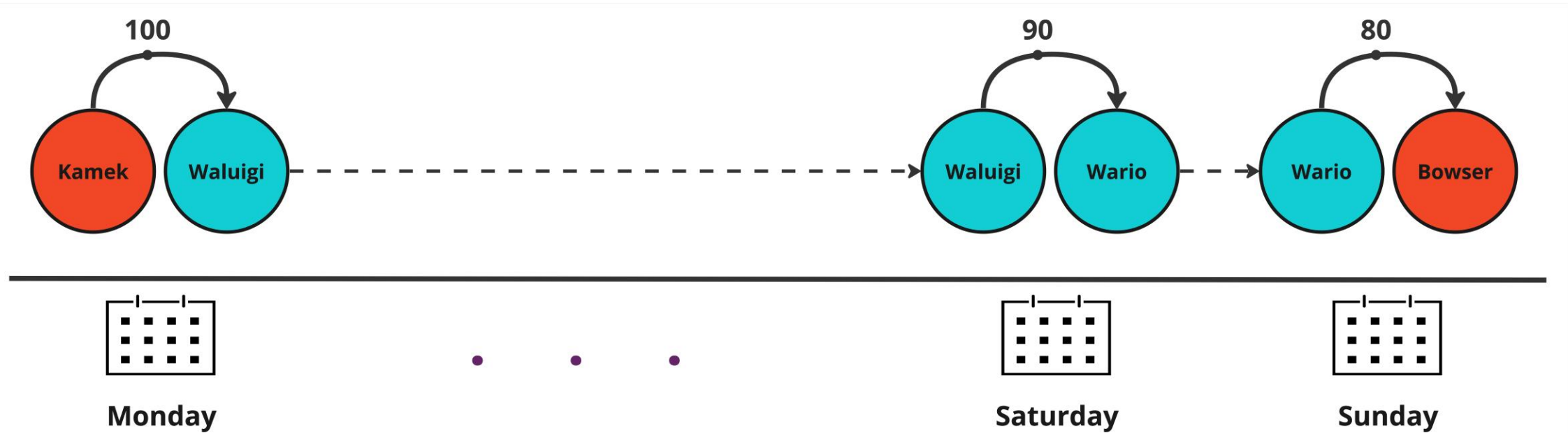
What is a flow?



Criminal | Agent

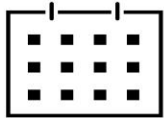
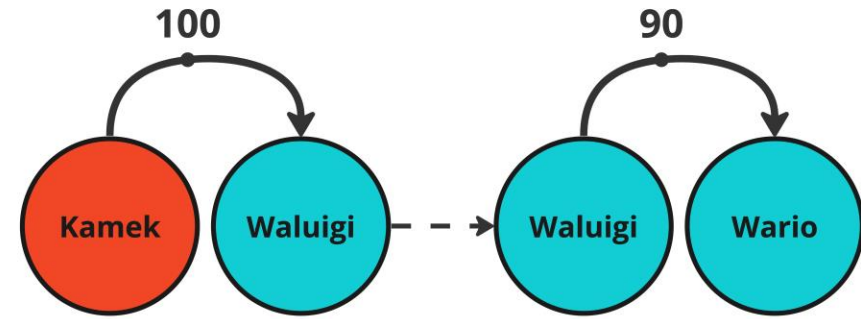
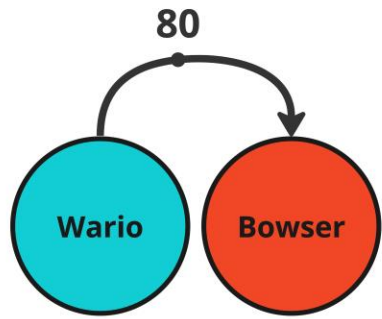
Background

What is a flow?

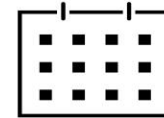


Background

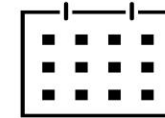
Is this a flow?



