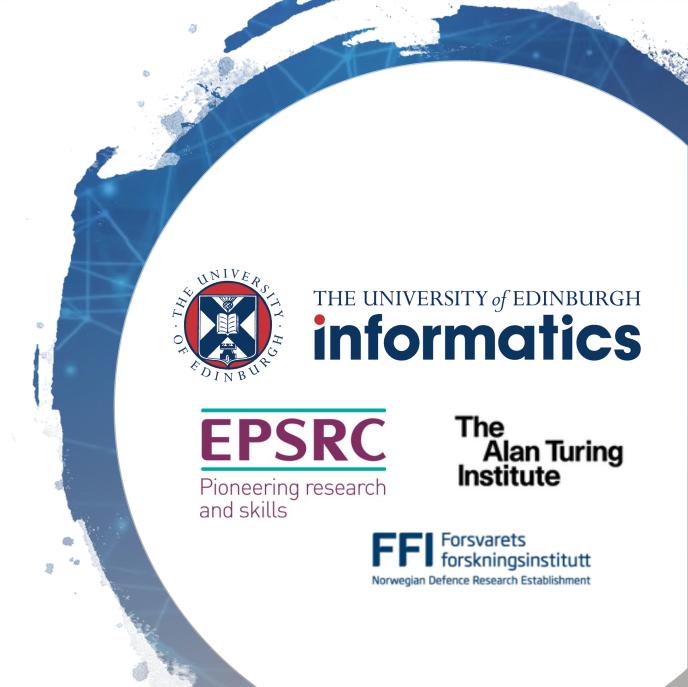
Henry Clausen, David Aspinall, Gudmund Grov, Marc Sabate

Better anomaly detection for access attacks using deep bidirectional LSTMs



### Contribution

### Novel deep LSTM-model:

- Designed for access attacks
- Flow-based
- Significantly improves detection rates

### Careful in-depth evaluation

- Comparison to SoA-models
- Longterm evaluation
- AUC-scores and det. Rates



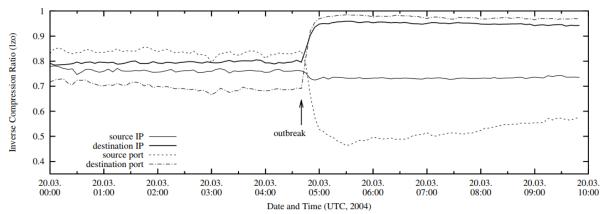
# Where network anomaly-detection works

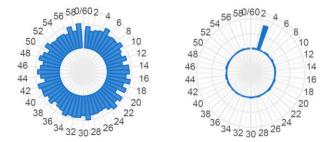
- DoS attacks
- Network probing
- Worms
- User active at strange times

#### Nisioti et al. (2018):

- Remote2Local & User2Root far less reliably detected
- Evaluation pitfalls make comparison difficult

Figure 1. Blaster - TCP address parameter compressibility







# Underlying idea

$\operatorname{Src}$	Dst	DPort	bytes #	packets
A	В	80	247956	315
A	В	80	7544	13
A	В	80	328	6
A	В	80	2601	10
A	В	80	328	6
A	В	80	328	6
A	В	80	380	7
A	В	80	328	6

:

SQL-injection-attack, CICIDS-17 data

$\operatorname{Src}$	Dst	DPort	bytes #	packets
D	C	N33	600	5
$\mathbf{C}$	D	445	77934	1482
D	$\mathbf{C}$	N33	600	5
$\mathbf{C}$	D	445	5202	10

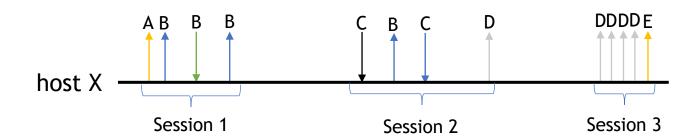
Benign SMB, LANL-16 data

$\operatorname{Src}$	Dst	DPort	bytes #	packets
$\overline{\mathbf{C}}$	D	445	4106275	2830
$\mathbf{C}$	D	445	358305611	242847

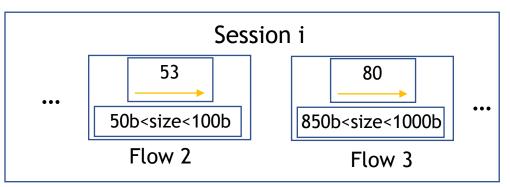
Malicious SMB, LANL-16 data



- sort outgoing and incoming connections on host X
- group them into intervals
  - flow separation less than 8s

