

Embedded Malware Detection using Markov n-grams

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Agenda

Introduction to problem domain

Mathematical Modeling

Discussion on Results

Conclusions

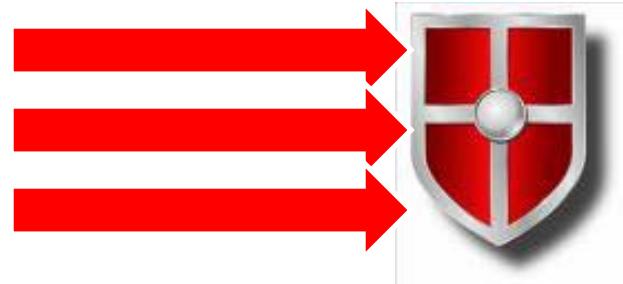
Future Work



Introduction to Problem Domain



Introduction



Problem with State-of-the-art antivirus

Signature matching

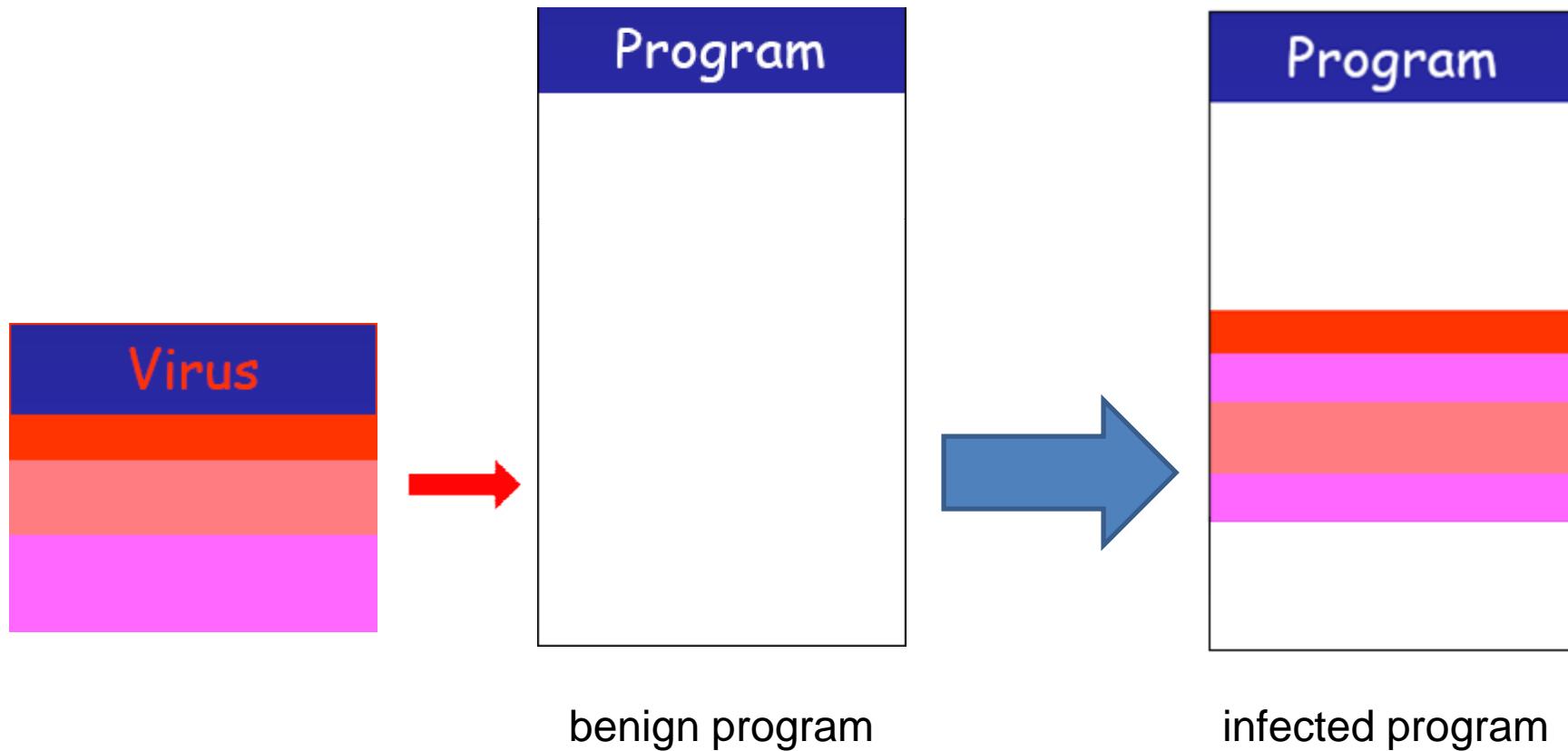
- Identify sequence of instructions unique to a virus => virus signature
- Match program to a database of signatures

