

Behavioural Network Traffic Analytics for Securing 5G Networks

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Presentation outline

- Problem formulation
- Proposed method
- Experimental results
- Conclusions



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- **Securing Mobile networks – Malware detection:**
 - Spam/Premium SMS/Call, DDoS SMS-flooding, DDoS by sending periodically Internet packets
- **Diversity of the malware types and behaviours**
 - Renders the problem of anomaly detection as a very challenging one
- **Multi-dimensional** nature of the data makes it difficult to analyse
 - SMS, Call, Internet, Services, Signalling
- More **challenging in 5G networks**, since one more dimension is added to the traffic, representing different network slices
 - Activity that is normal in one slice can be anomalous in another

- **Data types** in the mobile network:
 - **Signalling (control) plane:** all the signals that control or are needed for the network services (e.g. Call Forwarding enable/disable or Call handover)
 - **Billing (data) plane:** comprised of actual data sent/received by the mobile devices, including Call Detail Records (CDR), and Internet traffic
- Focus on the detection of malware on the **billing plane:**
 - No content used due to privacy concerns
 - Only high level communication events (who communicates with who and how/when)



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- **Behavioural-based approaches**

- Extract descriptors that capture different aspects of the behaviour of malicious and normal actors, allowing for their efficient discrimination

Behaviour: Range of actions taken by actors in conjunction with themselves and their environment.

In the context of mobile networks, the actors are the mobile devices, environment is the rest of the mobile devices and network, and actions are the communications among them.

- This paper proposes the **Behavioral Traffic Analysis** method, for discriminating between different user behaviors
- The method is an extension of the Multi-objective Clustering approach [Kalamaras et al. 2015] by extending the proposed behavioral descriptors

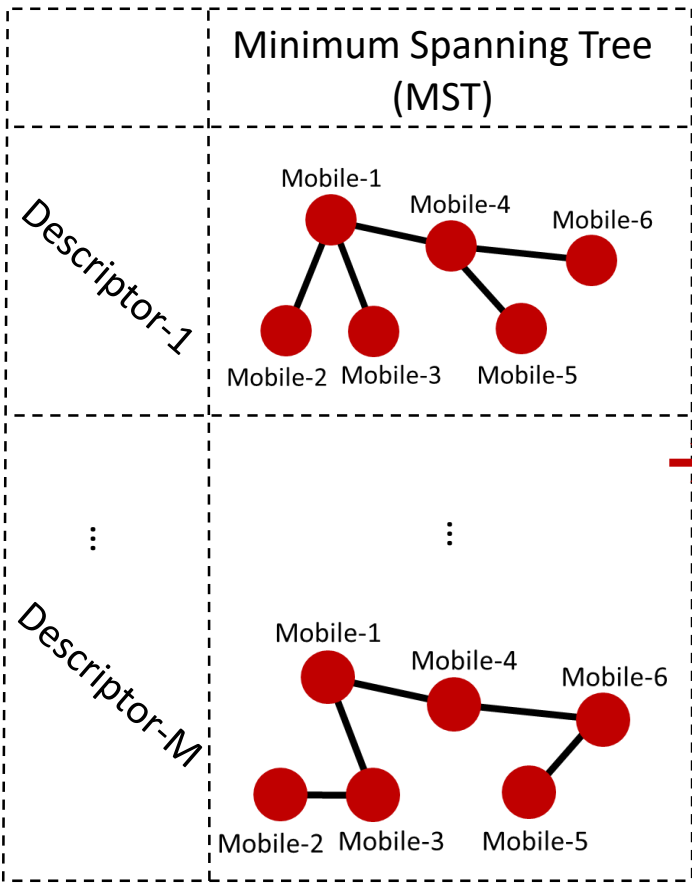
Proposed method

Multi-objective Clustering framework 1/2

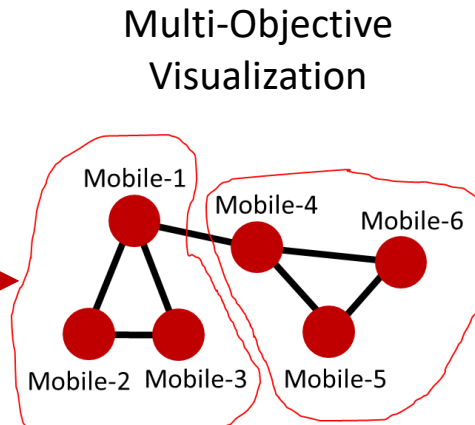
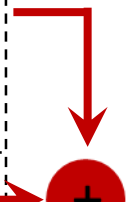