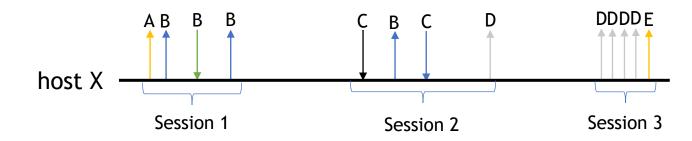


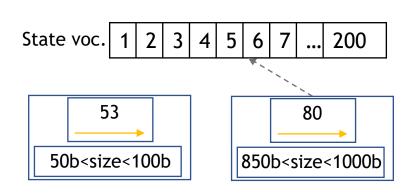
- Tokenise flows:
  - Direction
  - TCP/UDP/ICMP
  - Port
  - Size interval





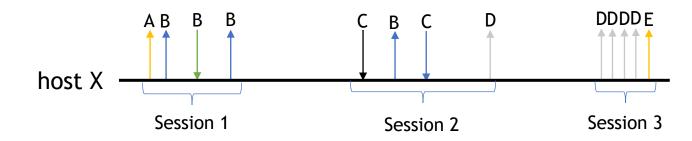
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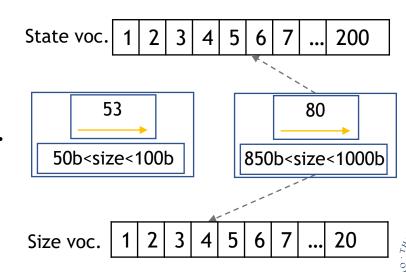






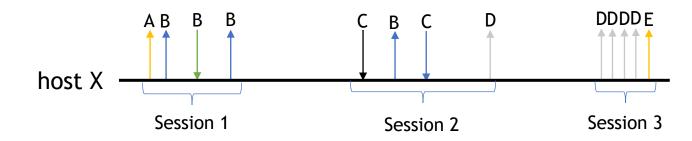
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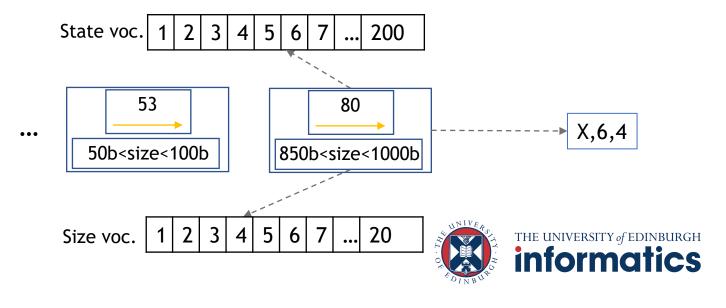




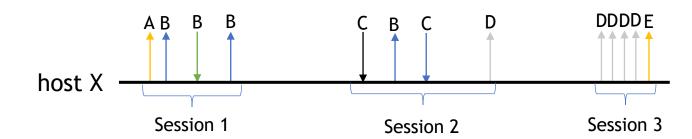


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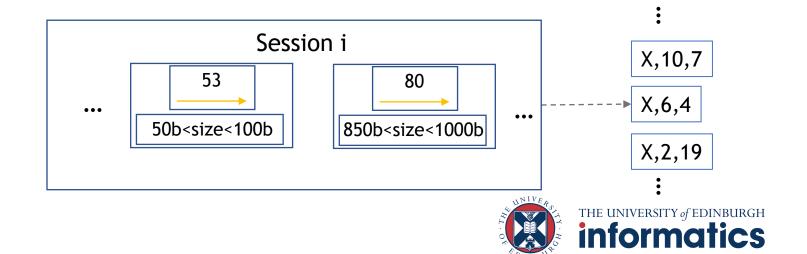




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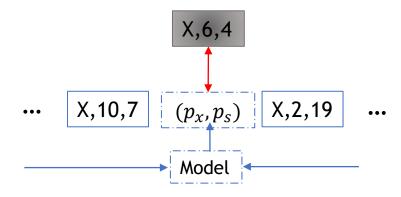