Problems

- Inability to detect zero-day attacks
- Scan only starting portions of files due to high overhead and false alarms; vulnerable!
- Size of signature database cannot scale in future



Embedded Malware



Mathematical Modeling



Our Approach

Anomaly detection

- Differentiate anomalous behavior from the normal workflow
- Primarily utilize *statistical modeling*

Statistical model of benign DOC, EXE, JPG, MP3, PDF, ZIP files using Markov n-grams

Feature extraction using well known information-theoretic measure, entropy rate

Threshold based detection using Gaussianity of sum of sampled entropy rate distribution



n-gram analysis

n-gram Definition [en-wikipedia]

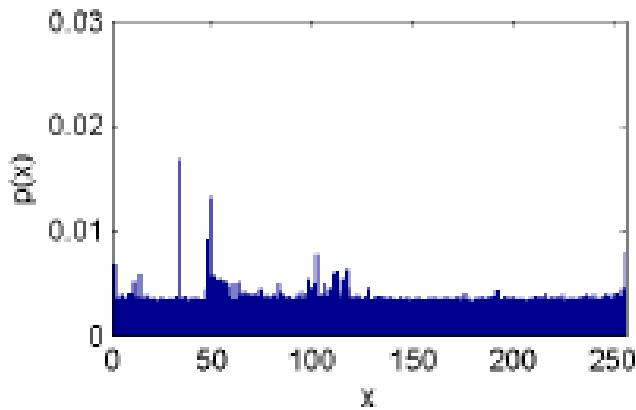
- An n -gram is a sequence of n symbols in a given sequence

11100111011001110011010011010010111000111111110
11100111011001110011010011010010111000111111110
11100111011001110011010011010010111000111111110

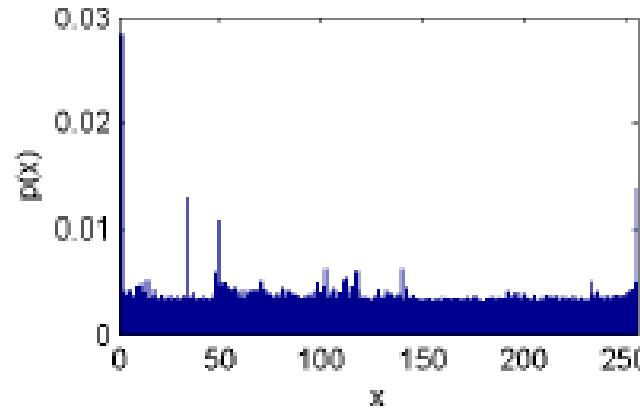
3-gram



Whole file n-gram analysis



1-gram benign PDF



1-gram infected PDF

Whole file, 1-gram analysis DOES NOT show significant perturbations 😞



Block wise 1-gram analysis

Calculate Mahanalobis distance between benign model and given file in a block wise manner (blocksize = 1000bytes)