Monday



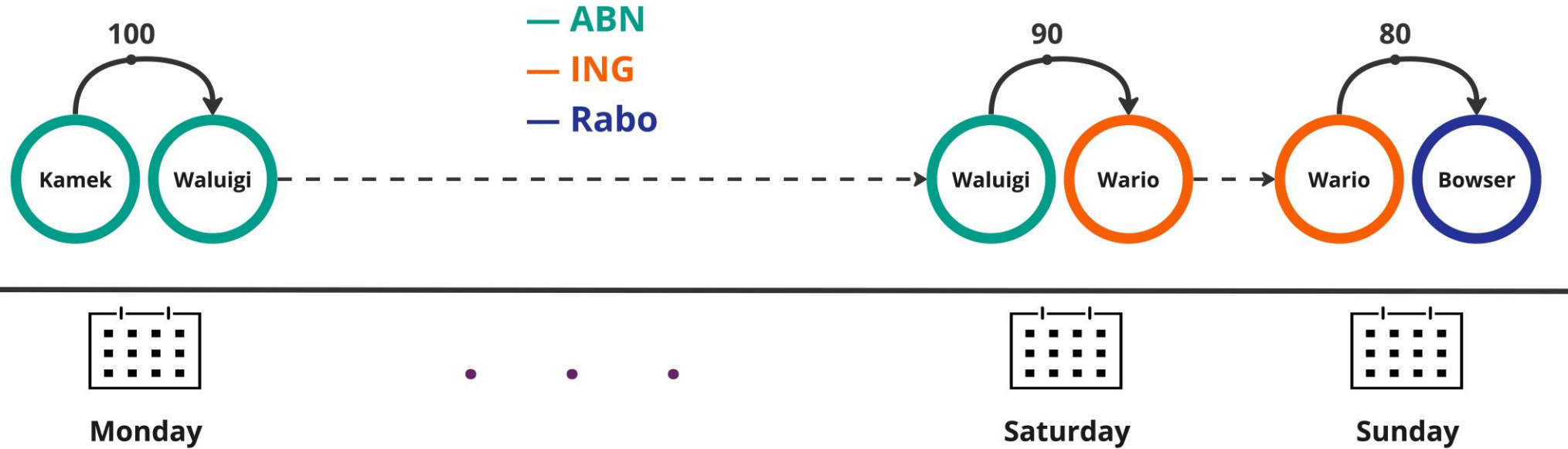
Saturday



Sunday

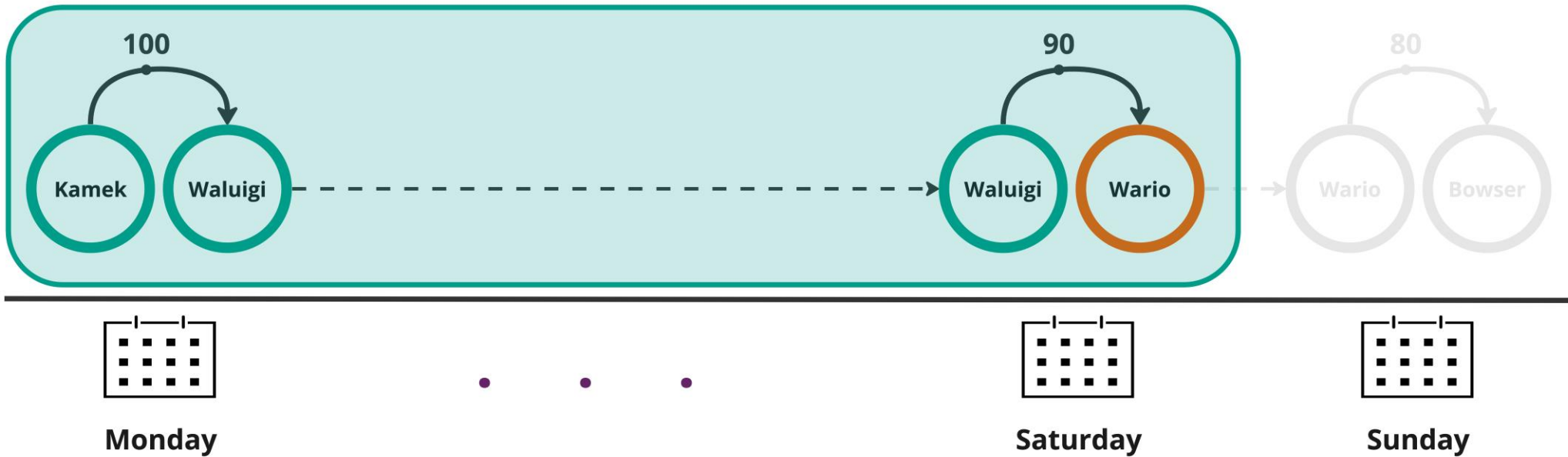
Background

Why complicate things?



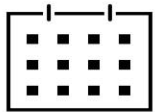
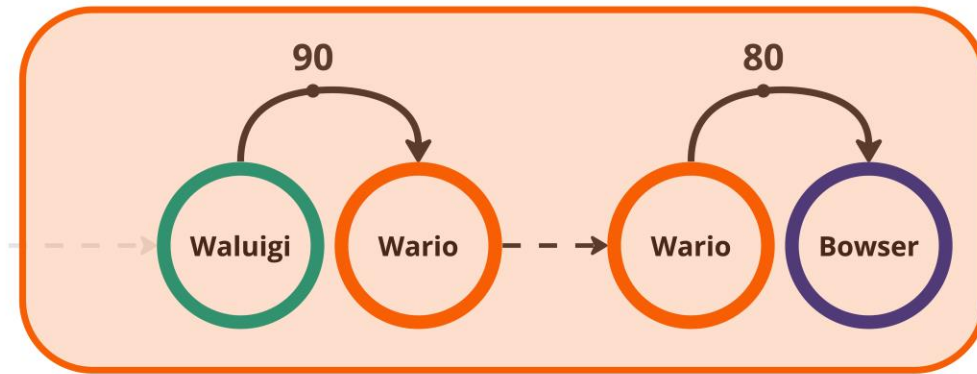
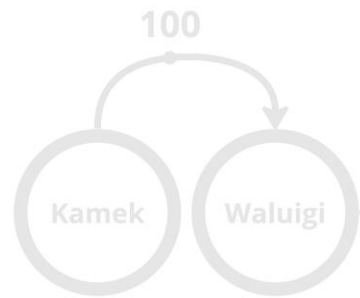
Background

Why complicate things?

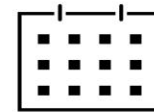


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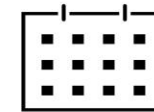
Why complicate things?



Monday



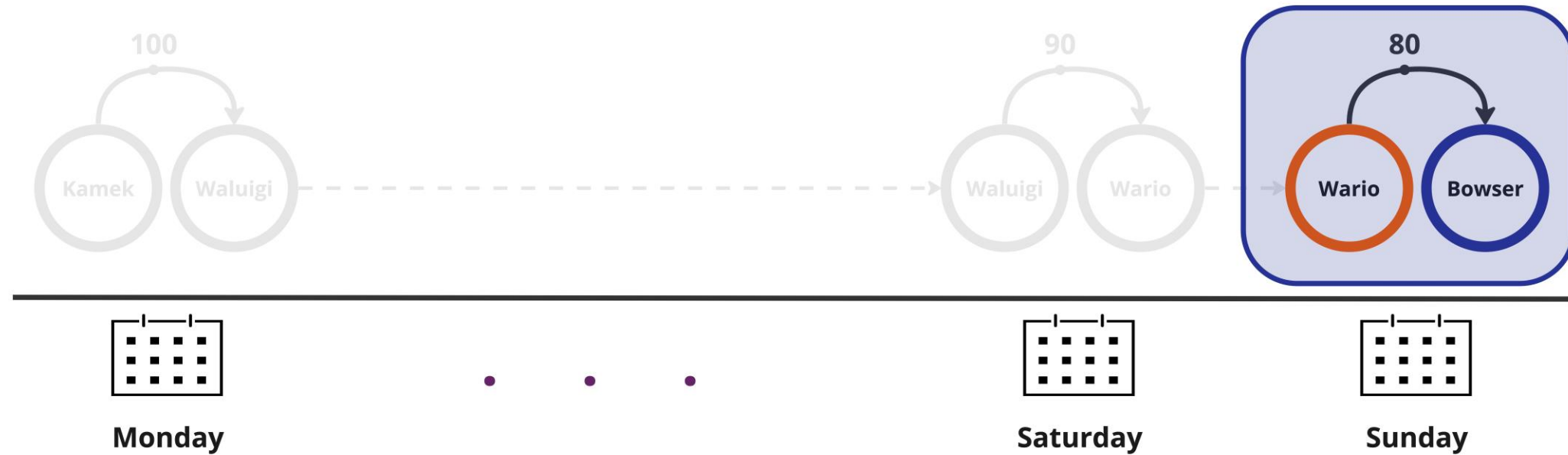
Saturday



Sunday

Background

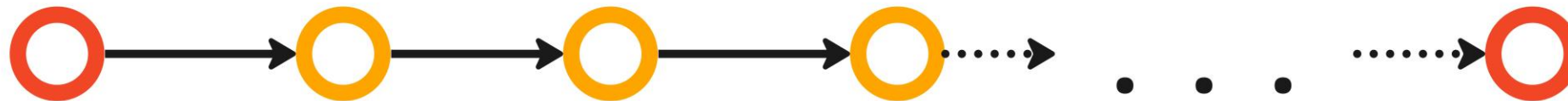
Why complicate things?



