

Web Server Anomaly Detection using Principle Component Analysis

NECTEC-
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Agenda

- **Introduction**

 - * **Motivation**

 - * **Features**

 - * **Access log and Error log**

 - * **PCA**

- **Methodology**

- **Experiment I: Determining Useful Dimensions**

- **Experiment II: Anomaly Detection**

- **Future Works**

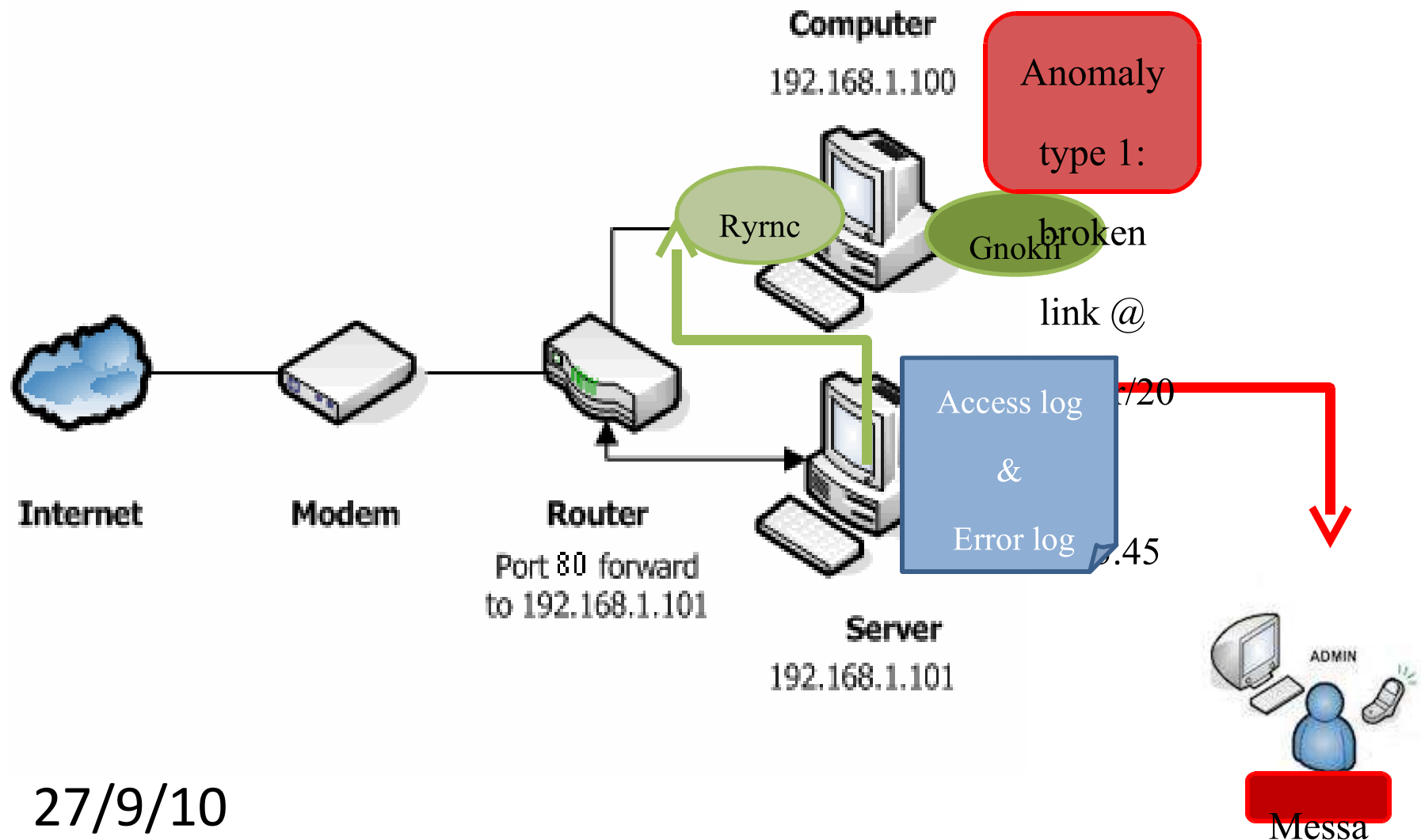
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Introduction: Motivation

- Network Monitoring Systems usually used host-specific thresholds to justify host status e.g. threshold on PING round trip time
- NTL teams are developing a NMS called NetHAM.
- One of the goal is to let NetHAM analyzes network hosts and their services without using host-specific threshold.
- We start at one of the most common internet services i.e. HTTP.
- Here comes this experiment...