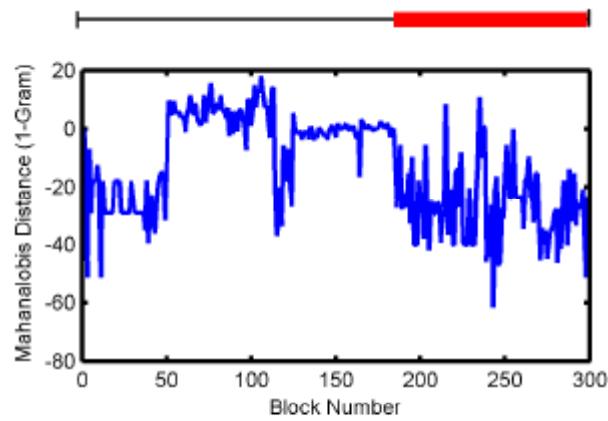
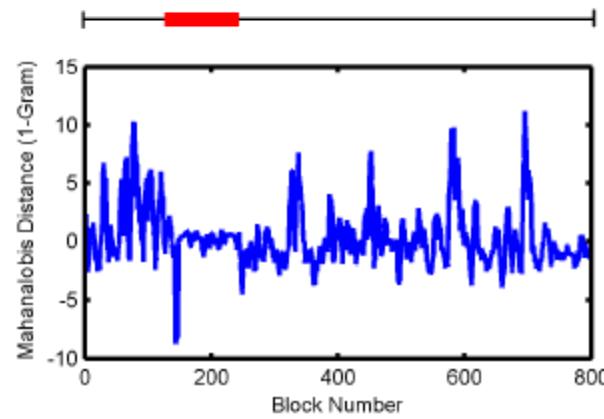
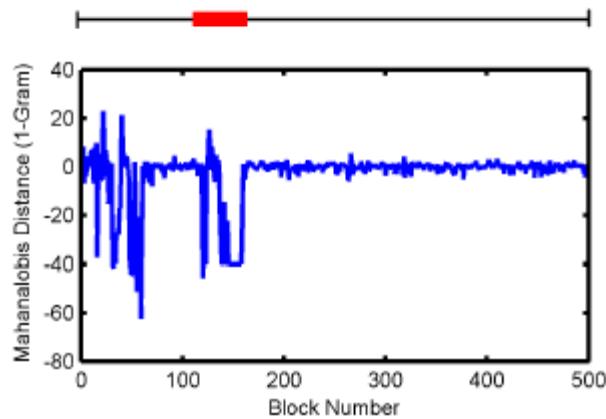
Benign model distribution is:

- average byte value frequency
- their standard deviation

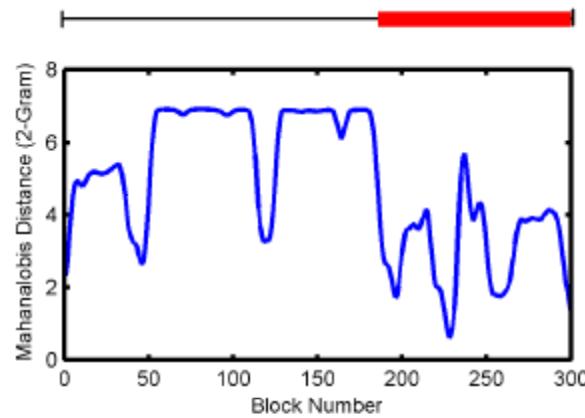
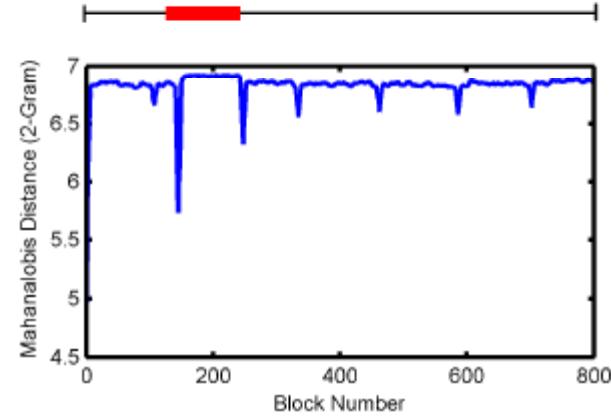
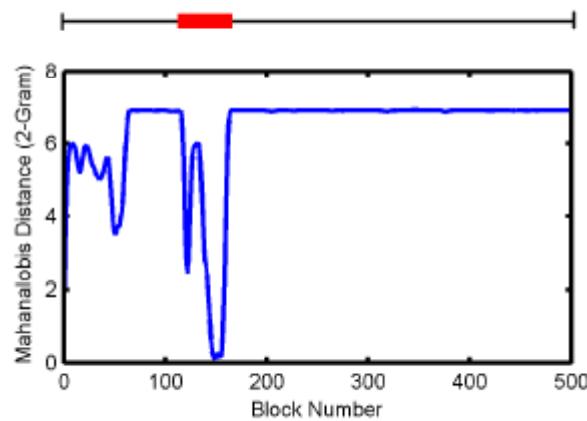
$$D(x, y) = \sum_{i=1}^{n-1} \frac{|x_i - y_i|}{\sigma_i + \alpha}$$



Block wise 1-gram analysis



Block wise 2-gram analysis



Analysis of results

2-gram is better than 1-gram; at the cost of exponentially higher computational complexity!

Cannot increase n due to a fixed block size of 1000 bytes!