	Descriptor-1	... Descriptor-M
Mobile-1	Descriptor-1 for Mobile-1	Descriptor-M for Mobile-1
Mobile-2	Descriptor-1 for Mobile-2	Descriptor-M for Mobile-2
⋮	⋮	⋮
Mobile-N	Descriptor-1 for Mobile-M	Descriptor-M for Mobile-N

(e.g. Call Descriptor) (e.g. SMS Descriptor)

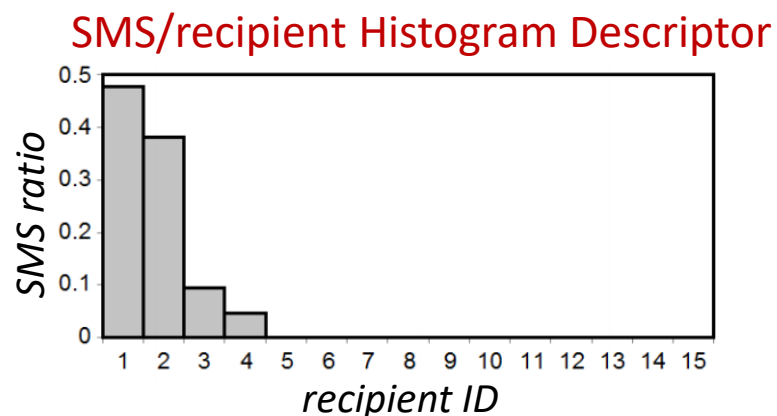
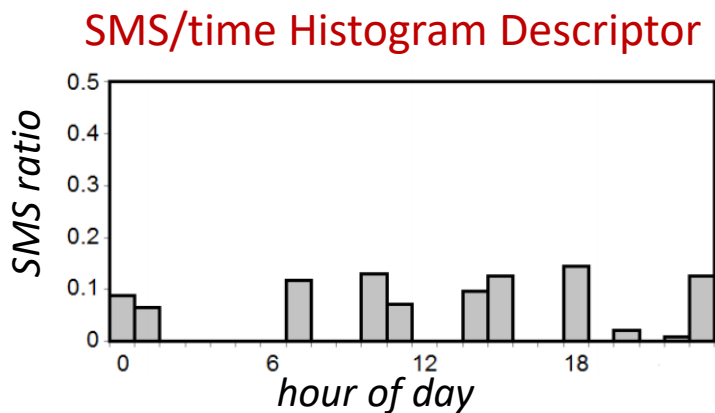


(MSTs created using a descriptor-specific distance metric)



(Positions computed using a weighted force directed graph layout)

- Inputs of Multi-objective Clustering framework
 - Descriptor definitions
 - Distance metric between descriptors
- Example of Multi-objective Clustering approach [Kalamaras et al. 2015]
 - Proposed Descriptors for both SMS and Call activities



**these descriptors are also defined for the call activity of each device (i.e. 4 descriptors in total)*

- Distance metric between descriptors: **L1**

- k-partite graphs created by a subset of billing attributes
- Each attribute value is mapped into a single graph node
- Continuous attributes (e.g. date-time, duration) are discretized

Origin	Dest	Slice	Type
m1	m4	s1	SMS
m1	m4	s1	SMS
m1	m2	s1	CALL
m1	m3	s2	CALL
m1	m2	s2	CALL
m2	m3	s1	CALL
m2	m3	s1	SMS
m2	m3	s1	CALL
m2	m1	s1	SMS

Billing data

Example of descriptors:

1. CALL descriptor:

Origin/Dest/Slice
for CALL activity

2. SMS descriptor:

Origin/Dest/Slice
for SMS activity

Origin	Dest	Slice	Type
m1	m2	s1	CALL
m1	m3	s2	CALL
m1	m2	s2	CALL

Billing data used for the
CALL descriptor of m1

?

