



Cybersecurity
Secure the Vision



Threat Hunting through Data Mining and Analytics

Scott Rodgers & Joel Amick



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Disclosure

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BUILT TO PERFORM.

CREATED TO SERVE.



Joel Amick
Director
Cyber Analytics



Scott Rodgers
Sr. Info Security Analyst
Cyber Analytics

The vision of Andrew Carnegie



\$1 trillion¹

assets under management



over
17,500
employees²



Serving
5M
individuals

15,000 
Institutions serviced by TIAA



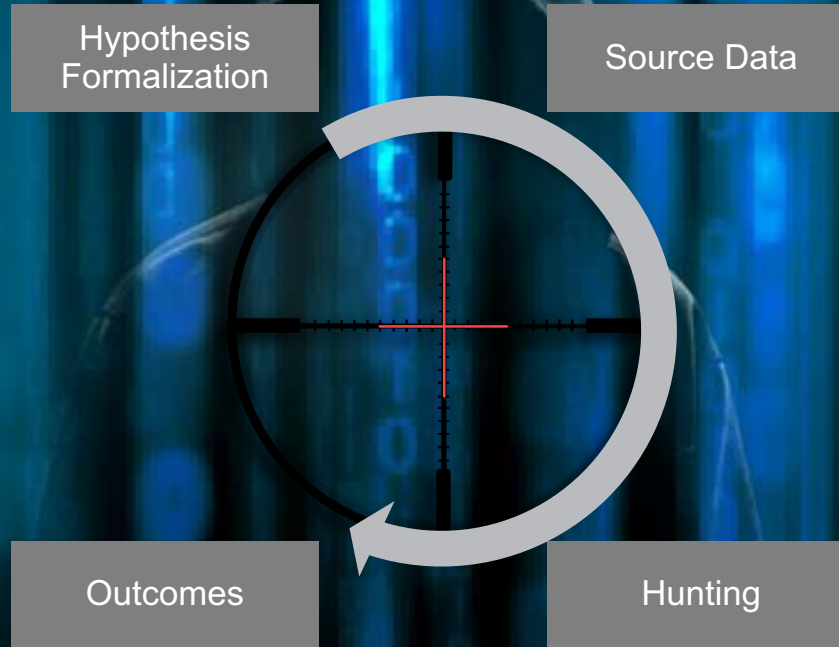
Threat Hunting is NOT:

- Using Static Tools/Black Boxes
- Reacting to a Dashboard or Console
- Monitoring Alerts
- A Single State



Threat Hunting IS:

- Proactive
- Hypothesis Based
- Timed
- Measureable
- Collaborative





Requirements

Requirements

- Access to required data
- Adequate understanding of data and data correlation methods



Hypothesis Criteria

- Specific
- Provable/disprovable
- Does not necessarily need to be *if/then* scenario



Ethical Concerns

- Partner with Human Resources and Legal

Example of Poor Hypothesis

“We have web account registration fraud on our website.”

Example of Good Hypothesis

“There is a statistically significant amount of web account registration fraud that can be identified by comparing IP geolocation and time zone.”

Data Scientist/Analyst



Statistical Background
Data Mining
Data Access

Cybersecurity Analyst/Investigator



Subject Matter Expert
Develop Hypotheses
Validating Results



External Threat

Freddy Fraud gathers stolen identity data to try to register fraudulent accounts on various websites.

Freddy has recently discovered your organization and decides to try a cache of stolen identity on your organization's website.

Insider Threat

Ivy Insider works as a developer at your organization. Recently, she was granted permissions that enables her to log on to any workstation and view the hard drive contents.

Ivy decides to seize this opportunity to log on to a number of executive machines to look for trade secret information to sell.





Target

Web Account Registration Fraud



Hypothesis

Mismatched attributes observed during web account registration have a statistically significant rate of fraud.