Background

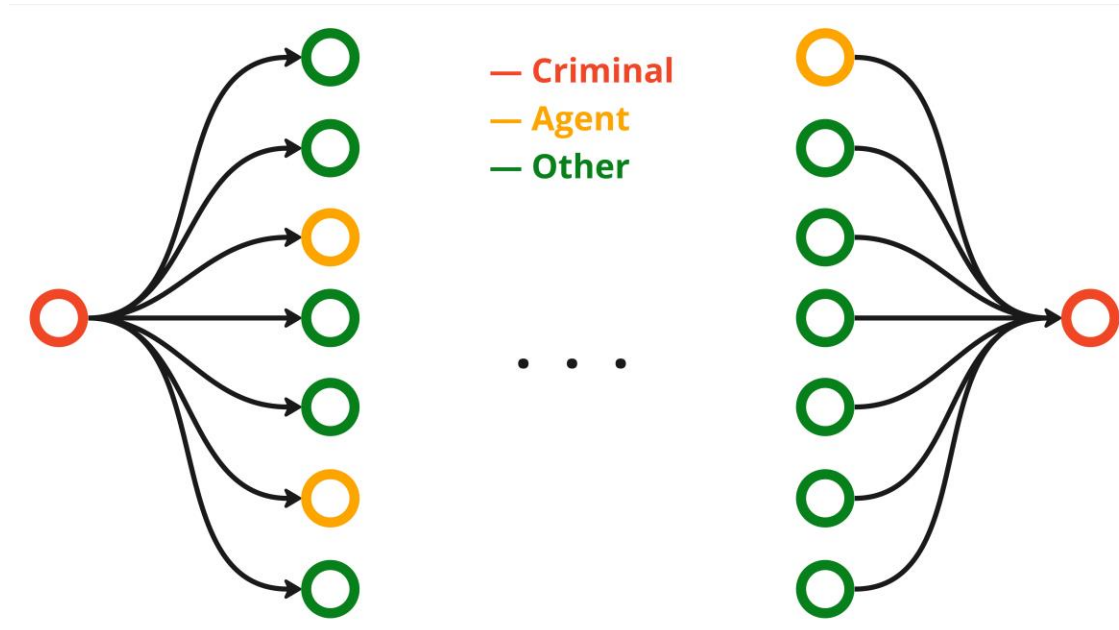
[Why Complicate] Transferring money via several hops