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Introduction: Motivation



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Introduction: Access log & Error log

Access Log : The server access log records all requests processed by the server.

```
202.44.135.34 - - [09/Sep/2003:20:30:11+0700] "Get /PROJECT/member.html HTTP/1.0" 200 19037
```

1 2 3 4 5 6 7 8 9

Error Log : The server error log records any errors that it encounters in processing requests.

```
[Fri Jun 20 22:19:44 2003] [error] [client 127.0.0.1] Invalid method in request\x80L\x01\x03
```

1 2 3 4

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Introduction: Features

The ten features are as follows.

1. Number of unique clients (F1)
2. Number of request (F2)
3. Number of error events (Er)
4. Number of HTTP Unauthorized Access events (401)
5. Number of HTTP Not Found events (404)
6. Number of HTTP Forbidden events (403)
7. Number of HTTP Internal Error events (500)
8. Maximum number of requests per client (Max)
9. Minimum number of requests per client (Min)
10. Average number of request per client (Avg)

Please note! All ten features are calculated over a 5-minute interval

Introduction: Basic Idea

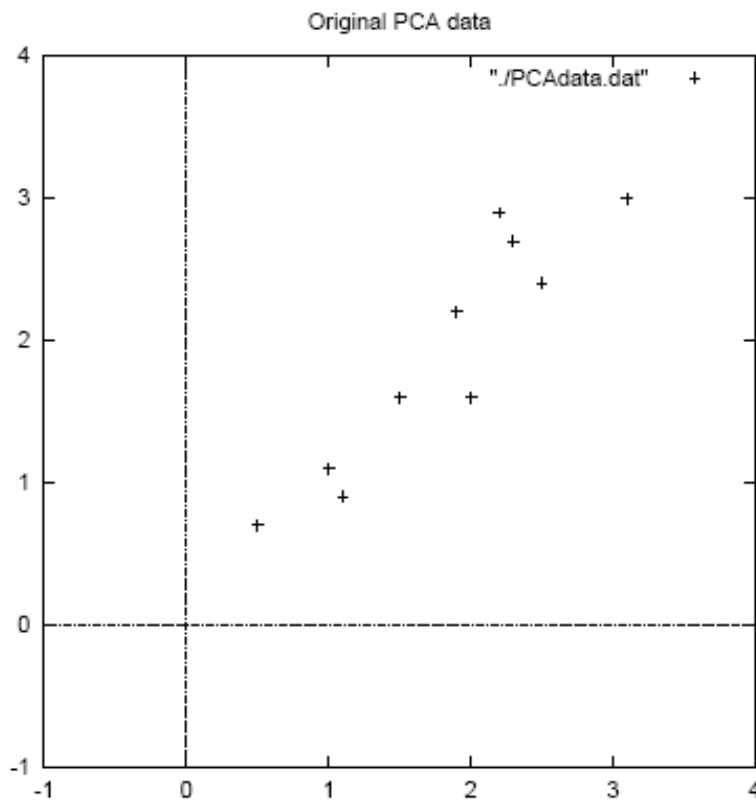
- Threshold alone may not be enough to indicate anomalous behavior because different servers may have different ranges of normality in such features.
- Need some mechanism to learn which pattern is normal and which pattern is abnormal.
- Here comes PCA...

Introduction: PCA

- Principal Component Analysis (PCA) is the technique used for identifying patterns in the data set in form of vectors.
- PCA transforms original data points to the new coordinate axes in order to express their similarities and differences.