2-gram distribution is a joint distribution

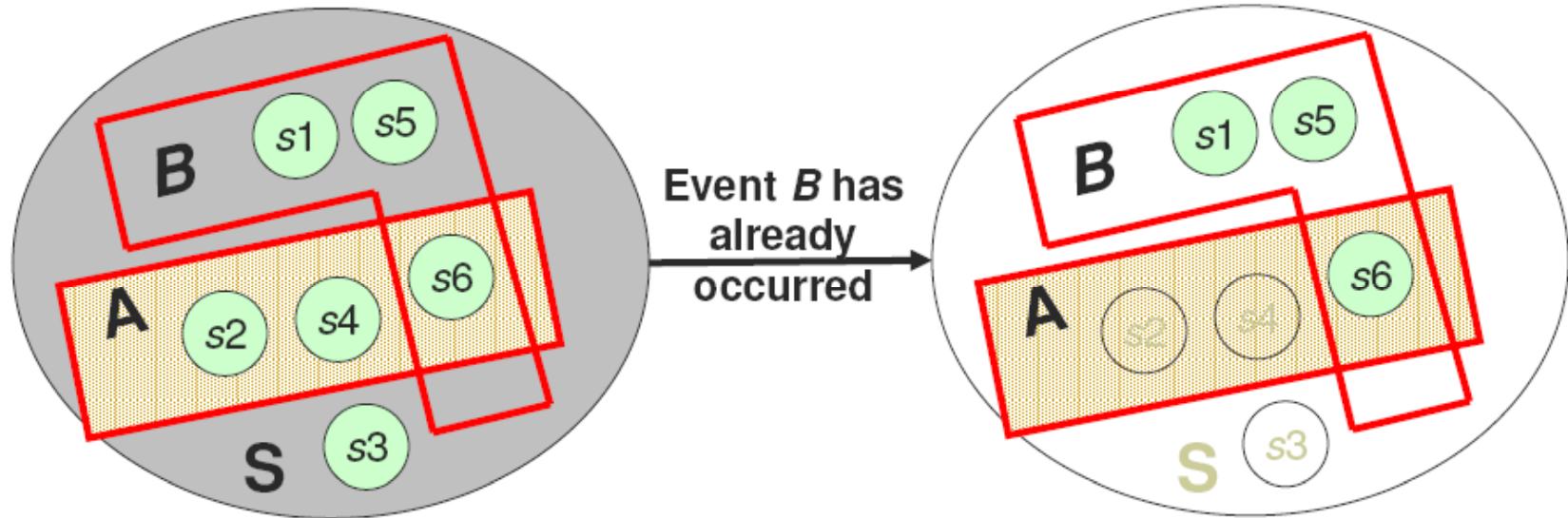
Some redundant information in joint distribution that can be removed using *conditional distribution*. **HOW?**



Advantage of Conditional Distribution

$$Pr\{A|B\} = \frac{Pr\{A \cap B\}}{Pr\{B\}}$$

Reduction in sample space



From Conditional Distribution to Markov Chain

A conditional distribution can be directly mapped to a Markov chain

Symbols of distribution mapped to states in Markov Chain

How to select order of Markov chain?

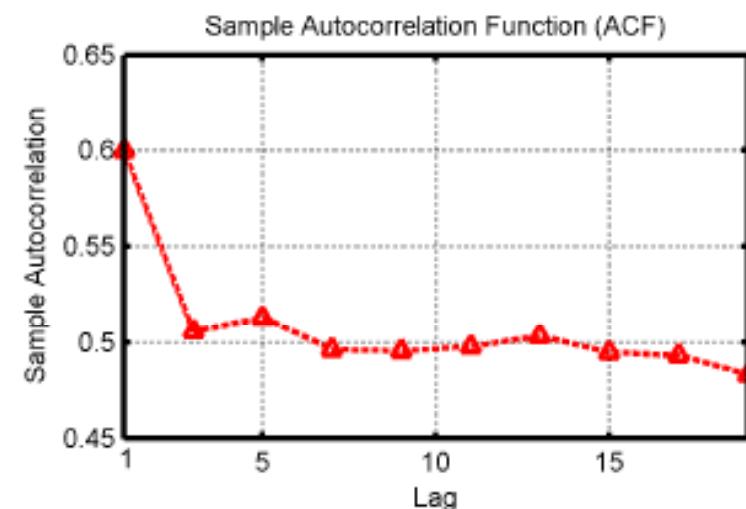
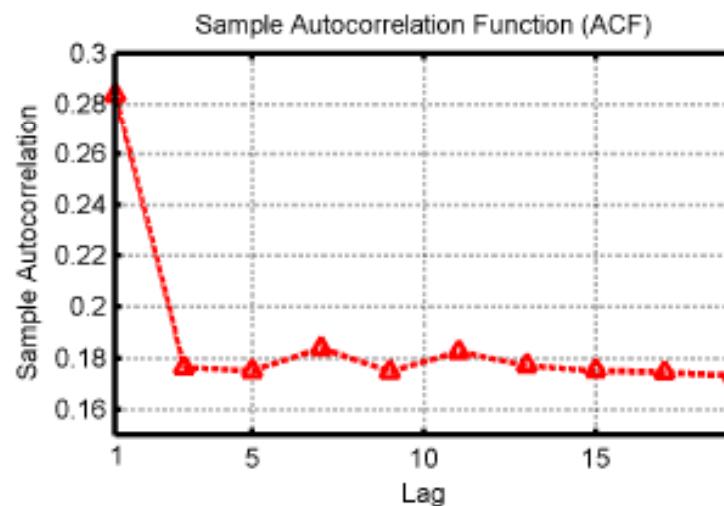
Byte level autocorrelation!