Source Data

Web Account Registration Data

Web Traffic Data

Participant Data



Hunt Steps

Identify Correlating Attributes

Join and Enrich Datasets

Create Additional Fields / Features

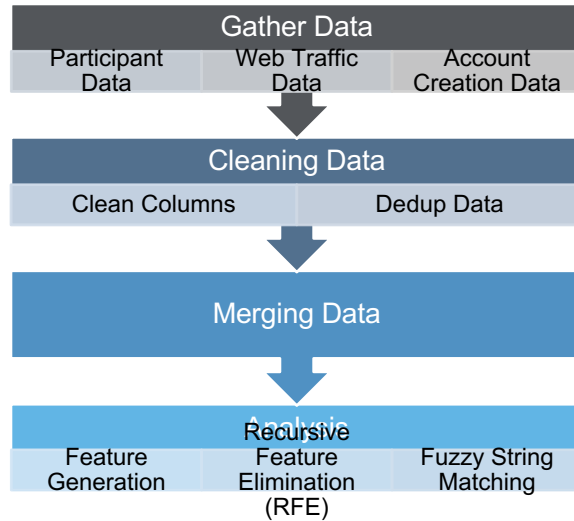


Outcomes

Actionable Intelligence

Rule Creation

Learning



```
## Read Participant CSV  
part_data = pd.read_csv(file_path)
```

```
## Get Web Account Registration Data  
conn = pyodbc.connect(connection_string)  
sql = "Select * from db.account_creation"  
acc_data = pd.read_sql(sql, conn)
```

```
## Drop Duplicates  
acc_data = acc_data.drop_duplicates(subset="user_id", keep="first")
```

```
## Merge  
merged_data = pd.merge(part_data, acc_data, left_index=True, right_index=True, sort=False)
```

```
## Calculate Fuzzy String Match Score  
matched_data = check_matches(merged_data, merged_col_names)
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## Save to CSV  
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```
## Calculate Fuzzy String Match Score
def check_matches(data, col_names):
    for combo in combinations(cols,2):
        compare_loop(Data, combo[0], combo[1])

def compare_loop(data, col1, col2):
    for index, row in Data.iterrows():
        fuzz.token_sort_ratio(row[col1], row[col2])
```