CALL descriptor of m1

Proposed Behavioural Analytics method

Proposed Descriptors

- k-partite graphs created by a subset of billing attributes
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m2	m3	s1	SMS
m2	m3	s1	CALL
m2	m1	s1	SMS

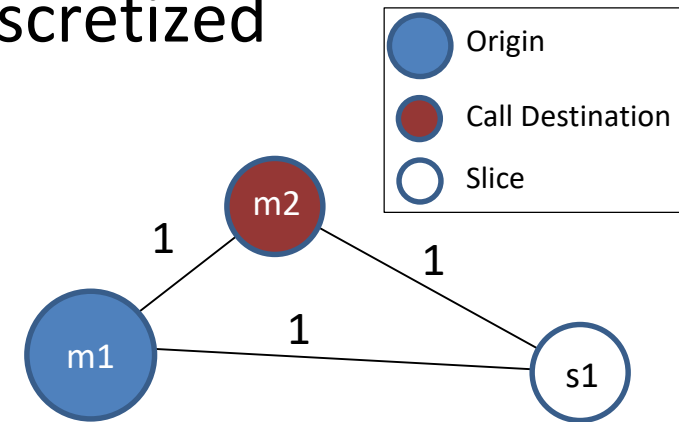
Billing data

Example of descriptors:

- CALL descriptor:**
Origin/Dest/Slice for CALL activity
- SMS descriptor:**
Origin/Dest/Slice for SMS activity

Origin	Dest	Slice	Type
m1	m2	s1	CALL
m1	m3	s2	CALL
m1	m2	s2	CALL

Billing data used for the CALL descriptor of m1



CALL descriptor of m1

Proposed Behavioural Analytics method

Proposed Descriptors

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m2	m3	s1	SMS
m2	m3	s1	CALL
m2	m1	s1	SMS

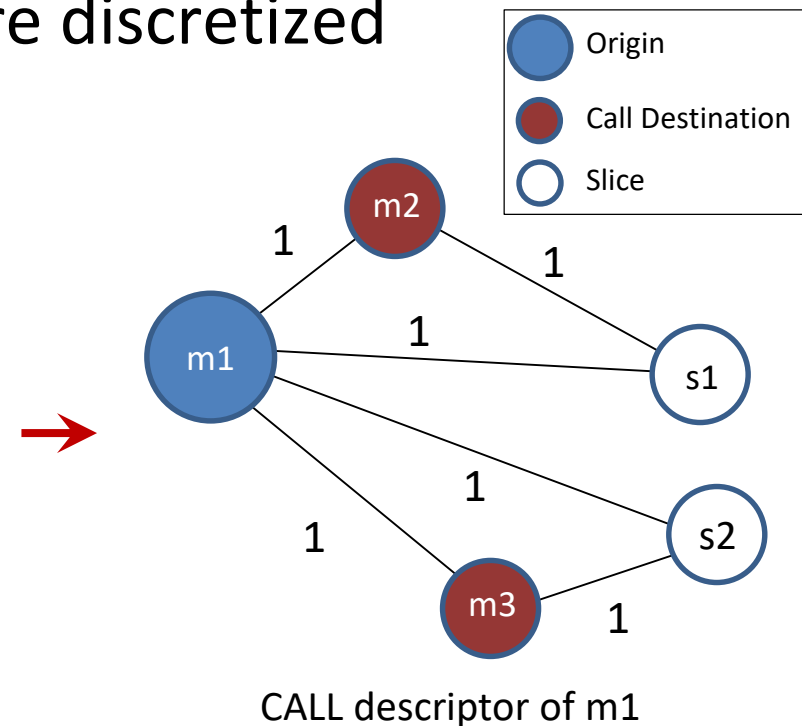
Billing data

Example of descriptors:

- CALL descriptor:**
Origin/Dest/Slice for CALL activity
- SMS descriptor:**
Origin/Dest/Slice for SMS activity