Byte-level autocorrelation results

Random information beyond lag=1

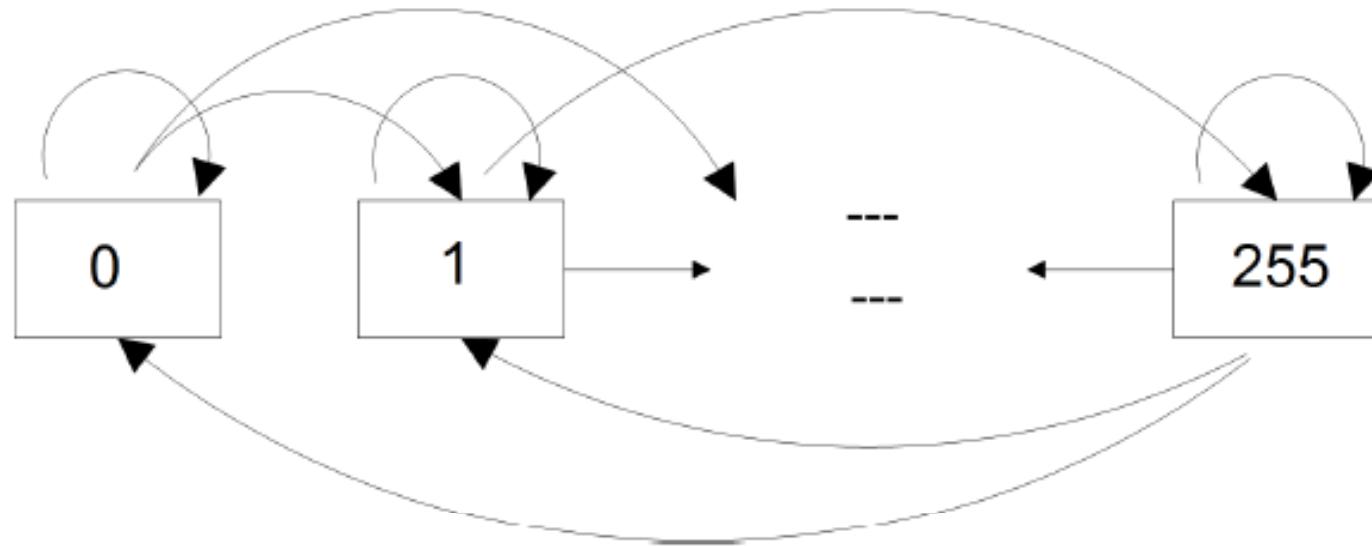
1st order Discrete Time Markov Chain



Markov n-gram Model

1st order, 256 state Markov Chain

Can be constructed using conditional 2-gram distribution



Quantification feature

average information in the Markov Chain

Entropy rate gives