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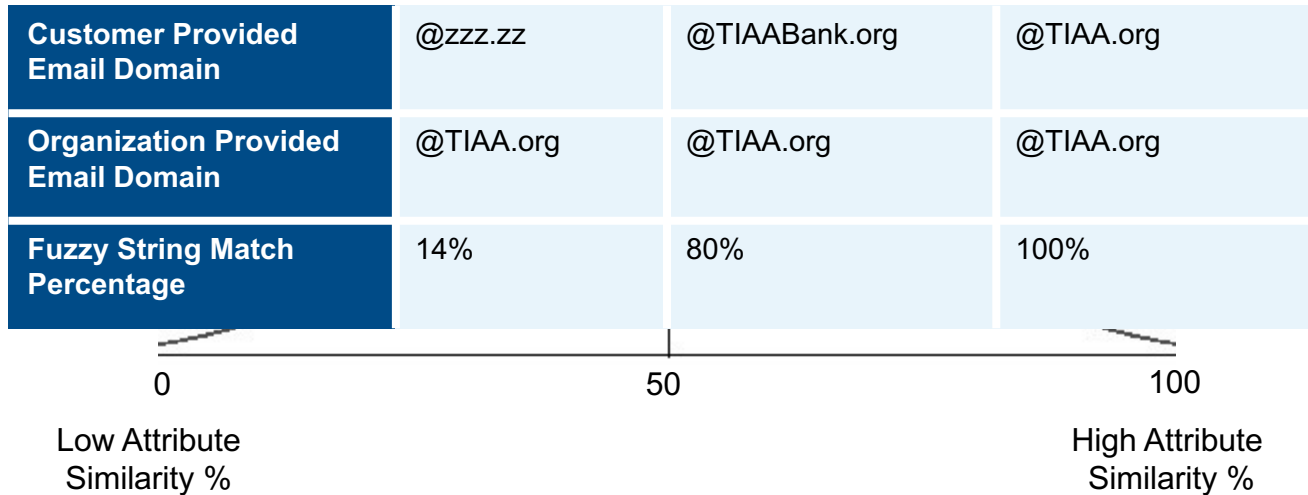
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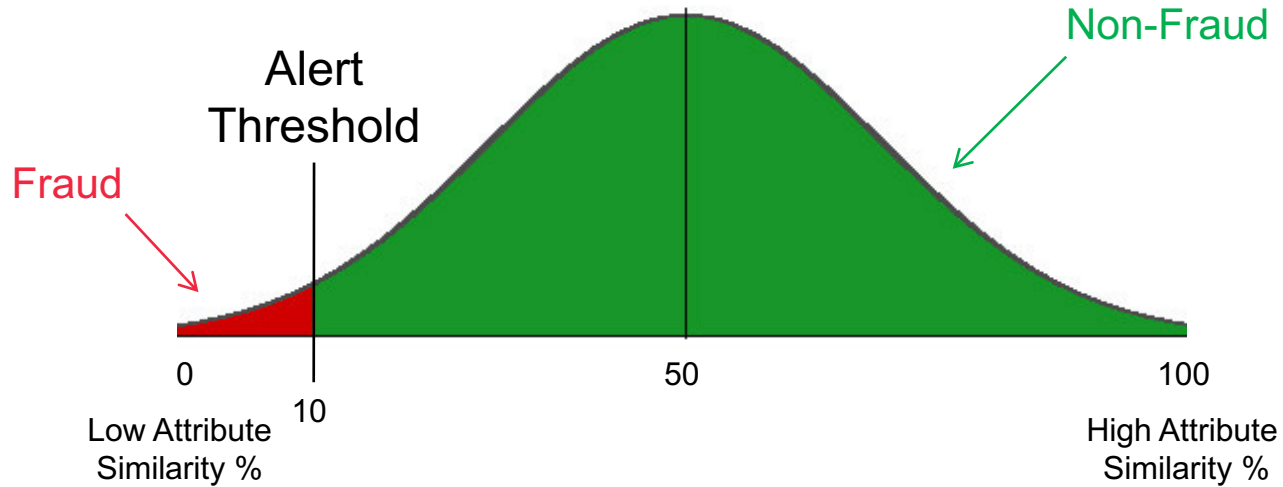
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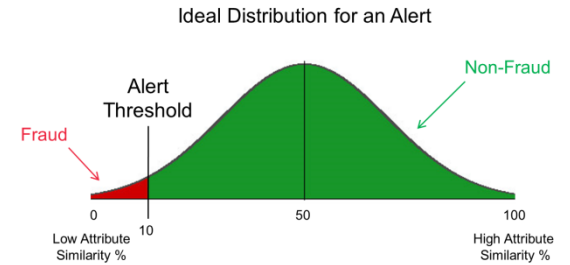
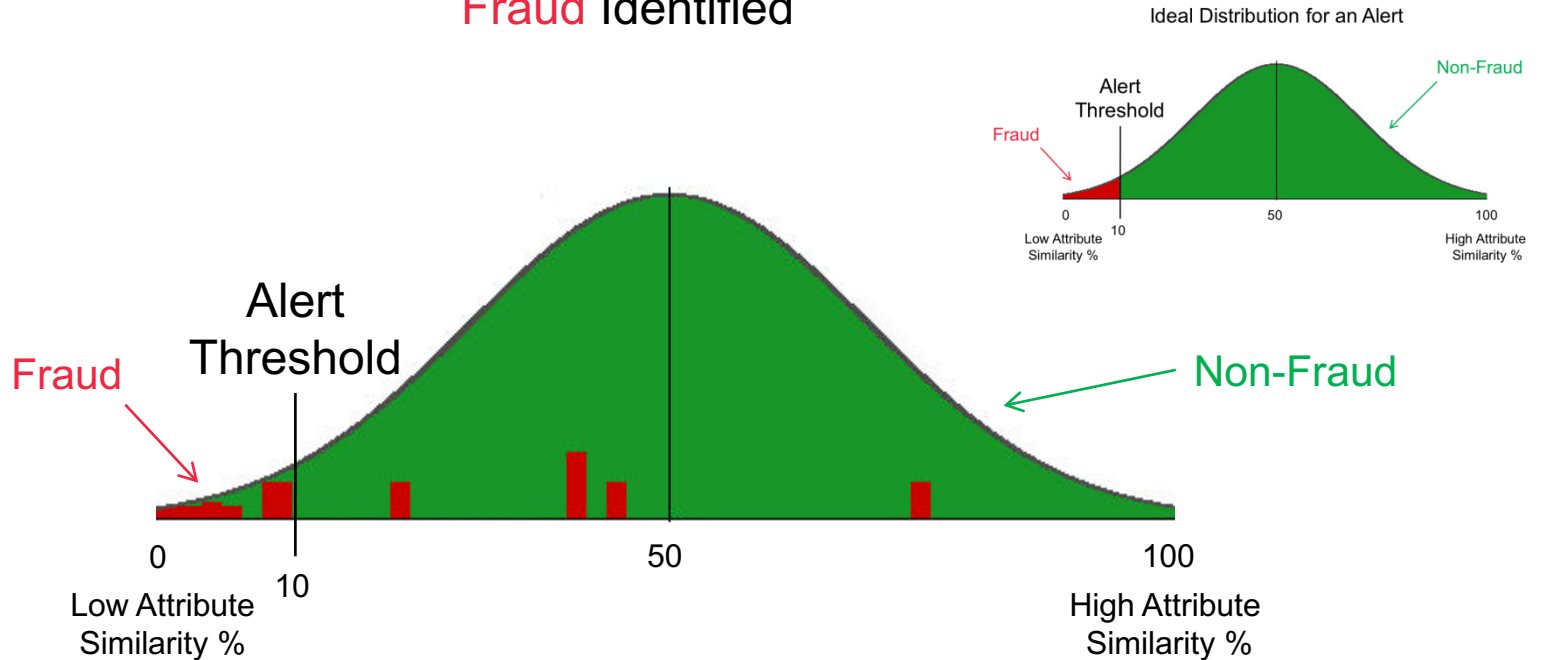
What Are We Looking For?