Introduction: PCA

- A quick example



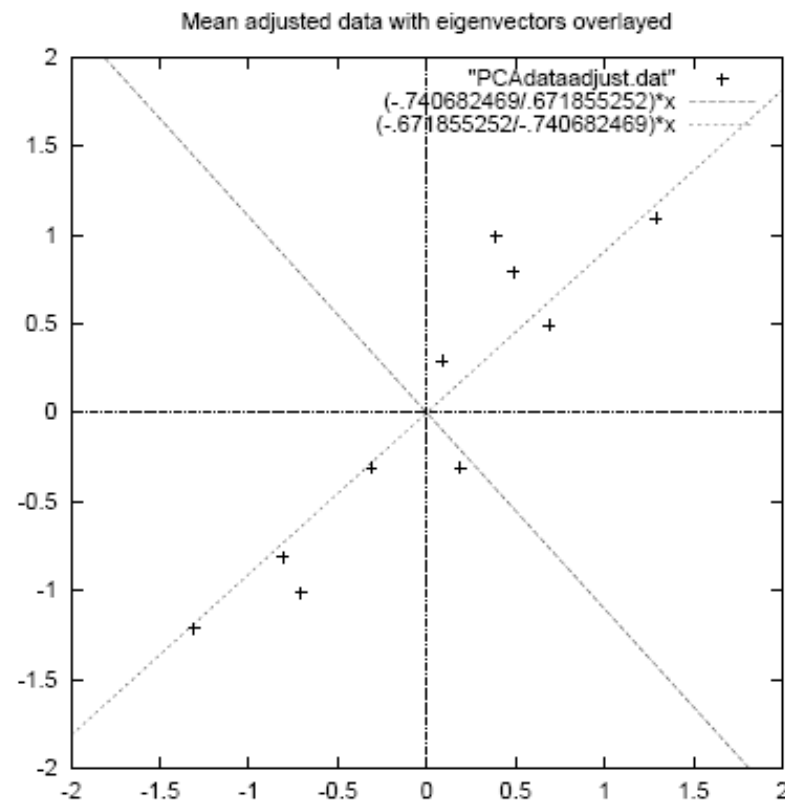
X	Y
2.5	2.4
0.5	0.7
2.2	2.9
1.9	2.2
3.1	3.0
2.3	2.7
2.1	1.6
1	1.1
1.5	1.6
1.1	0.9