— Criminal
— Agent



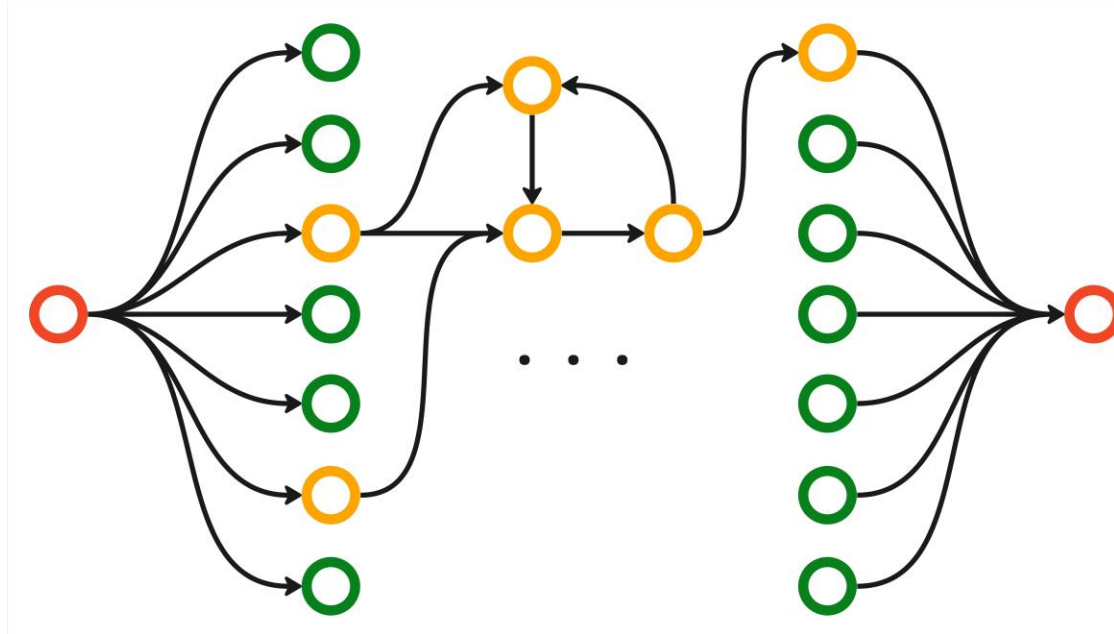
Background

[Why Complicate] Few interactions with the accomplices



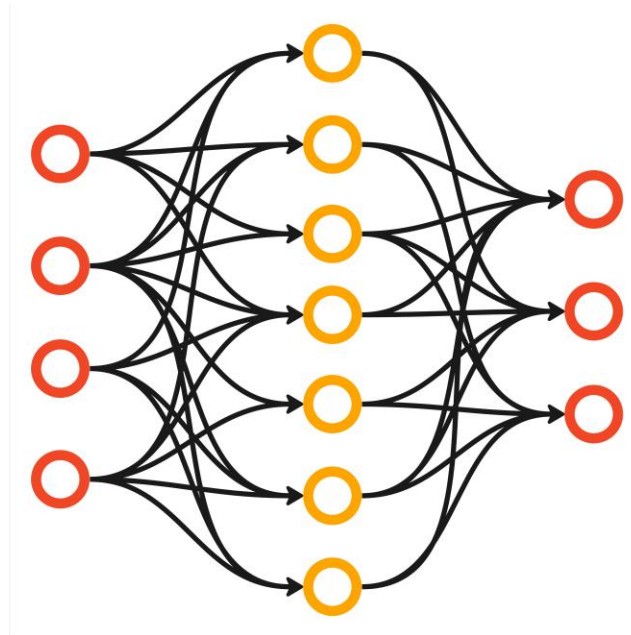
Background

[Why Complicate] More interactions among the accomplices



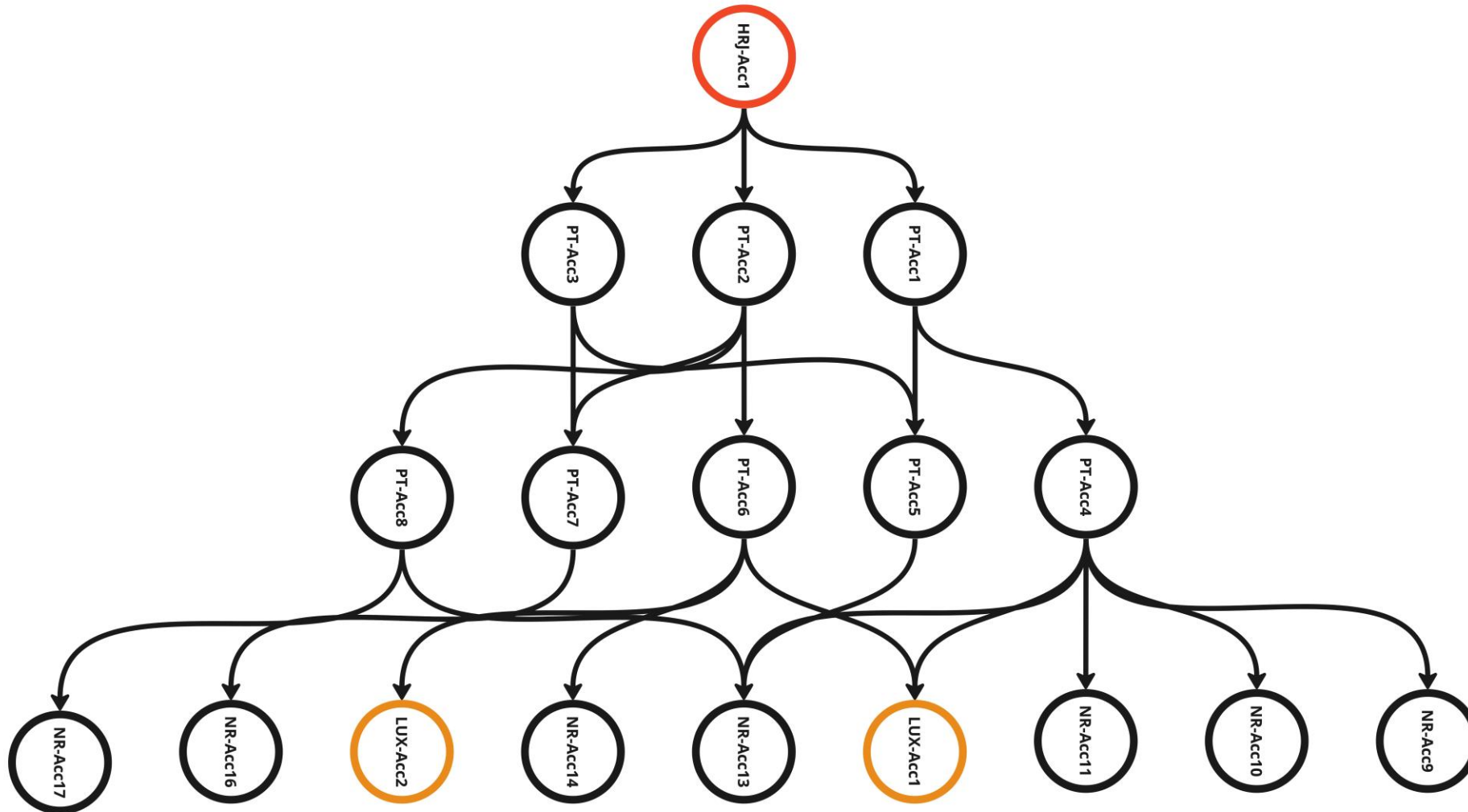
Background

[Why Complicate] Breaking down big transactions into many (small) transactions



Background

Motif queries complexity



Limitations in existing methods

- Define start and end of a flow
- Define number of hops
- Every path has the same importance
- Naïve grouping of flows

Limitations in existing methods

	Dynamic Grouping	Parameter-free for # of hops	Complex Flows	Suitable for multi-bank data
DBJ [28]	x	x	x	✓
FlowScope [26]	✓	x	x	x
FaSTM\forallN (Ours)	✓	✓	✓	✓

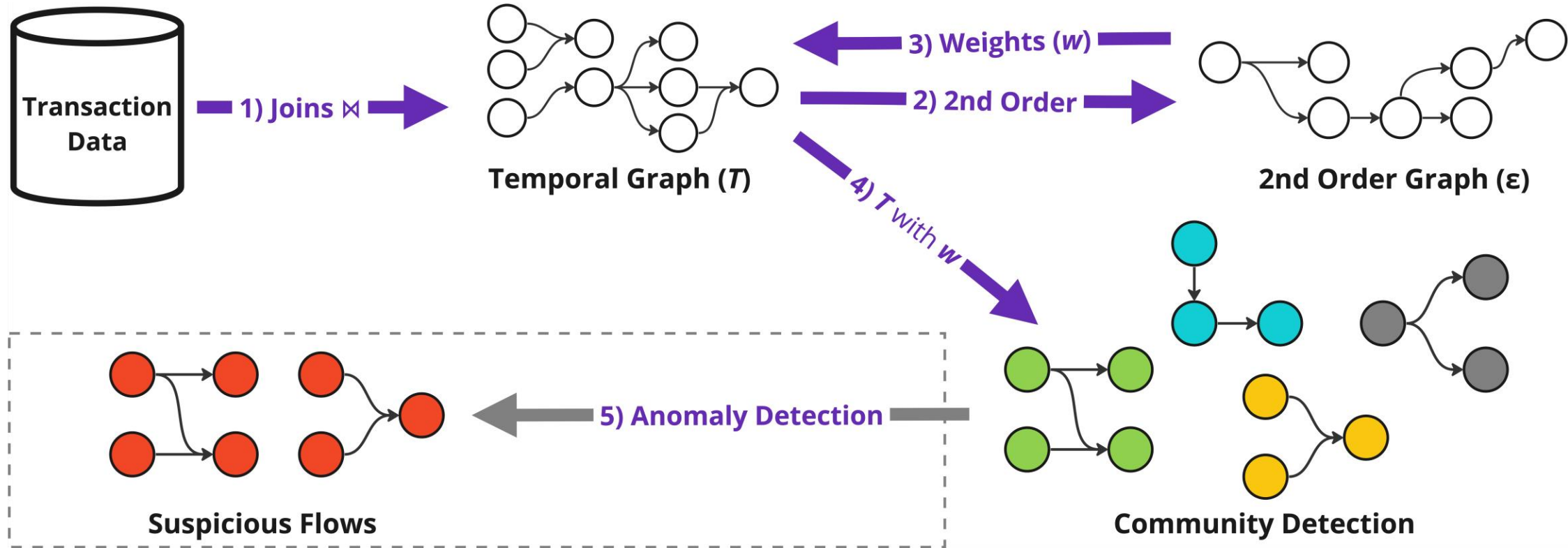
Table 1: Features Comparison of State-of-the-art AML approaches and FaSTM \forall N

[28] Michele Starnini et al. "Smurf-Based Anti-money Laundering in Time-Evolving Transaction Networks", ECML PKDD 2021.