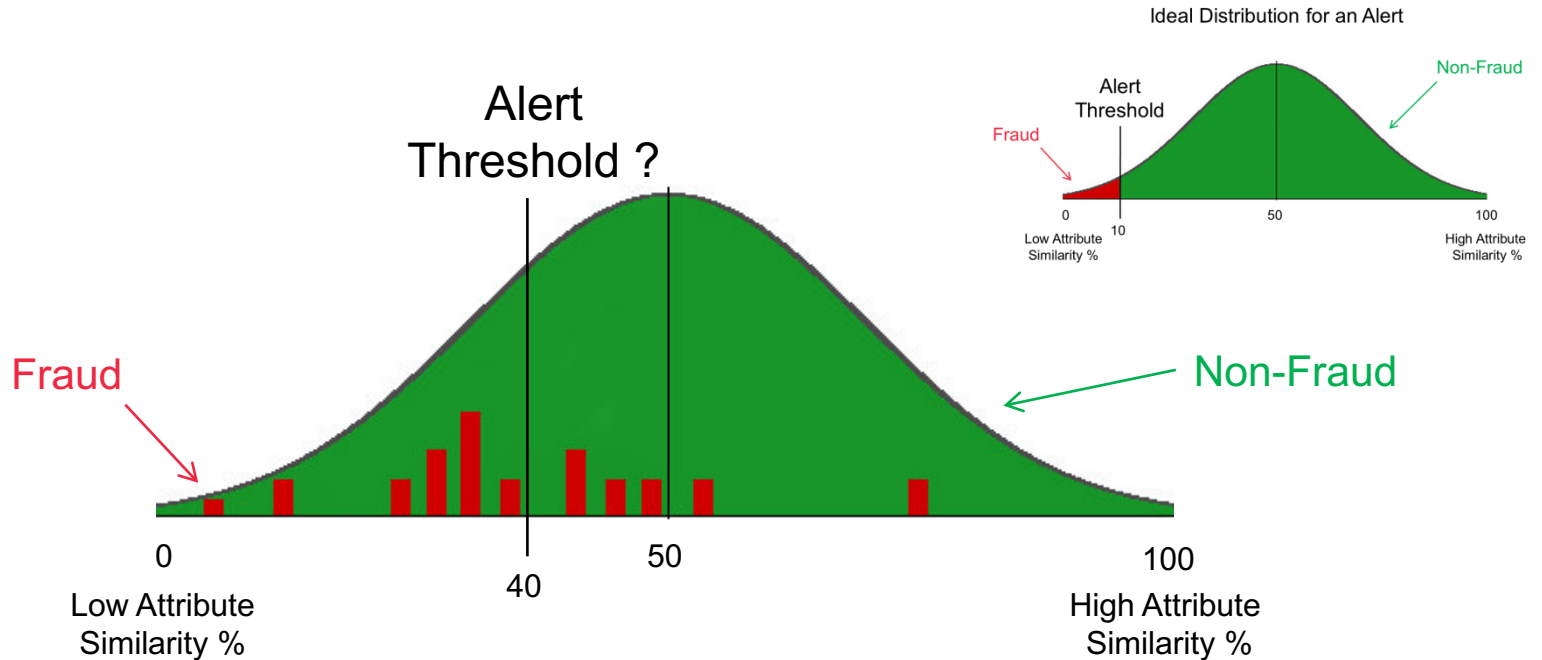


Ideal Distribution for an Alert



Fraud Identified





Hypothesis – Proven

- Mismatched attributes (Email Address Domain – Customer Organization Domain) observed during web account registration have a statistically significant rate of fraud.