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Introduction: PCA

- After some calculations, we have the new PCA axes

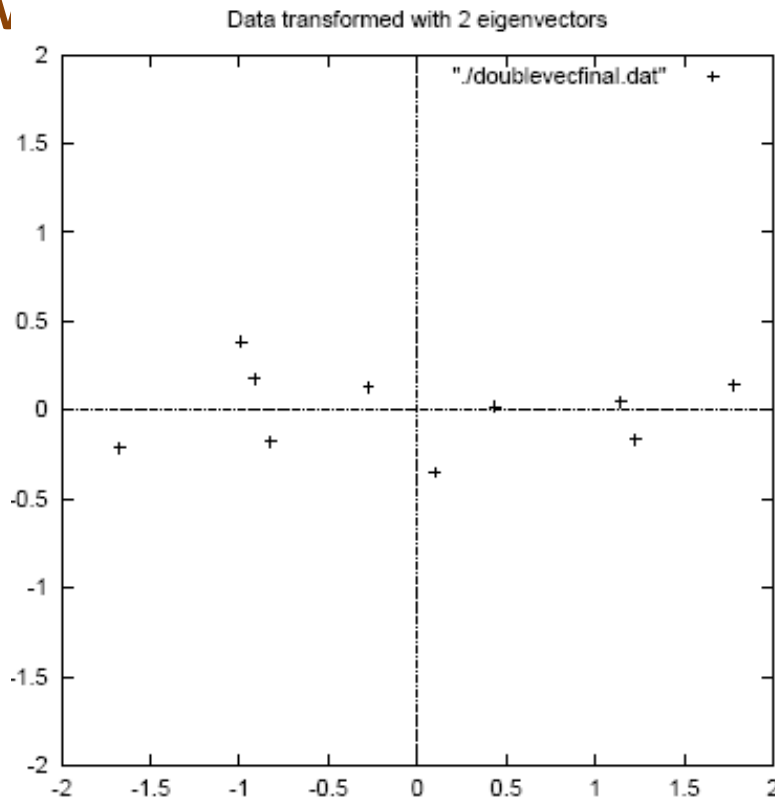


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Introduction: PCA

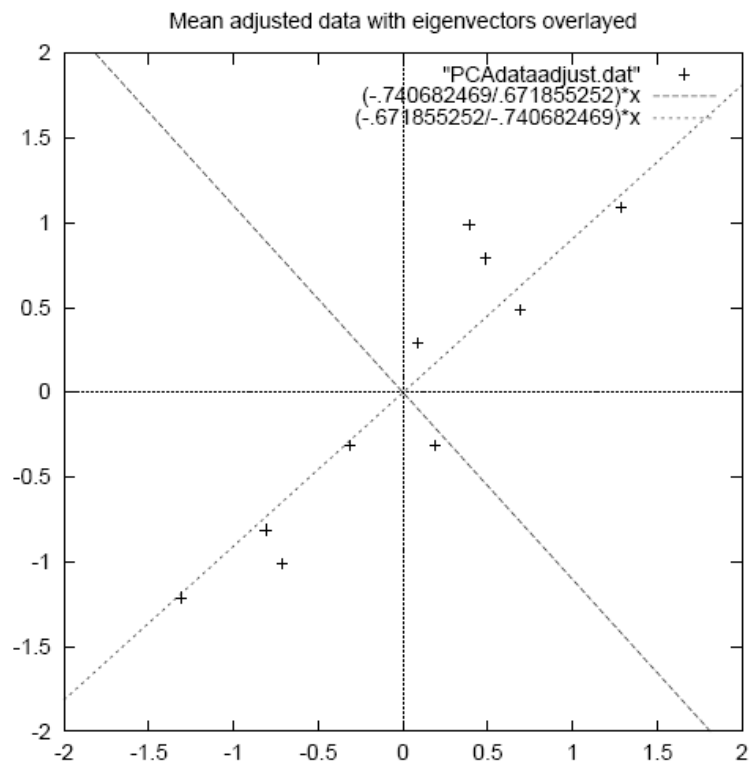
- Map all data points to the new axis to get the new set of coordinates



X	Y
-	-
0.827970186	-0.17511530
1.77758033	7
-	0.14285722
0.992197494	7
-	-
0.274210416	0.38437498
-1.67580142	9
-	-
0.912949103	0.13041720
0.099109437	7
5	-
1.144572111	1.20949846

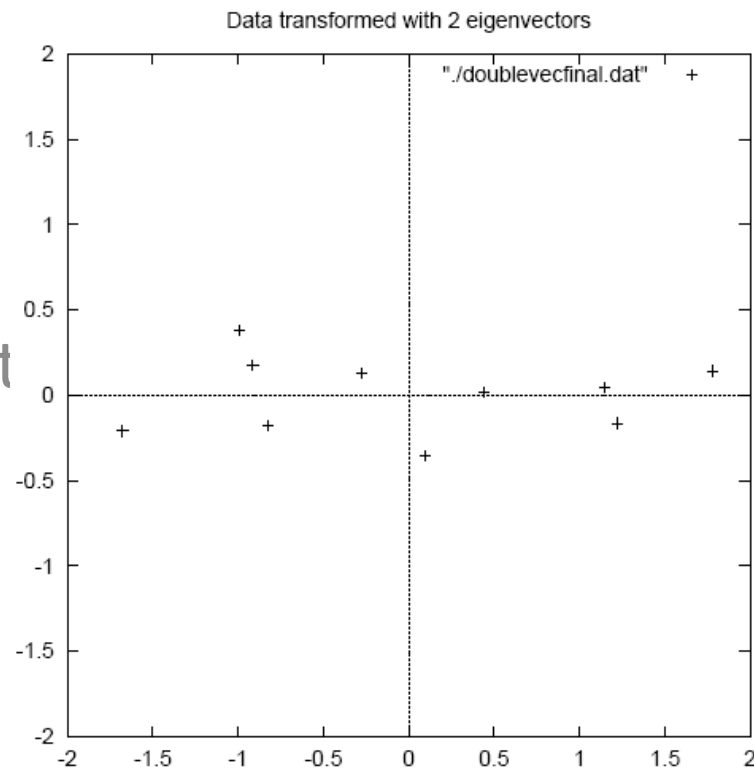
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Introduction: PCA