- Time density of average information in the Markov chain

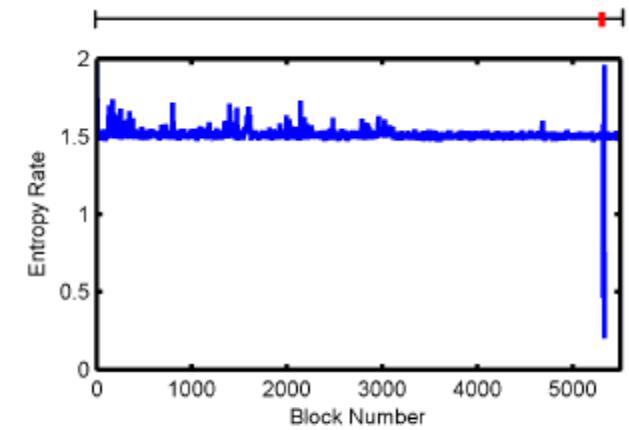
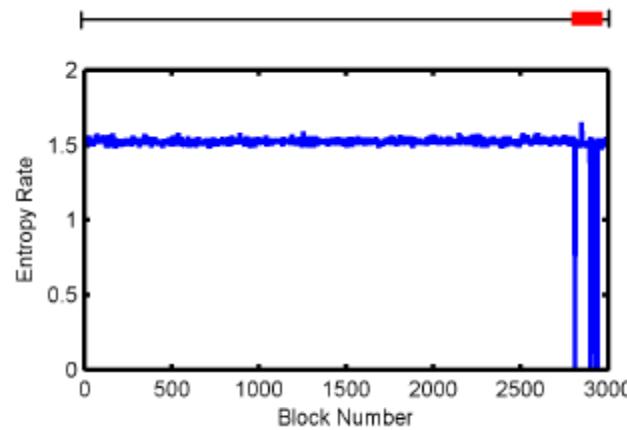
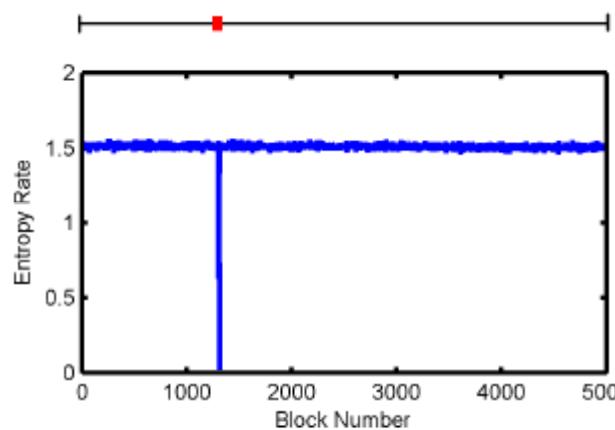
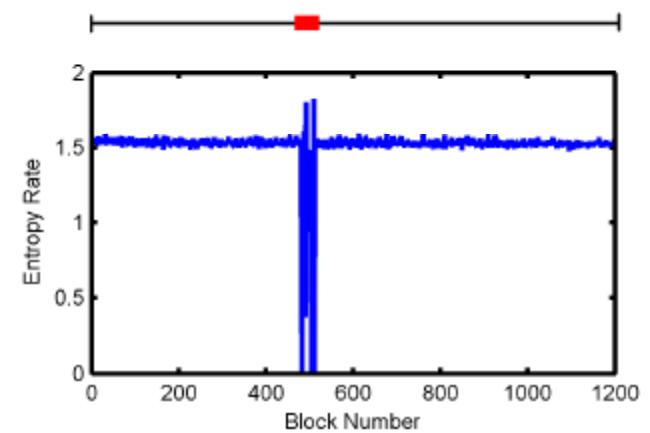
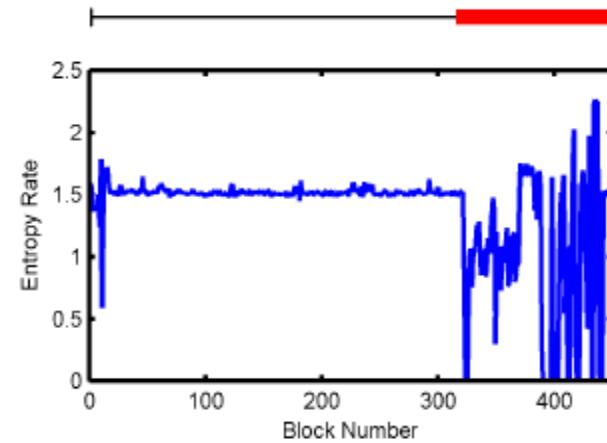
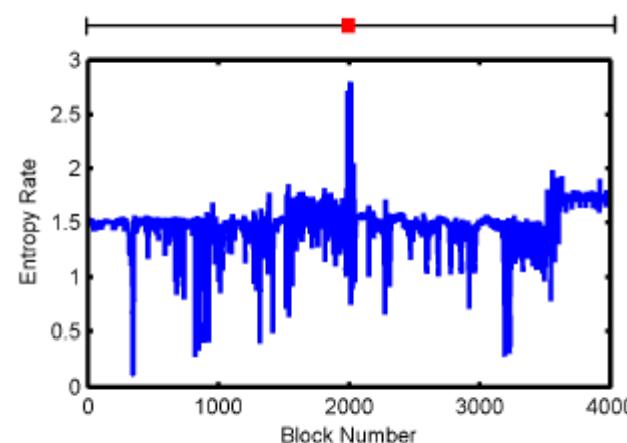
$$R = \lim_{N \rightarrow \infty} \frac{H(X_1, X_2, \dots, X_n)}{N}$$