Actionable Intelligence

- Additional fraudulent web account registration events identified by mismatched attributes.

Policy Issues Identified

- Why are users allowed to register a new web account with an email domain that does not match their organization?

Learning

- Hunters gained hands-on experience using and correlating web account registration data.



Target

**Executive
Computers**



Hypothesis

Users with elevated permissions are accessing executive workstations without business justification.



Source Data

**Windows
Event Logs**

**Employee
Titles**

**Workstation
Data**

**Support
Ticketing
Data**



Hunt Steps

**Identify
Correlating
Attributes**

**Join and
Enrich
Datasets**

**Investigate
Results**



Outcomes

**Report
Actionable
Intelligence**

**Identified
Policy Issue**

**Developed
Mitigating
Alerting
Strategy**

```
1 -- Select columns
2 select w.WorkstationName, u.Userid, u.Name, u.JobTitle from db.Users u
3 -- Join Workstation Data
4 join db.workstation w
5 on u.UserId=d. UserId
6 -- Filter based on Job Title
7 where u.JobTitle Like '%exec%' OR u.JobTitle Like '%vp%'
```

Workstation Name	User ID	Name	Job Title
exec-pc-001	exec001	Eric Exec	Executive
exec-pc-002	exec002	Eddie Exec	Executive
exec-pc-003	exec003	Erin Exec	Executive
vp-pc-005	vp005	Val VP	Vice President

Windows Event Log 4624 – Successful logon

An account was successfully logged on.

Subject:

Security ID: SYSTEM
Account Name: EXEC-PC-001\$
Account Domain: WORKGROUP
Logon ID: 0x3E7

Process Information:

Process ID: 0x30c
Process Name: C:\Windows\System32\lsass.exe

Logon Information:

Logon Type: 3
Restricted Admin Mode: -
Virtual Account: No
Elevated Token: No

Network Information:

Workstation Name: EXEC-PC-001
Source Network Address: -
Source Port: -

Impersonation Level: Impersonation

Detailed Authentication Information:

Logon Process: Negotiate
Authentication Package: Negotiate
Transited Services: -
Package Name (NTLM only): -
Key Length: 0

New Logon:

Security ID: AzureAD\exec001
Account Name: exec001@org.com
Account Domain: AzureAD
Logon ID: 0xFD5113F
Linked Logon ID: 0xFD5112A
Network Account Name: -
Network Account Domain: -
Logon GUID: {00000000-0000-0000-0000-000000000000}

Fields used in hunt

Look for successful windows login events

```
eventtype=microsoft-windows-events AND eventcode=4624
```

Filter logins to executive workstations

```
AND (workstation_name=exec-pc-* OR workstation_name=vp-pc-005)
```

Aggregate login events by workstation

```
| stats count(EventCode) as "Login Count", dc(Security_ID) as "User Count", values(Security_ID) as "Users", values(Logon_Type) as "Logon Type" by workstation_name
```

Only show workstations with more than 1 distinct user

```
| where "User Count" > 1
```

Computer Name	User Count	Login Count	Users	Title	Logon Type	Logon Type Description
exec-pc-001	2	1	admin_ivy_insider	Developer	3	Network
		9	exec001	Executive	2	Interactive
exec-pc-002	2	2	admin_ivy_insider	Developer	3	Network
		38	Exec002	Executive	2	Interactive
exec-pc-003	3	1	admin_service_desk_ron	Sr. Desktop Support	10	Remote Interactive
		2	service_desk_elliott	Desktop Support	10	Remote Interactive
		15	exec003	Executive	2	Interactive
vp-pc-005	2	2	service_desk_elliott	Desktop Support	10	Remote Interactive
		23	vp005	Vice President	2	Interactive

Actionable Intelligence

Hypothesis – Proven

- Users with elevated permissions are accessing executive workstations without business justification.

Actionable Intelligence

- User Ivy Insider abusing privileges and logging into executive systems.

Policy Issues Identified

- Is least privilege being applied to user permissions?

Learning

- Correlated disparate data & built understanding of data



External Threat

- Web Account Creation
 - Web Channel
 - Phone Channel
- Account Takeover
- Cross-Channel Fraud
- One Time Pin (OTP) Abuse



Insider Threat

- Unapproved or Portable Applications
- Personal VPN Clients to avoid Data Loss Prevention Tools
- File Sharing
- Remote Desktop / Access Tools



Q&A