Original Data

subt



Final Data

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Agenda

- Introduction

 - * Motivation

 - * Features

 - * Anomalies Types

 - * PCA

 - * Access log and Error log

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- Methodology

- Experiment I: Determining Useful Dimensions

- Experiment II: Anomaly Detection

27/9/10 Future Works

Methodology

- Prepare the set of data in the form of 10 dimensional feature vectors obtained from the log files.
- Subtract the mean from each vectors
- Prepare the covariance matrix :

$$\begin{bmatrix} \text{Cov}(X_1, X_1) & \text{Cov}(X_1, X_2) & \dots & \text{Cov}(X_1, X_N) \\ \text{Cov}(X_2, X_1) & \text{Cov}(X_2, X_2) & \dots & \text{Cov}(X_2, X_N) \\ \vdots & \vdots & & \vdots \\ \text{Cov}(X_N, X_1) & \text{Cov}(X_N, X_2) & \dots & \text{Cov}(X_N, X_N) \end{bmatrix}$$

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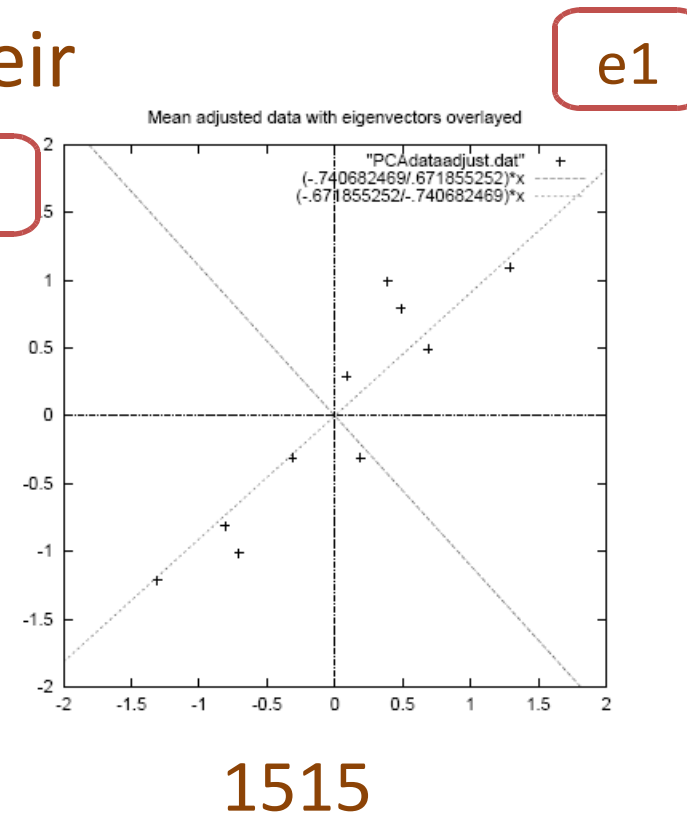
$$\text{Cov}(X, Y) = \sum \frac{(x_i - \bar{x})(y_i - \bar{y})}{N}$$