[26] Xiangfeng Li et al. "FlowScope: Spotting Money Laundering Based on Graphs", AAAI Conference on Artificial Intelligence 2020.

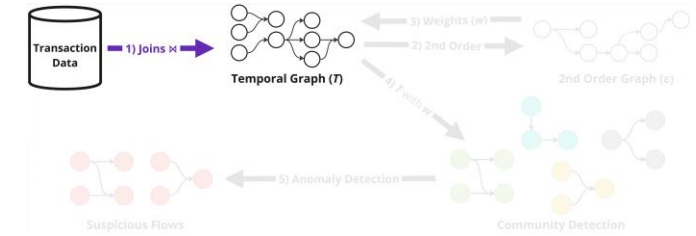
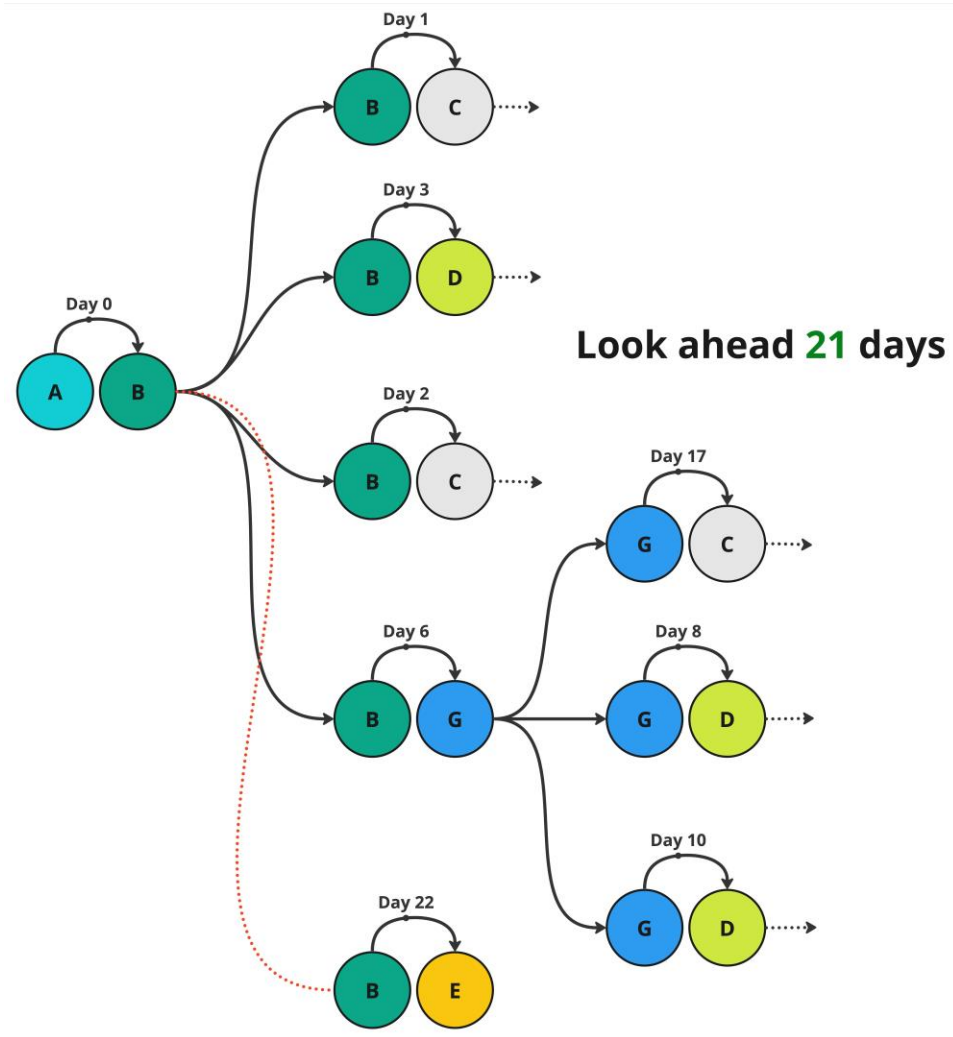
Method

Framework diagram



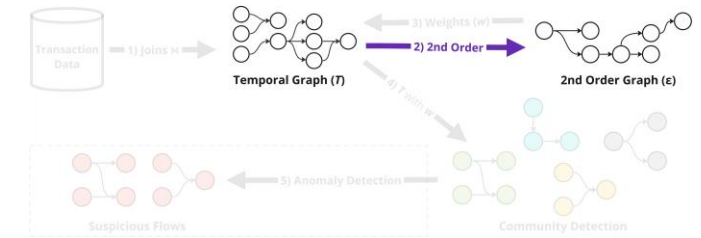
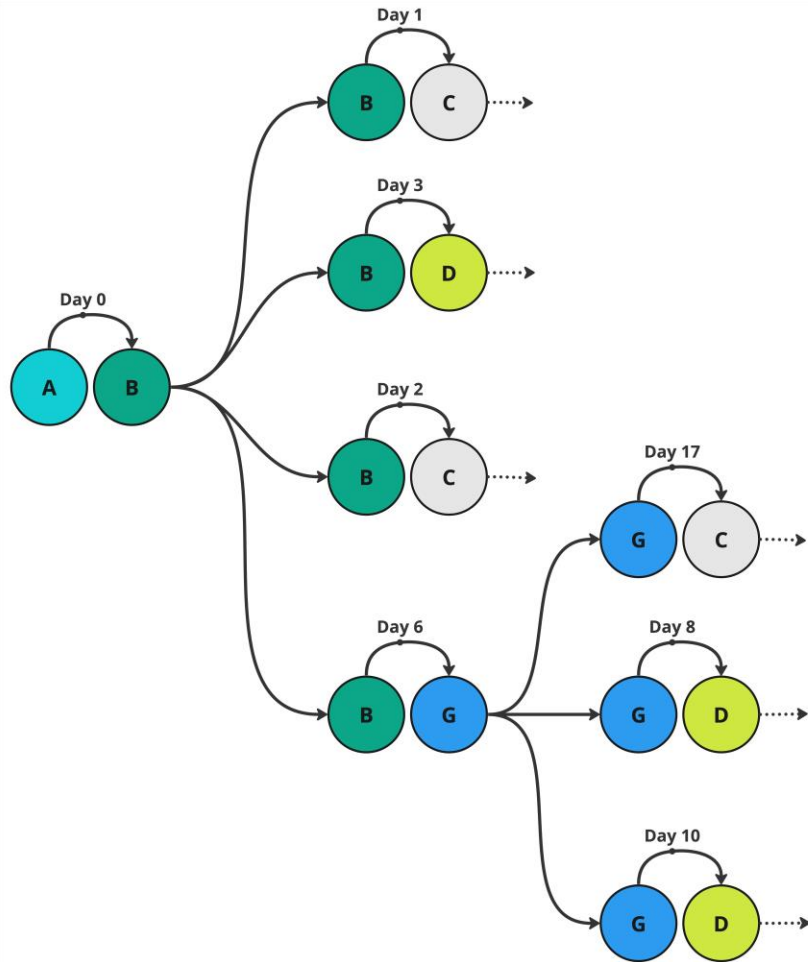
[Method] 1) Joins