Origin	Dest	Slice	Type
m1	m2	s1	CALL
m1	m3	s2	CALL
m1	m2	s2	CALL

Billing data used for the CALL descriptor of m1



CALL descriptor of m1

Proposed Behavioural Analytics method

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m2	m3	s1	SMS
m2	m3	s1	CALL
m2	m1	s1	SMS

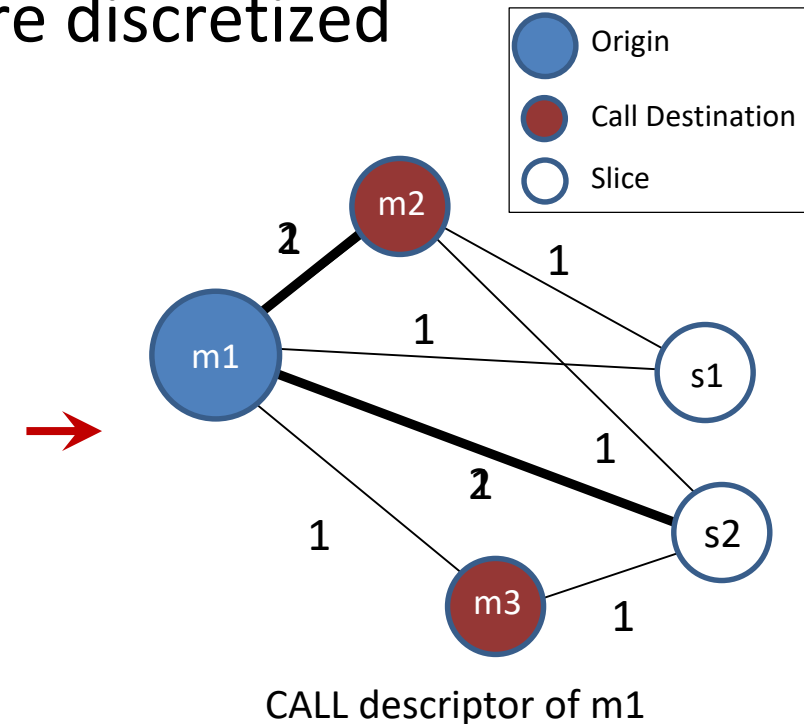
Billing data

Example of descriptors:

- CALL descriptor:**
Origin/Dest/Slice for CALL activity
- SMS descriptor:**
Origin/Dest/Slice for SMS activity

Origin	Dest	Slice	Type
m1	m2	s1	CALL
m1	m3	s2	CALL
m1	m2	s2	CALL

Billing data used for the CALL descriptor of m1



- k-partite graphs created by a subset of billing attributes
- Each attribute value is mapped into a single graph node
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Origin	Dest	Slice	Type
m1	m4	s1	SMS
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m1	m2	s1	CALL
m1	m3	s2	CALL
m1	m2	s2	CALL
m2	m3	s1	CALL
m2	m3	s1	SMS
m2	m3	s1	CALL
m2	m1	s1	SMS

Billing data

Example of descriptors:

1. CALL descriptor:

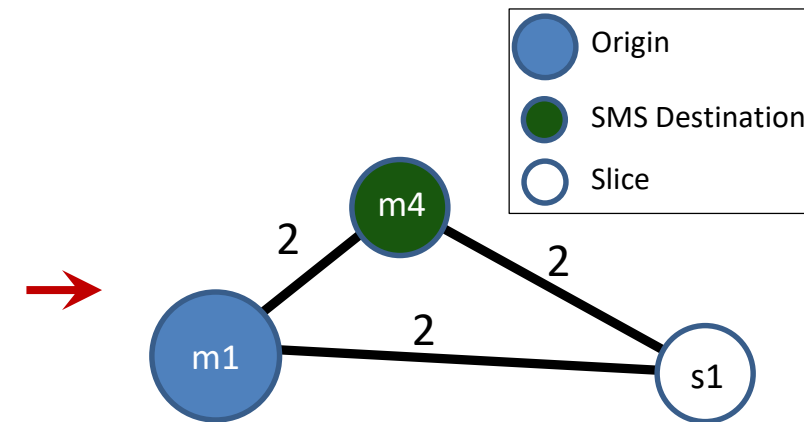
Origin/Dest/Slice for CALL activity

2. SMS descriptor:

Origin/Dest/Slice for SMS activity

Origin	Dest	Slice	Type
m1	m4	s1	SMS
m1	m4	s1	SMS

Billing data used for the SMS descriptor of m1



SMS descriptor of m1

- Distance metric defined using **graph matching techniques**
- For mobile- i and mobile- j , their distance with respect to descriptor- k is defined as:

$$D_k(G_k^i, G_k^j) = w_{eig} D_k^{eig}(G_k^i, G_k^j) + w_{adj} D_k^{adj}(G_k^i, G_k^j)$$

[Koutra et al. 2011] structural information
using the graph eigenvalues λ

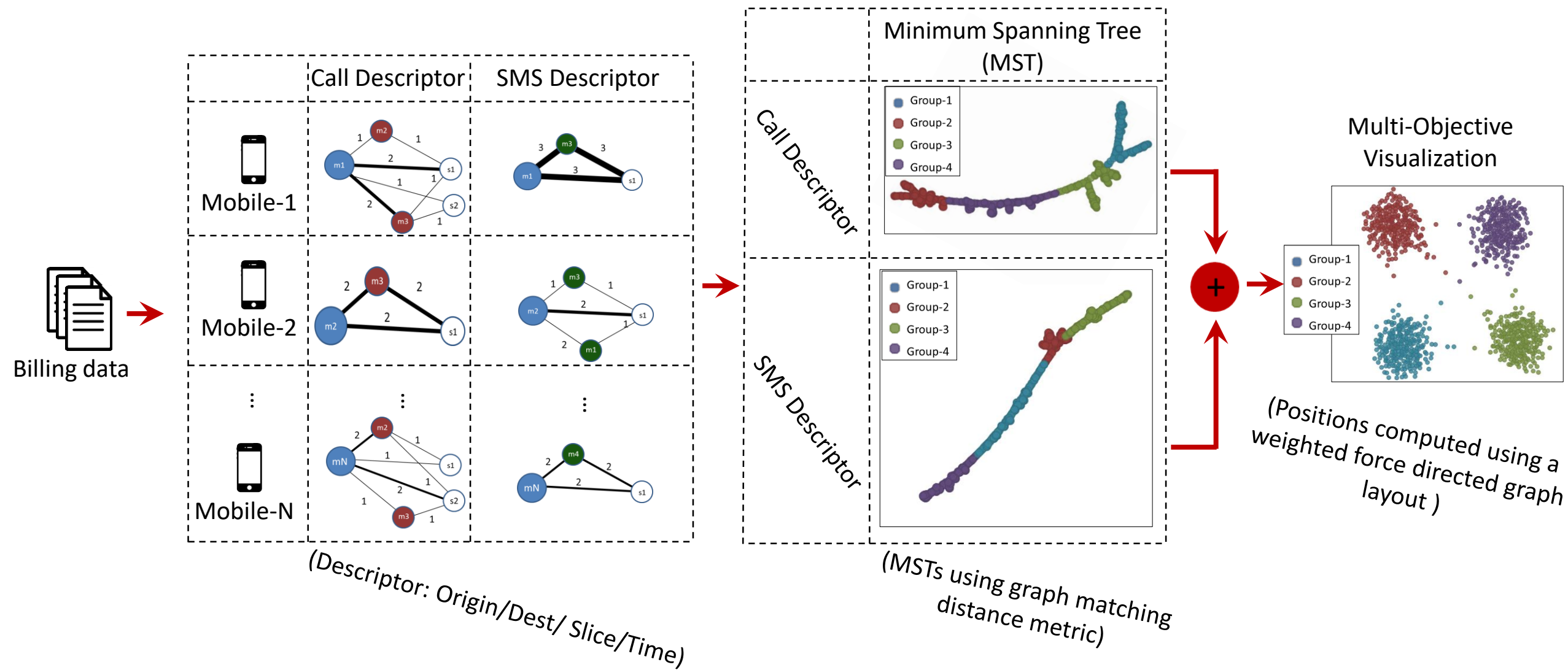
$$D_k^{eig}(G_k^i, G_k^j) = \sum_{h=1}^{h_{max}} (\lambda_k^{i,h} - \lambda_k^{j,h})^2$$