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Methodology

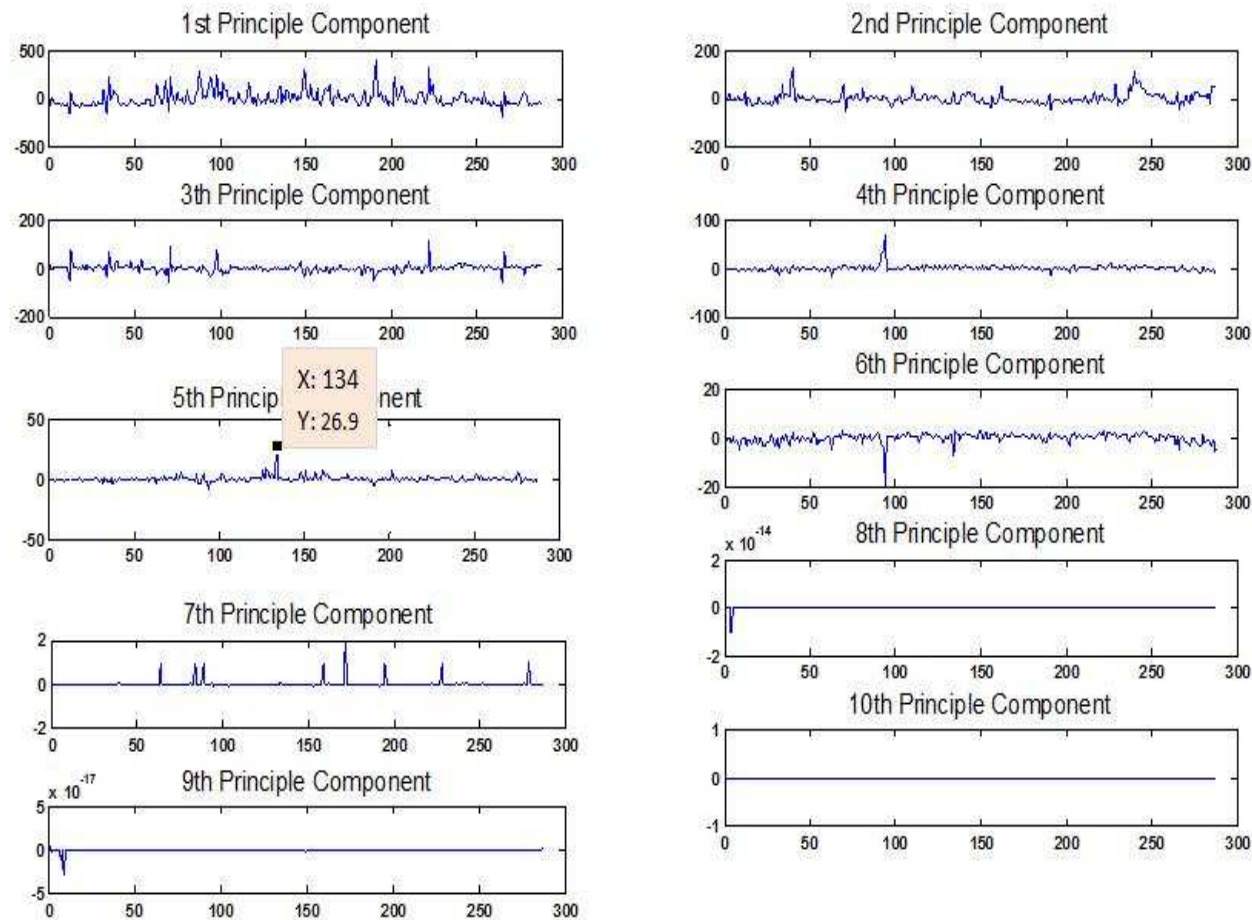
- Solve for eigenvectors and eigenvalues of the covariance matrix.
- Rearrange eigenvectors by their eigenvalues descending to obtain the new axes sorted by their significance (e.g. the eigenvector with the highest eigenvalue becomes the first principal

axis)