Connect every transaction to every other possible transaction



[Method] 2) 2nd Order

Quantify the connections



Definition 2 (Co-occurrence Weight). Using \mathcal{S} , the co-occurrence weight between a source node $A \rightarrow B$ and a destination node $B \rightarrow C$ is calculated as,

$$W(A \rightarrow B, B \rightarrow C) = \max(\mathcal{P}(A \rightarrow B, B \rightarrow C), \mathcal{P}'(A \rightarrow B, B \rightarrow C))$$

where,

$$\mathcal{P}(A \rightarrow B, B \rightarrow C) = \frac{|\mathcal{S}(A \rightarrow B \sim B \rightarrow C)|}{|\mathcal{S}(A \rightarrow B \sim B \rightarrow [*])|}$$

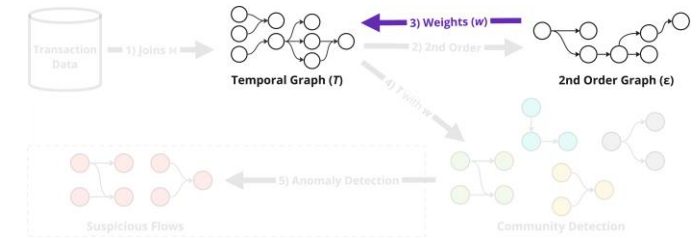
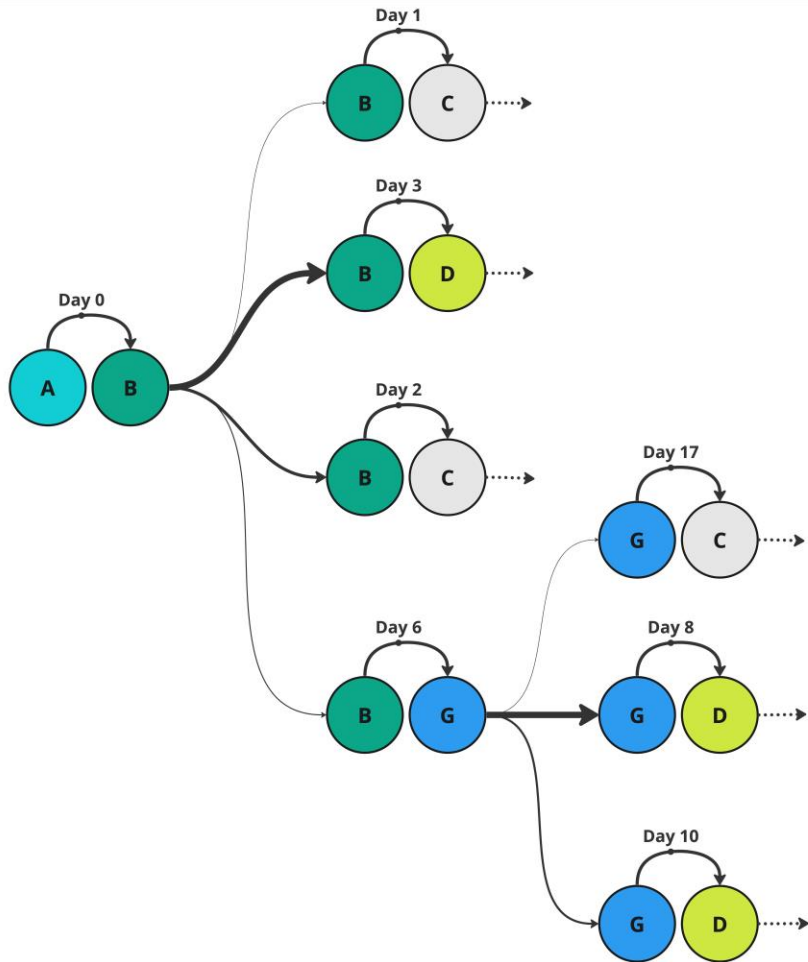
and,

$$\mathcal{P}'(A \rightarrow B, B \rightarrow C) = \frac{|\mathcal{S}(A \rightarrow B \sim B \rightarrow C)|}{|\mathcal{S}([*] \rightarrow B \sim B \rightarrow C)|}$$

where, $[*]$ represents **any** account and \sim represents directed adjacency from the left to the right node(s).

[Method] 3) Weights

Apply the weights

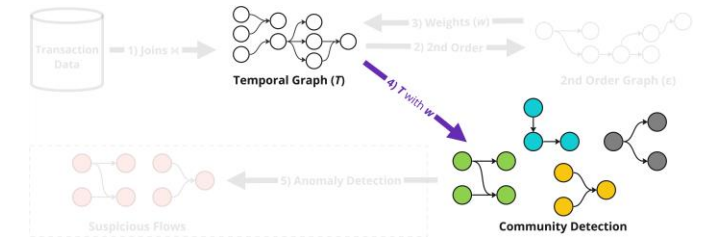
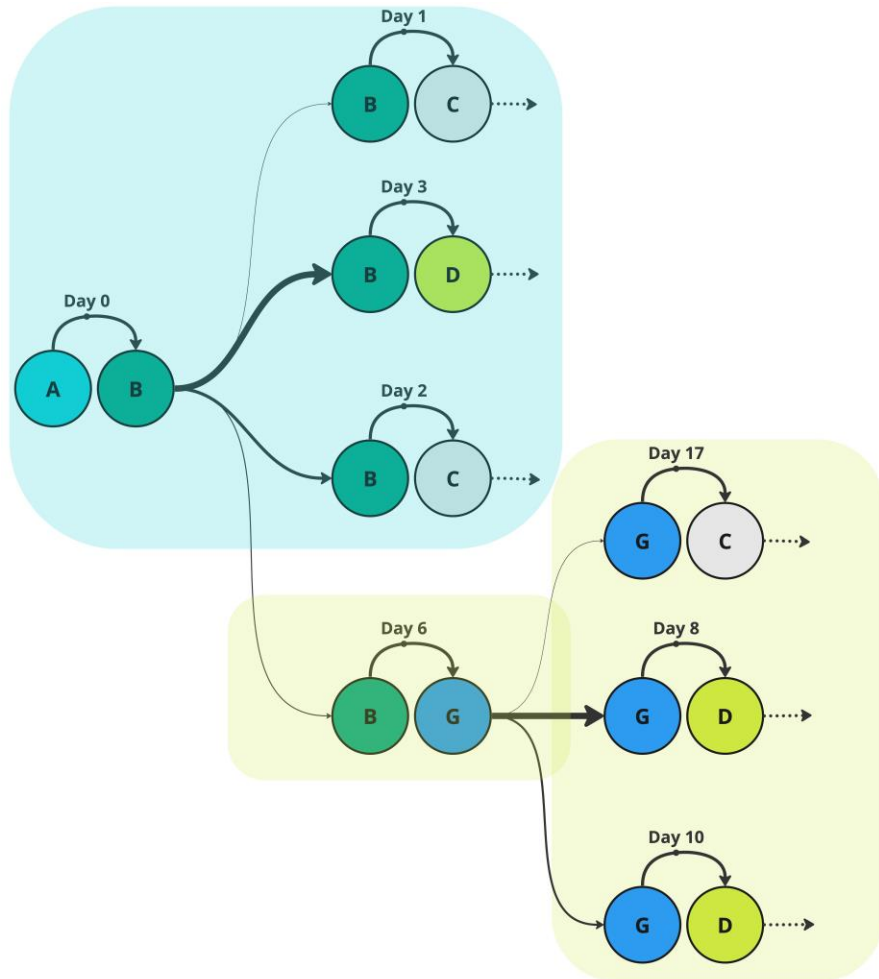


