

Domain Generation Algorithms (DGAs)

- First time detected in the Conficker malware family [2]
- A DGA generates domain names following behaviour similar to a pseudo-random number generator. These domain names are known as *Algorithmically Generated Domains* (AGDs)
- Few examples of AGDs [3]: atuquhuswcvjehti.com, test.takemuchexpression.club, ecfdb206098b6b12d069f58e4da6d66c5f2.cc, gzauh44cvh14f52i35m29m29crn10g63dqd30e51c39.biz

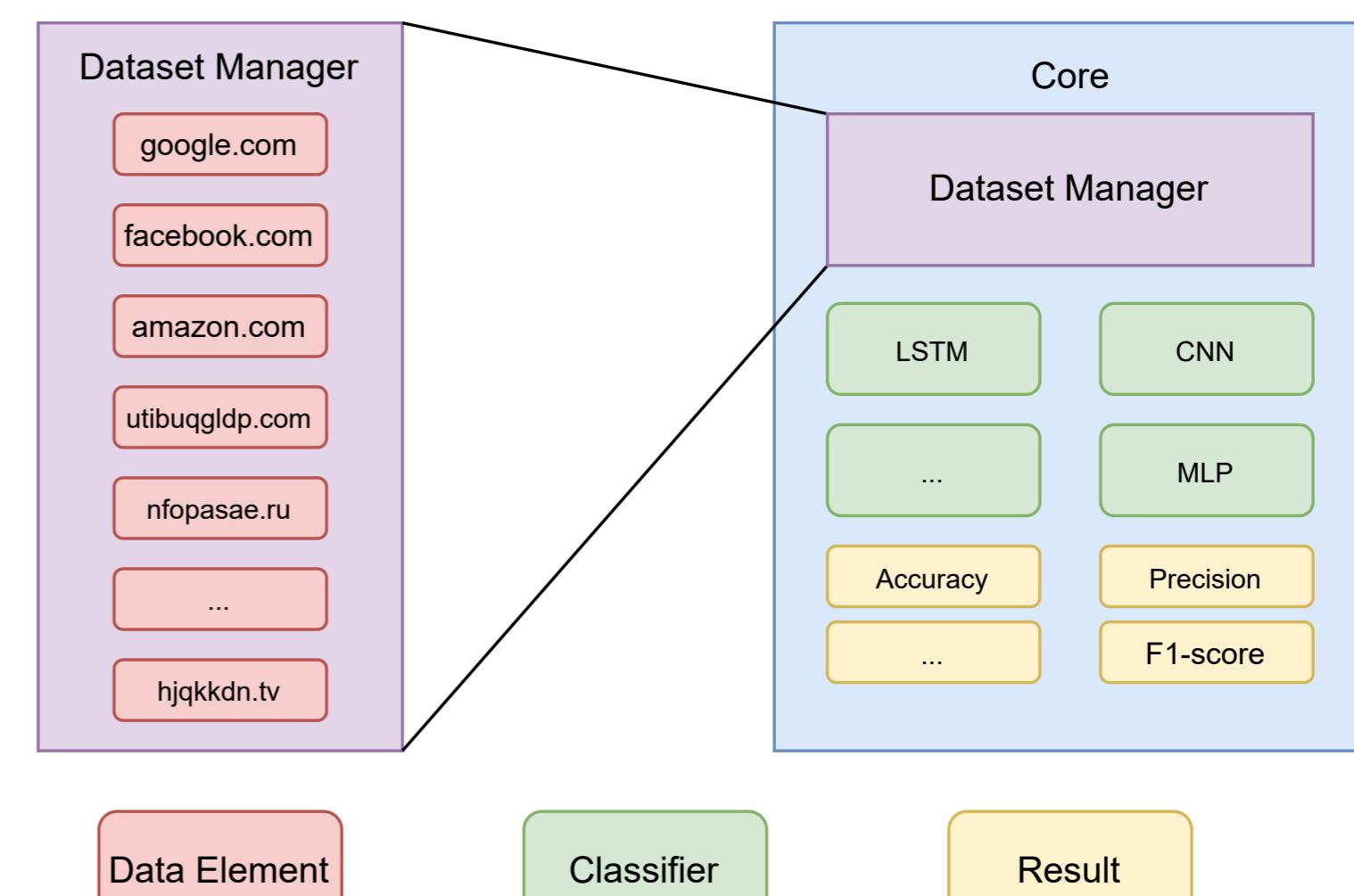
Detection Techniques

- Different solutions in the literature, the most recent ones adopting machine learning and deep learning approaches
- Most studies are based on the analysis of the domain name only
- There is no common methodology for comparing models with each other, even though the datasets follow the same structure

Preliminary Results

Model (Year)	Acc	Prec	Rec	F1	FPR	TPR	MCC	κ
LSTM [5] (2016)	95.42	97.39	95.69	96.53	5.12	95.69	89.82	0.8045
LSTM [7] (2017)	95.44	97.25	95.87	96.55	5.40	95.87	89.84	0.8059
CNN [7] (2017)	94.96	97.39	94.98	96.17	5.07	94.98	88.86	0.7849
LSTM [6] (2018)	95.02	96.82	95.67	96.24	6.27	95.67	88.88	0.7896
CNN [6] (2018)	92.94	96.29	92.99	94.61	7.16	92.99	84.49	0.7056
CMU [8] (2018)	94.87	97.46	94.77	96.10	4.92	94.77	88.69	0.7810
MIT [8] (2018)	95.48	96.96	96.23	96.59	6.03	96.23	89.87	0.8083
Parallel CNN [8] (2018)	93.48	96.64	93.48	95.03	6.49	93.48	85.68	0.7265
Baseline [8] (2018)	86.51	93.36	85.87	89.46	12.19	85.87	71.31	0.4745
MLP [8] (2018)	92.59	96.41	92.32	94.32	6.86	92.32	83.84	0.6907
CNN [1] (2019)	95.28	97.08	95.81	96.44	5.76	95.81	89.48	0.7998
Max Pooling [1] (2019)	90.48	95.62	89.84	92.64	8.21	89.84	79.53	0.6107
LSTM [1] (2019)	92.40	96.98	91.44	94.13	5.68	91.44	83.67	0.6804
LSTM + CNN [1] (2019)	83.88	94.12	80.87	86.99	10.09	80.87	67.44	0.3796
Bidireccional [1] (2019)	93.40	95.92	94.10	95.00	8	94.10	85.33	0.7261
DBD [4] (2019)	94.19	96.92	94.28	95.58	5.98	94.28	87.18	0.7545

Acc: Accuracy; Prec: Precision; Rec: Recall; F1: F1-score; FPR: False Positive Rate; TPR: True Positive Rate; MCC: Matthews's Correlation Coefficient; κ : Cohen's Kappa Score



- Core module includes all execution logic
- Dataset Manager manages dataset processing
- New models must adhere to the Classifier schema
- Data Element and Result represent training data and evaluation metrics, respectively. They are user-defined

Conclusions

- Our framework allows anyone to train and compare models in a simple and fast way
- Simpler models (but Baseline) tend to achieve better results when considering a large number of different malware families
 - They need to generalize more
 - More robust to detect DGAs from different families

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Acknowledgements

Try It!



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