Methodology

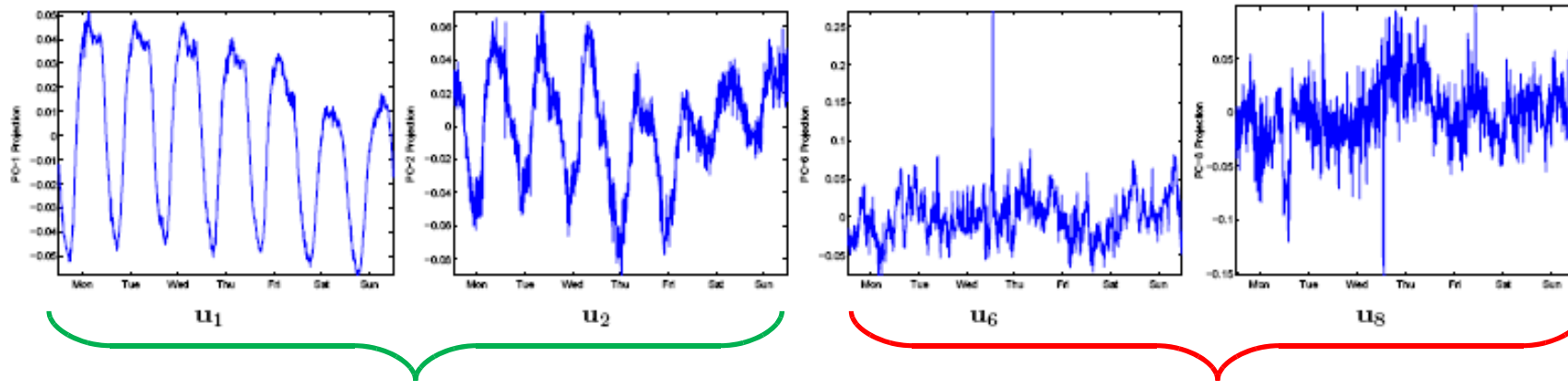
Project data vectors onto PCA axes.



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Methodology

- Apply PCA on this set of vectors.
- Select an appropriate separating axis between normal part and anomalous part. The first k -principal axis will form the normal subspace



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Methodology

- Let k be the dimension of the normal subspace. Pick the time bins whose residual norm

$$\|\tilde{x}\| = \left\| x - \sum_{i=1}^k \langle x, e_i \rangle \mathbf{e}_i \right\|$$

exceeds the x standard deviation threshold. These time bins will be considered to contain anomalies.

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Experiment I: Anomaly Types

Broken link [HTTP status code 404]

A situation when requested files or requested directories from a web browser cannot be found

Directory Scan [HTTP status code 404, 403]

A situation when a brute-force search is performed with attempts to find valid directories and files in the web server

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Password Cracking [HTTP status code 401]

It is indicated by a series of failed logins within a short period of time by an attacker

Script Error [HTTP status code 500]

Unintentional syntax errors of PHP, Perl, or CGI scripts

Experiment I: Data

ApacheLog

- Obtained Apache log files from Traffy Server.
- We use 5 sets of log files from December 23 to December 27, 2009, with average of 2000 visitors per day.
- To simulate anomalous log, we inject four types of events into a dummy webserver.

The log files are then merged into the Traffy server.

Experiment I: Determining Useful Dimensions

Purpose

- To determine dimension of PCA results that would indicate web anomalies

Procedures

*Simulated anomalies in 2 rates
and injected each anomaly type
at 5 different intervals for each
background traffic

*Find 10 features

*Took these features in each

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background traffic to PCA

Types	Occurring rate 1	Occurring rate 2
Broken Link	60 times per minute	30 times per minute
	For 1 minute	For 1 minute
Password Cracking	60 times per minute	30 times per minute
	For 10 minute	For 10 minute
Scrip Error	60 times per minute	30 times per minute
	For 1 minute	For 1 minute
Directory Scan	805 scans in 35	805 scans in 35
	seconds	seconds

Experiment I: Determining Useful Dimensions

Conclusion

Click

<u>Dimensions</u>	Day 1		Day 2		Day 3		Day 4		Day 5	
	Rate	Rate	Rate	Rate	Rate	Rate	Rate	Rate	Rate	Rate
	1	2	1	2	1	2	1	2	1	2
Broken Link	5	5	4,5	-	5	5	4,5	5	5	-
Password Cracking	2,4	2,4	3,4	4	2,4	4	2,4	2,4	2,3	4
Scrip Error	5	6	5,6	6	5	6	4,5	6	5	6
Directory Scan	1,2	1,2	1,2,3	1,2,3	1,2	1,2	1,2	1,2	1,2	1,2

- Cannot

found

- The first six dimensions seem to be useful for detecting all web anomalies.
- As the attack rate decreases anomalies tend to show up in lower-variance

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Experiment II : Anomaly Detection