Source Perspective		Destination Perspective	
A→B	B→C	A→B	B→C
D→B	B→E	D→B	B→E
D→B	B→E	D→B	B→E
A→B	B→W	A→B	B→W
A→B	B→C	A→B	B→C
K→B	B→L	K→B	B→L
Z→B	B→T	Z→B	B→T
Z→B	B→C	Z→B	B→C
A→B	B→L	A→B	B→L
A→B	B→G	A→B	B→G
= 2/5		= 2/3	

= MAX(2/5, 2/3)
= 2/3

[Method] 4) Community Detection

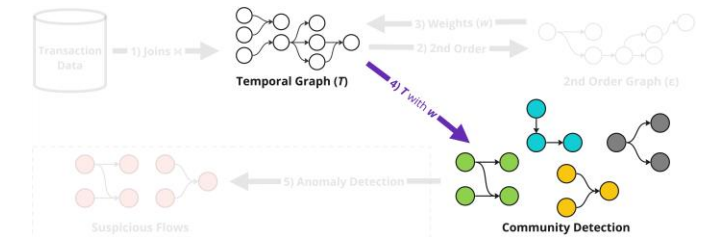
Detect communities of connected transactions



[Method] 4) Community Detection

Detect communities of connected transactions

- Transactions that are *strongly* connected form a community
 - It will not appear in any other community
 - The other transactions in community-y have strong dependence on transaction-x
 - The transactions in other communities have weak(er) dependence on transaction-x

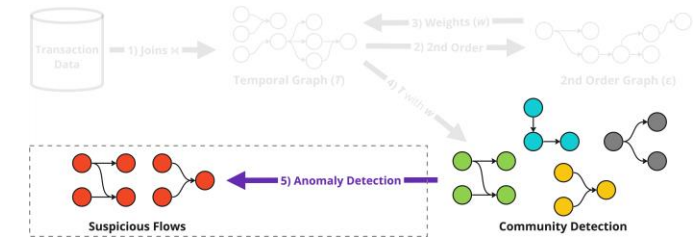


[Method] 5) Suspicious Flows

Marking communities of transactions as suspicious

- Max-flow based approach
 - Cash deposits as sources
 - HRJ deposits as sinks

- Graph level Anomaly Detection (GLAD)
 - Graph embeddings
 - Autoencoders
 - Isolation forest
 - ...?



Follow all Suspicious Trails of Money *for all* Nodes (FaSTM \forall N)

Space Complexity

Step	Transactions	\mathcal{T} Edges
Initial state	1.1 billion	-
Pre-processing	510 million	-
\mathcal{T} creation	475 million	25 billion
Remove weak edges	325 million	2.3 billion