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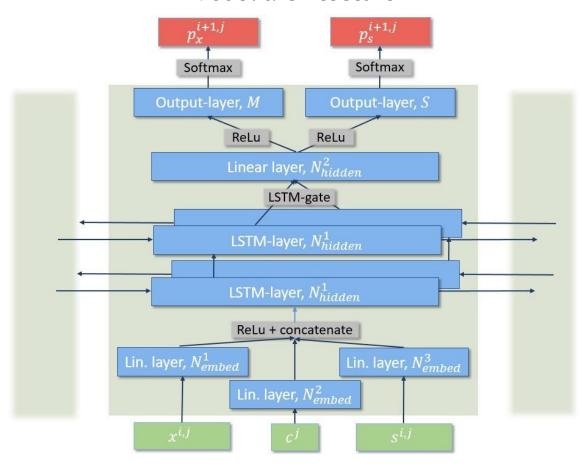
# Modelling - Architecture

Leave-one-out prediction training



Anomaly-score: averaged likelihood of session

#### Model architecture





# Modelling - Architecture

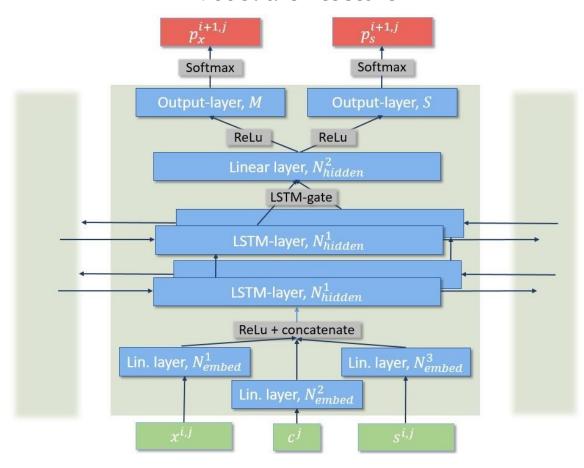
### Embedding layer

- separately to reduce parameters
- Two LSTM layers
  - **Both directions**
- Linear layer to postprocess
  - Softmax-output for state and size

• 
$$N_{hidden}^{1,2} = 50$$
  
•  $N_{embed}^{1,2,3} = 5$ 

• 
$$N_{embed}^{1,2,3} = 5$$

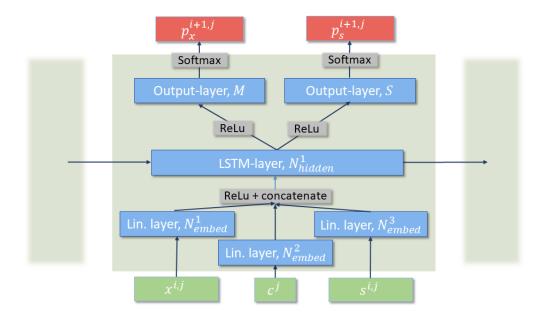
#### Model architecture



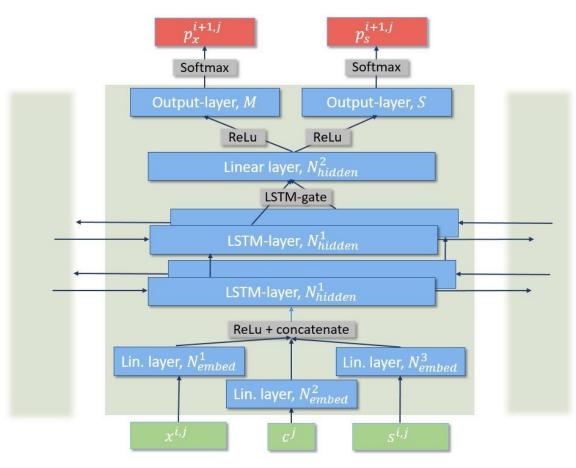


## Modelling - Architecture

### Shallow comparison architecture



#### Model architecture





# Datasets and comparison

- CICIDS-17
  - 7 access attacks
  - SQL-i., Heartbleed, XSS, ...

- UGR-16
  - 6 months
  - longterm evaluation

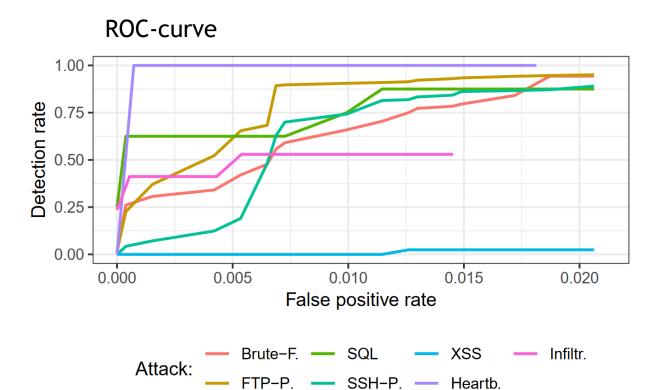
### SoA-models:

- UNIDS (2013)
  - Clustering-based
  - Best access-attack detection rates in survey
- Radford et al. (2018)
  - LSTM-based
- Niyaz et al. (2016)
  - Deep autoencoder



### Evaluation





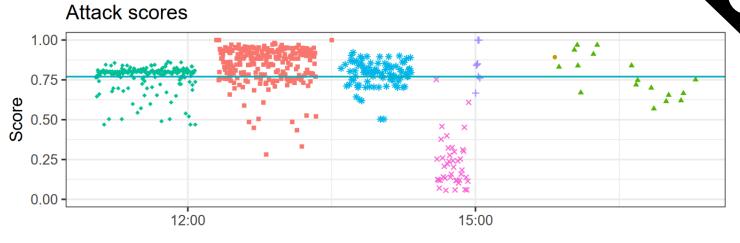
Attack:

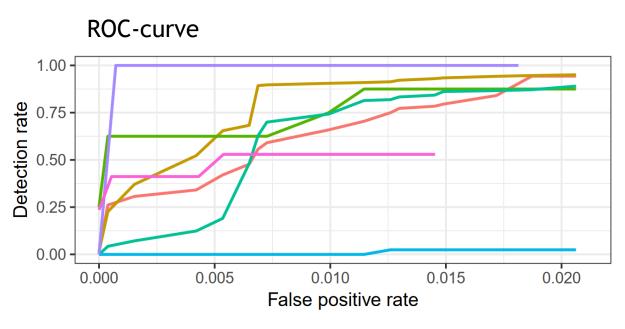
FTP-P Infilt. \* Brute-F. × XSS

Heartbl. • SSH-P. + SQL



### **Evaluation**







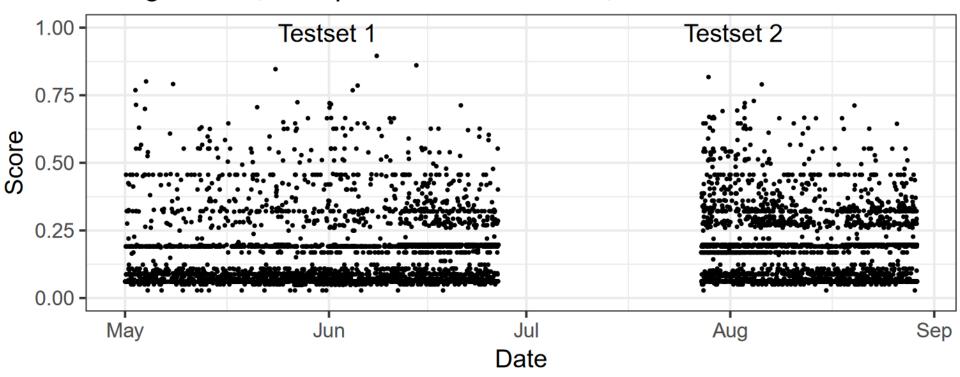
A Harale	FTP-P	<b>A</b>	Infilt.	*	Brute-F.	×	XSS
Attack:	Hearthl		SSH-P	4	SOL		

	1-AUC scores							
	Our model	UNIDS	Radford	Niyaz	shallow m.			
Brute Force Web	0.016	0.49	0.027	0.32	0.048			
FTP-Patator	$\boldsymbol{0.0025}$	0.011	0.0048	0.16	0.0052			
Heartbleed	0.0003	0.0057	0.032	0.077	0.012			
Infiltration	0.046	0.033	0.35	0.15	0.11			
SQL-injection	0.005	0.44	0.497	0.39	0.019			
SSH-Patator	0.009	0.013	0.035	0.011	0.005			
XSS	0.127	0.02	0.03	0.16	0.13			
Average	0.044	0.144	0.135	0.18	0.091			



# Evaluation - long-term stability

Benign traffic, Computer 42.219.154.44, UGR-16





### Limitations

Traffic overlay

• Events separated in time

Isolated flow events



# Thank you for your attention

IN SECURITIES



(C) AMANDA ROUSSEAU



### Limitations

Traffic overlay

• Events separated in time

Isolated flow events



### Conclusion

- Large public dataset
  - Realistic interactions
  - Evasive tactics
  - github.com/detlearsom/detgen/stepping-stone-data
- Evaluation of current state-of-the-art
  - Lower overall detection rates
  - Lack of robustness against chaff
  - Watermarking and deep-learning performs best