Purpose

To automatically detect anomalies from the PCA results

Procedures

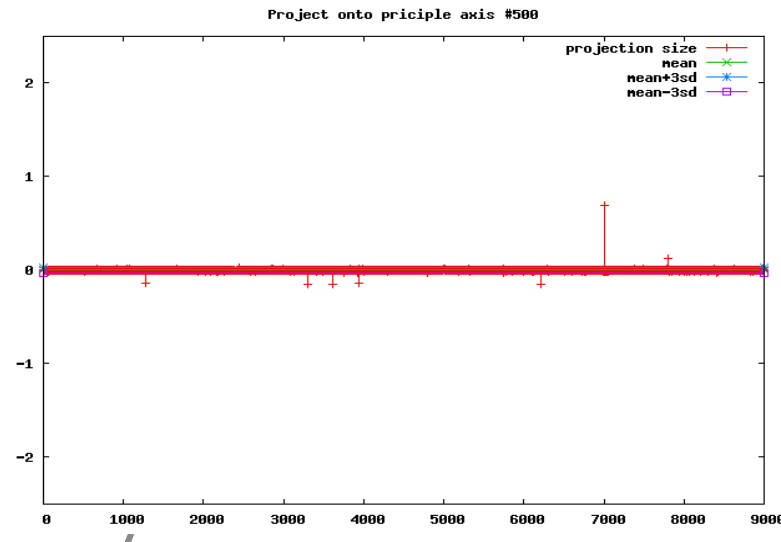
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*Simulate anomalies at rate2 and injected each anomaly

type at 5 different intervals for each log file.

*The web anomalies can be identified within 6 highest-variance dimensions of PCA results.

*Applied thresholds of 4,5 and 6 S.D. to obtain peaks in within those 6 axes



Types of anomalies

Occurring rate 2

Thresholds

Broken Link

30 times per minute
For 1 minute

4S.D.

Password Cracking

30 times per minute
For 10 minute

5S.D.

6S.D.

Scrip Error

30 times per minute
For 1 minute

Directory Scan

805 scans in 35 seconds

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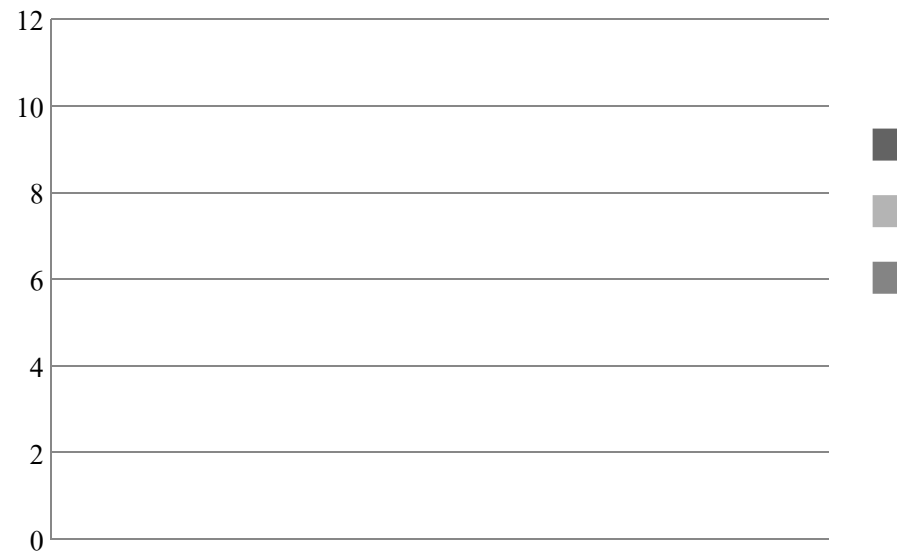
Experiment II : Anomaly Detection

Conclusion

Average Detection Hit Rate



Total number of false positives



27/9/10 The threshold can detect a directory scan event with the highest rate

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Future Works

- Classification algorithms (Done)
- Extend the set of features, therefore more problem types could be discovered.
- Develop the more robust algorithms for anomaly cutting points.
- Take periodic behaviors into consideration.
- Incremental updating algorithm.

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Any question ?



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