Table 2: Space explosion and implosion after each step

Runtimes

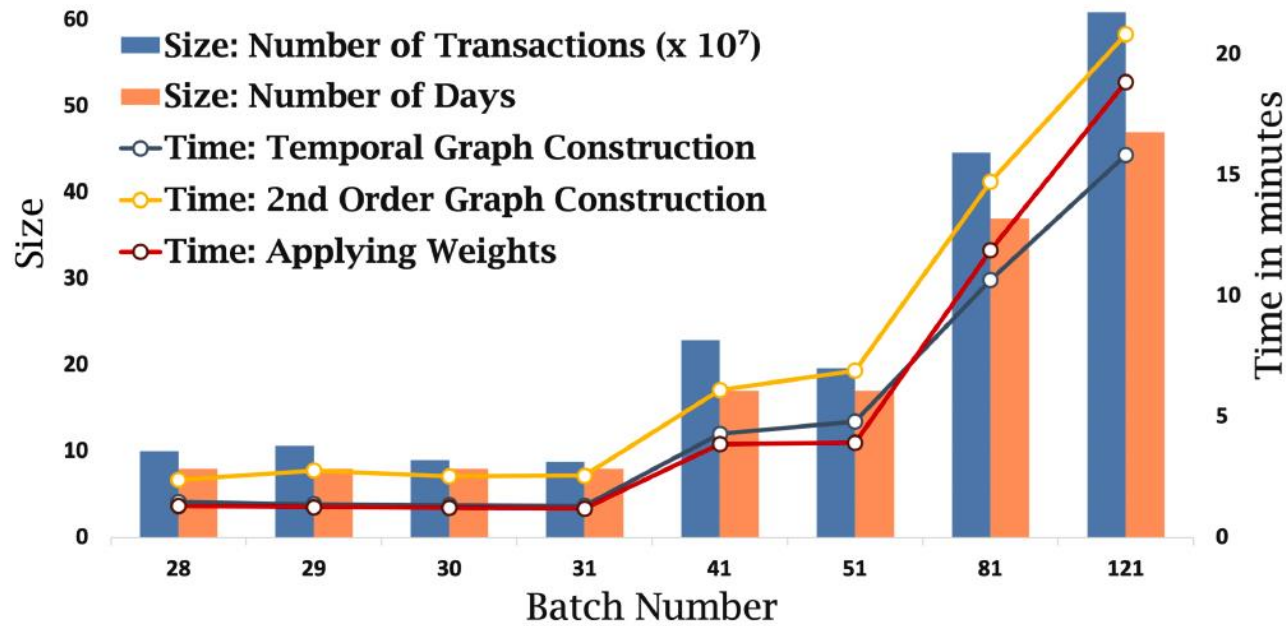


Fig. 8: Runtimes for batches with different number of days in the data

Functional and usability comparison

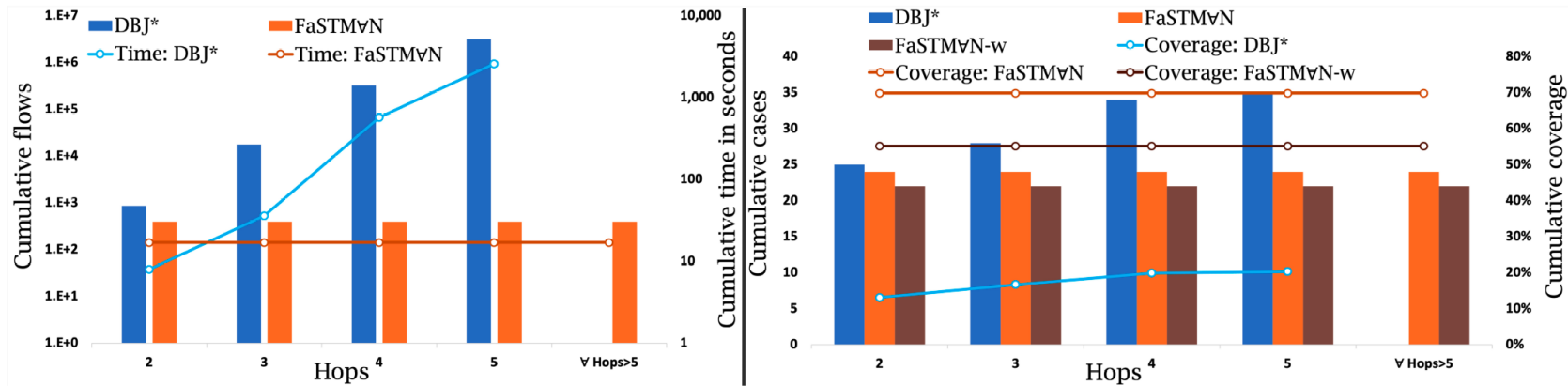


Fig. 10: Suspicious flow detection comparisons. (a) A comparison for runtimes. Both y-axes scales are logarithmic. (b) A functional comparison for the suspicious flows. Higher coverage with lower number of cases is the desired outcome.

Topology-agnostic nature

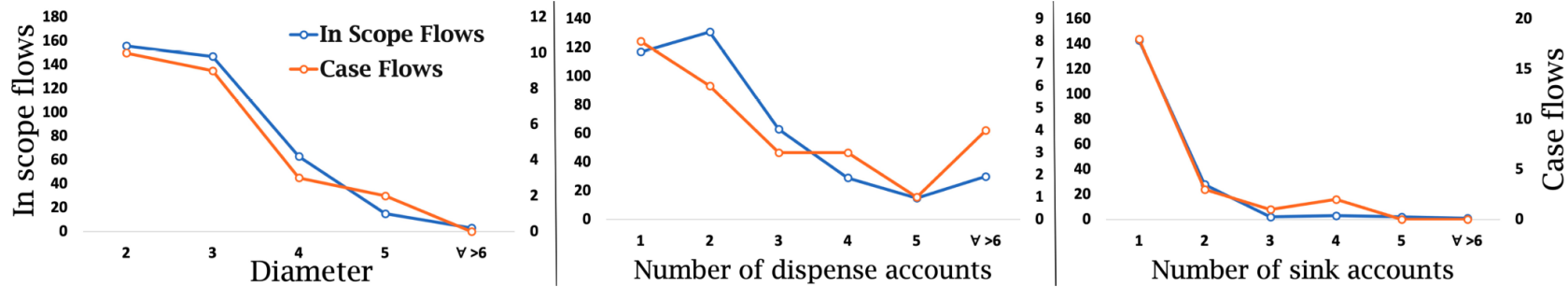


Fig. 11: Topological diversity of the flows

Functional and usability comparison

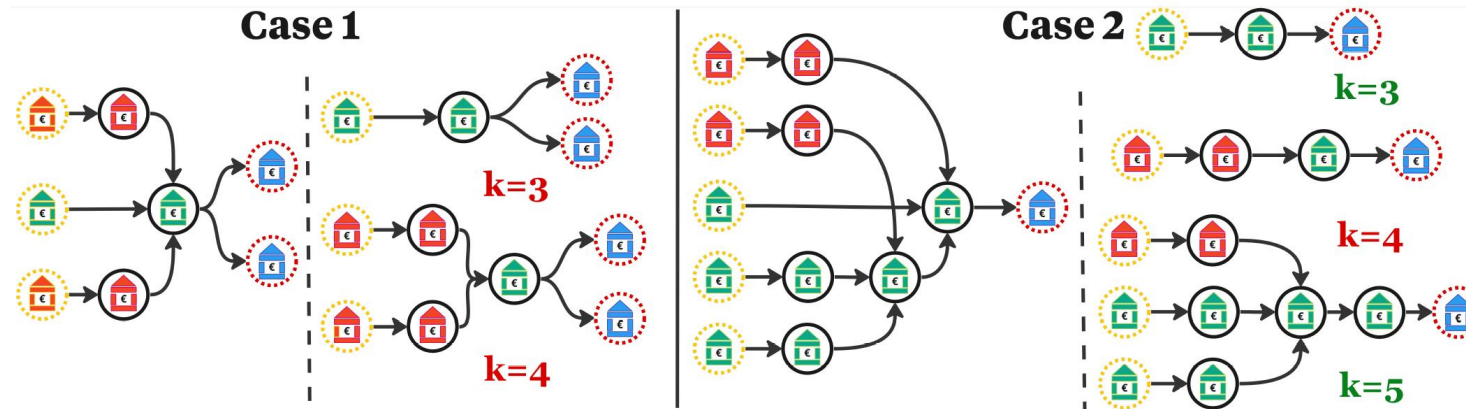


Fig. 12: Two cases of **real** flows. On the left of the dashed lines are the flows detected by FaSTM/N, and on the right, the series of **separate** flows detected by FlowScope. The red font for $k=x$ indicates that the flow was not flagged suspicious by FlowScope, based on risk criterion $C3$.

Conclusion

Future work and improvements

- Using higher ($> 2^{\text{nd}}$) order or multi-order representations may reveal more interesting relationships
- Experimentation with the edge weights is important based on business problem – you are looking to capture meaningful relationships based on what you deem important for the modus operandi
- Community detection
 - Based on recurring flows, over different periods of time, detect communities of entities
- Targeted network search
 - Return all the dominant flows a query account is involved in

Conclusion

Questions

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- Transaction Monitoring Netherlands (TMNL)
 - <https://tmnl.nl/>

- <https://github.com/mhaseebtariq/fastman>