content information using the graph
adjacency matrices M

$$D_k^{adj}(G_k^i, G_k^j) = \sum |M_k^i - M_k^j|$$

Proposed Behavioural Analytics method

Overview





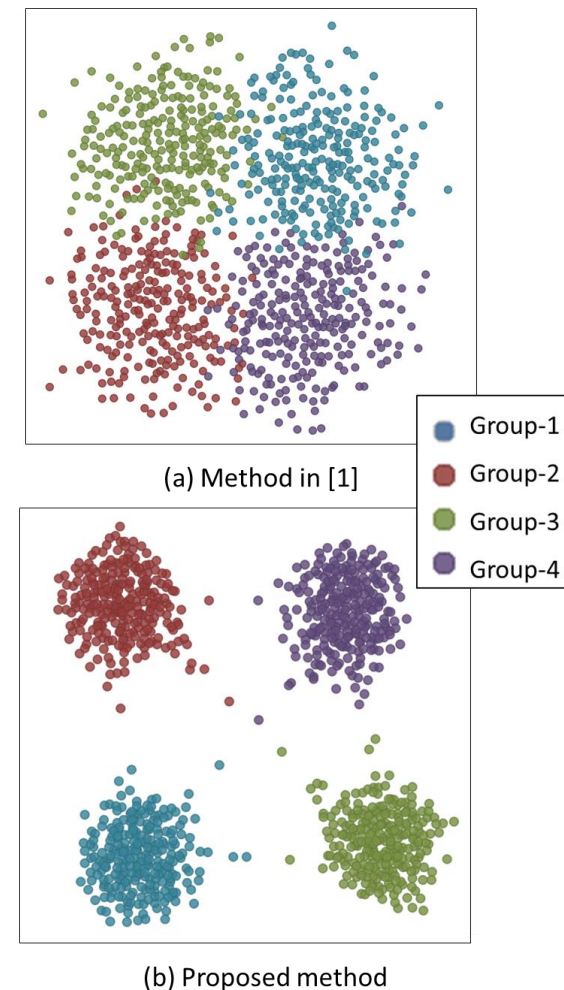
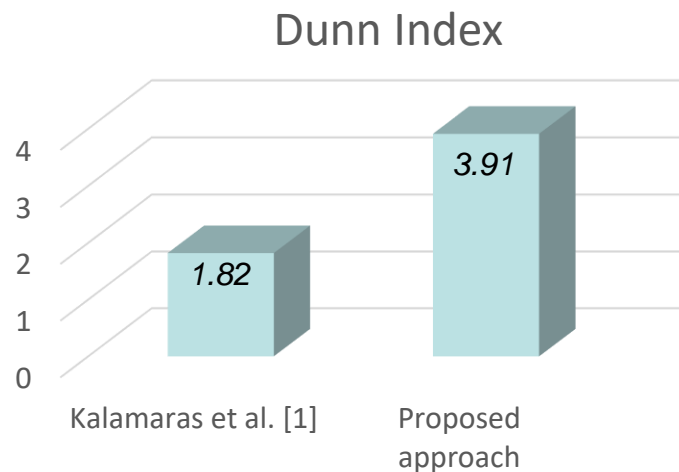
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Experimental results (1/2)

- Simulation of different behavioral groups:

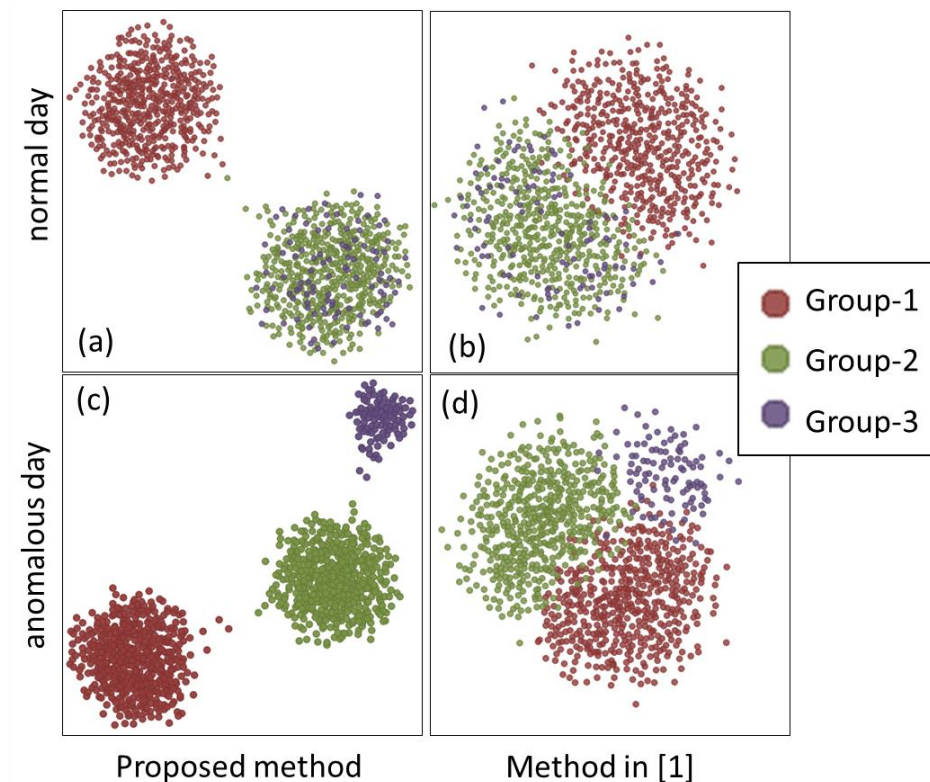
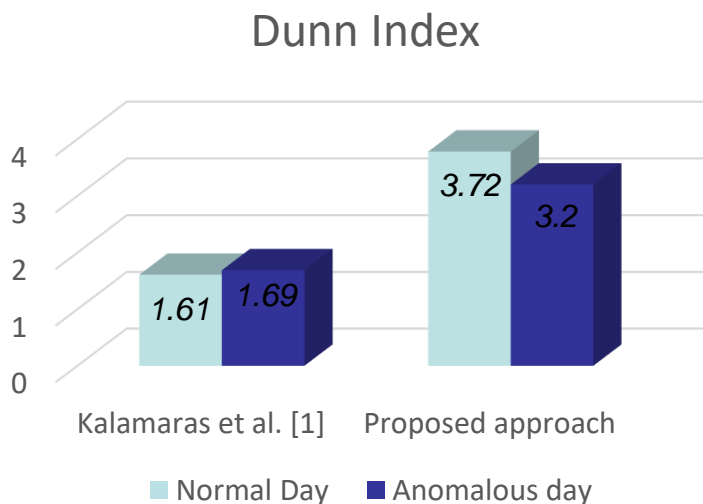
Group ID	Short description
Group-1	250 users with normal SMS, and normal Call behaviour
Group-2	250 users with high SMS, and normal Call behaviour
Group-3	250 users with normal SMS, and high Call behaviour
Group-4	250 users with high SMS, and high Call behaviour



Experimental results (2/2)

- Simulation of different behavioral groups for 7 days
- First 6 days: normal behavior, 7th day: anomalous group emerges

Group ID	Short description
Group-1	500 users with normal SMS, and normal Call behaviour
Group-2	500 users with high SMS, and normal Call behaviour
Group-3	100 users (anomalous users active in only the last day of the simulation) with anomalous SMS behaviour, and normal Call behaviour



[1] Kalamaras et al., "A multi-objective clustering approach for the detection of abnormal behaviors in mobile networks," ICCW 2015

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- Proposed a method of behavioral analytics for securing mobile networks
- Extension of previous approach → using graph descriptors
- Advantages:
 1. No feature engineering → **scenario agnostic**
 2. Can be used for **clustering** of entities based on their behavioral characteristics
 3. Graph nodes do not need to represent network entities, e.g. they can represent timestamps, slices etc. → **generic**
- Future work:
 - Apply anomaly detection to extract an anomaly label for each mobile device
 - Further 5G network simulations



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