For a 256 state Markov chain

$$R = \sum_{i=0}^{255} \pi_i H(X_i)$$



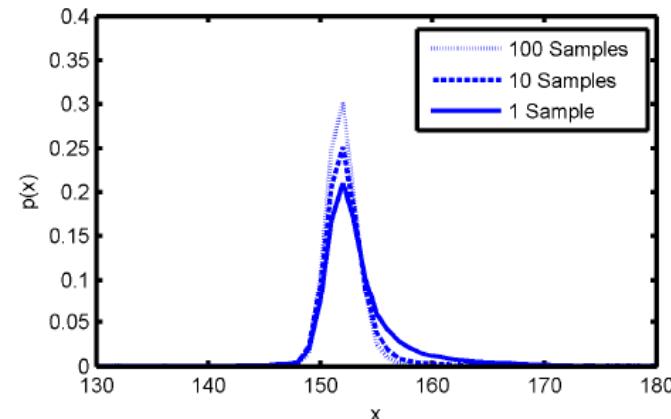
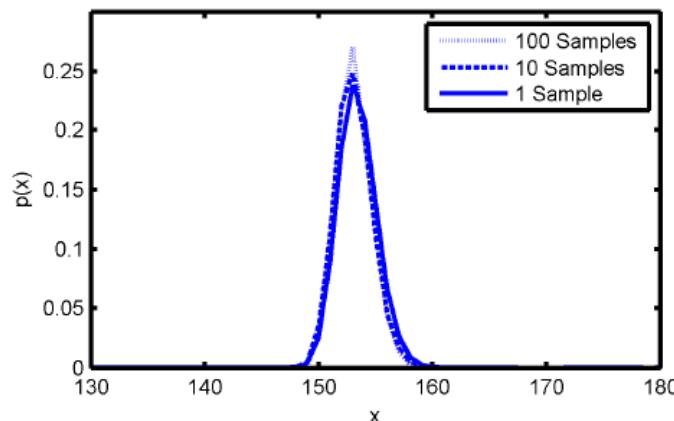
Entropy rate Results



Threshold selection

Sum of sampled entropy rate distributions approach Gaussianity*

Threshold selected at $\mu \pm 5\sigma$ (99.99%)



* Direct consequence of central limit theorem



Results

	Mahalanobis n-gram Detector (%)	Markov n-gram Detector (%)	Percentage Improvement (%)
MP3			
TP rate	63.8	95.0	31.2
FP rate	32.3	0.2	32.1
JPG			
TP rate	76.3	95.4	19.1
FP rate	35.7	2.7	33.0
PDF			
TP rate	75.4	84.5	9.1
FP rate	46.8	31.8	15.0
ZIP			
TP rate	60.0	90.4	30.4
FP rate	29.9	8.3	21.6
EXE			
TP rate	54.1	84.9	30.8
FP rate	47.3	16.7	10.6
DOC			
TP rate	65.6	66.3	0.7
FP rate	48.8	29.2	19.6

Conclusion & Future Work



Conclusion

Advantages

- Ability do identify location of malware
- Significantly improved *TP rate* and *FP rate* as compared to Mahanalobis detector

Limitations

- Still slightly high false positive rate for DOC and PDF files
- Embedded Malware -> dormant malware
- Mimicry Attacks



Future Work

How to reduce false positive rate?

Complement with signature based detector

- Inability to detect zero-day attacks 😞

Multiple features and correlation

- Initial results have shown promise 😊
- Subject of forthcoming journal publication...

Patent pending on current work



Thank you!



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