



# Hunting SMB Shares

With Data, Graphs, Charts, and LLMS

Scott Sutherland



# Scott Sutherland

GitHub: nullbind (nullbind)  
X: @\_nullbind  
Bsky: @nullbind.bsky.social

## VP of Research at NetSPI

Service & Product Development

Find, exploit, and detect things that go boom on your network

## GitHub Projects

[github.com/netspi](https://github.com/netspi)/PowerHuntShares  
/PowerUpSQL  
/DAFT  
/SQLC2  
/PowerHunt  
/PowerShell/Crypt-It

## Blogs

<https://www.netspi.com/authors/scott-sutherland/>



# Two Parts One Story

1. A legacy of excessive privileges.
2. Hunting for context in a sea of share data.



# Story Time

A legacy of excessive privileges.

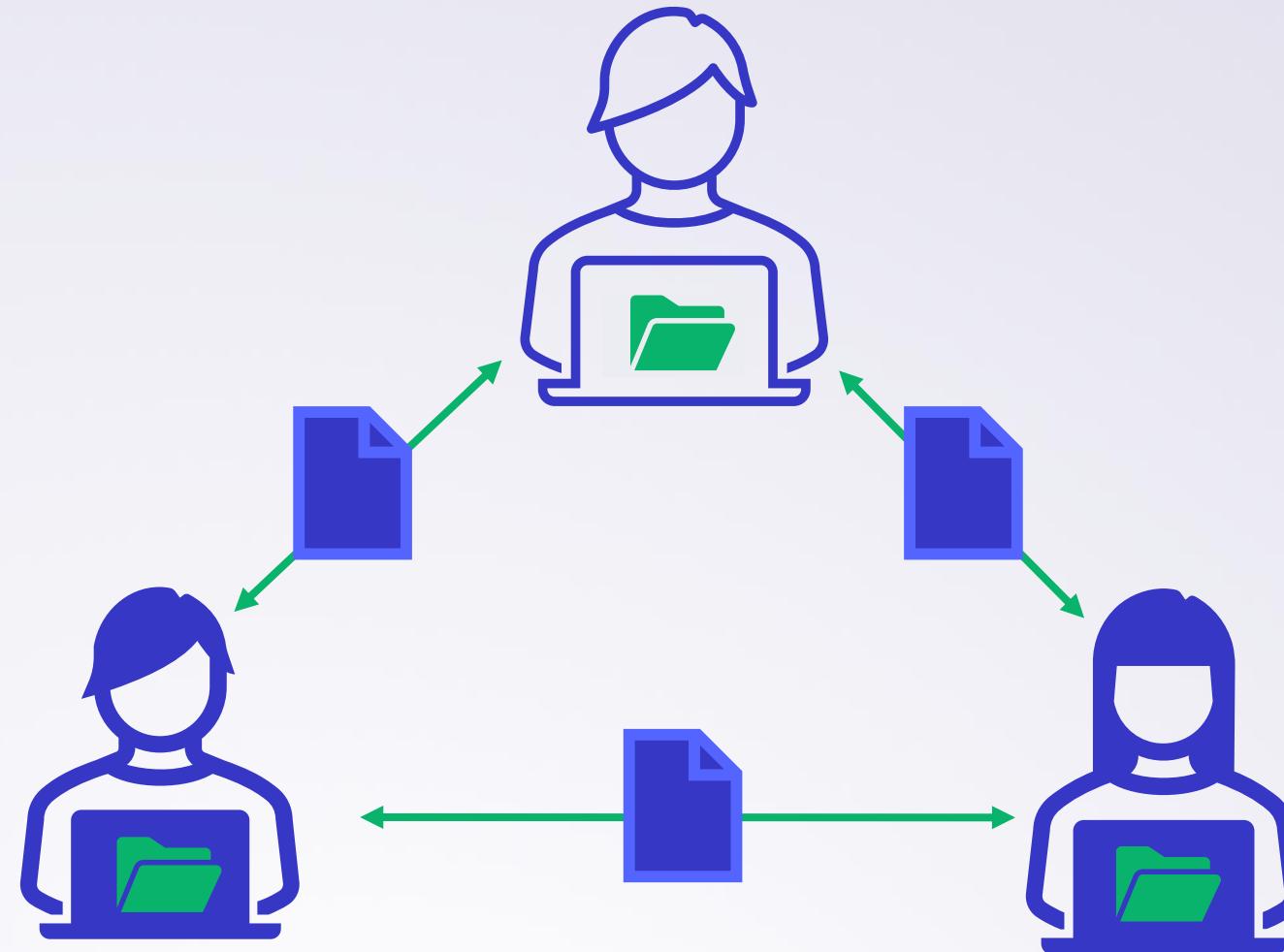




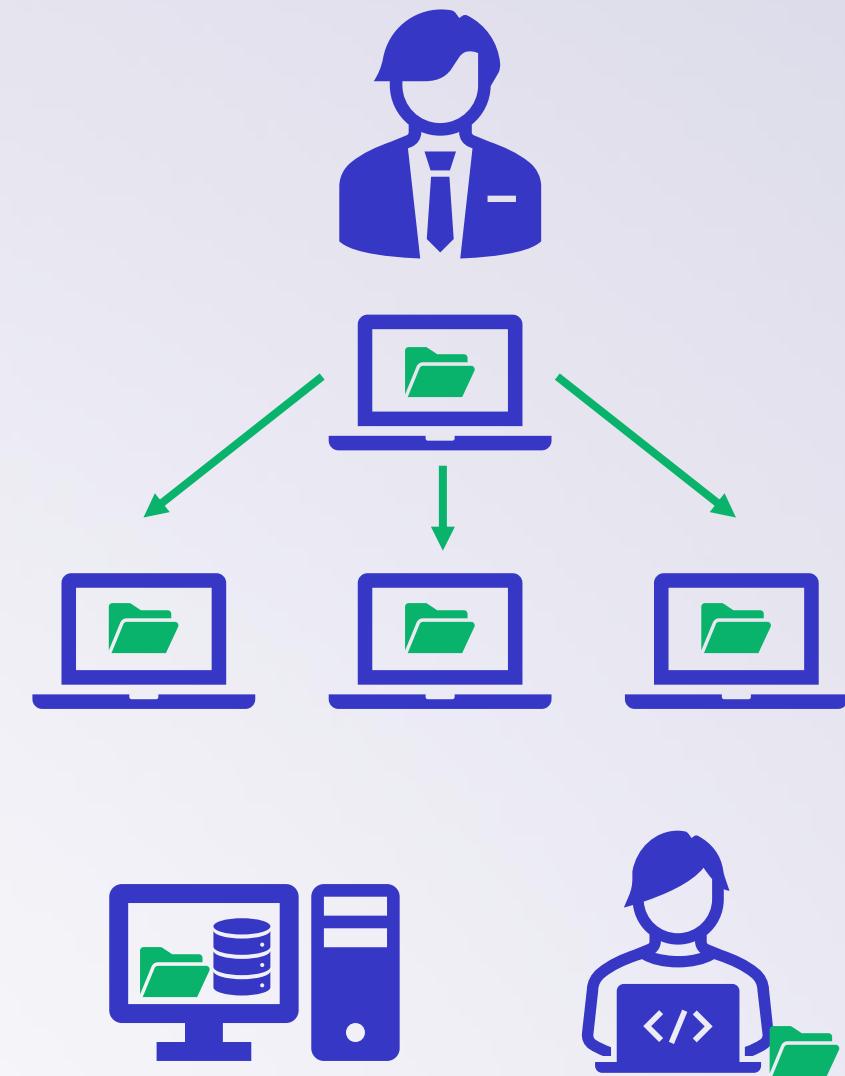
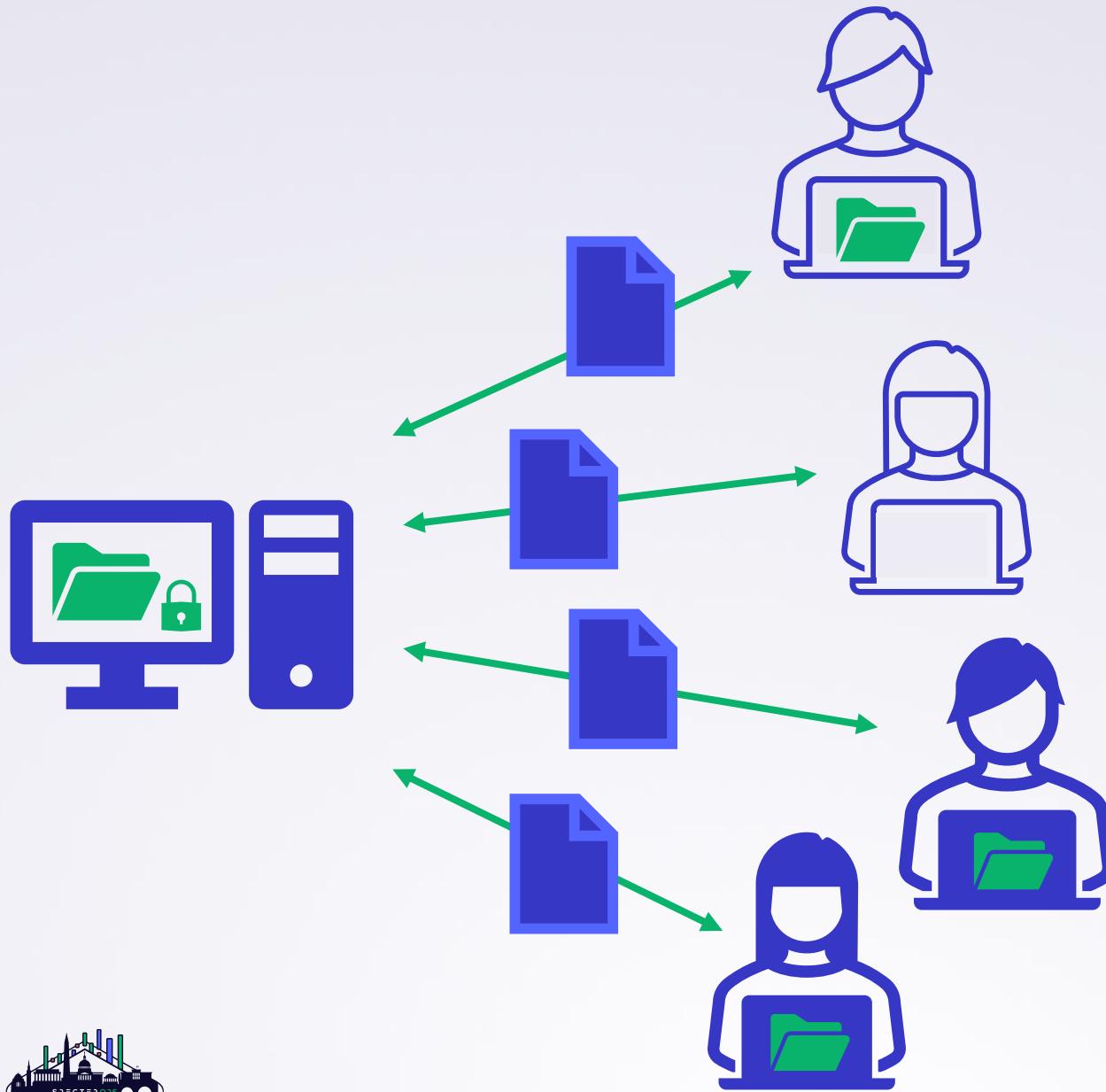
Story Time



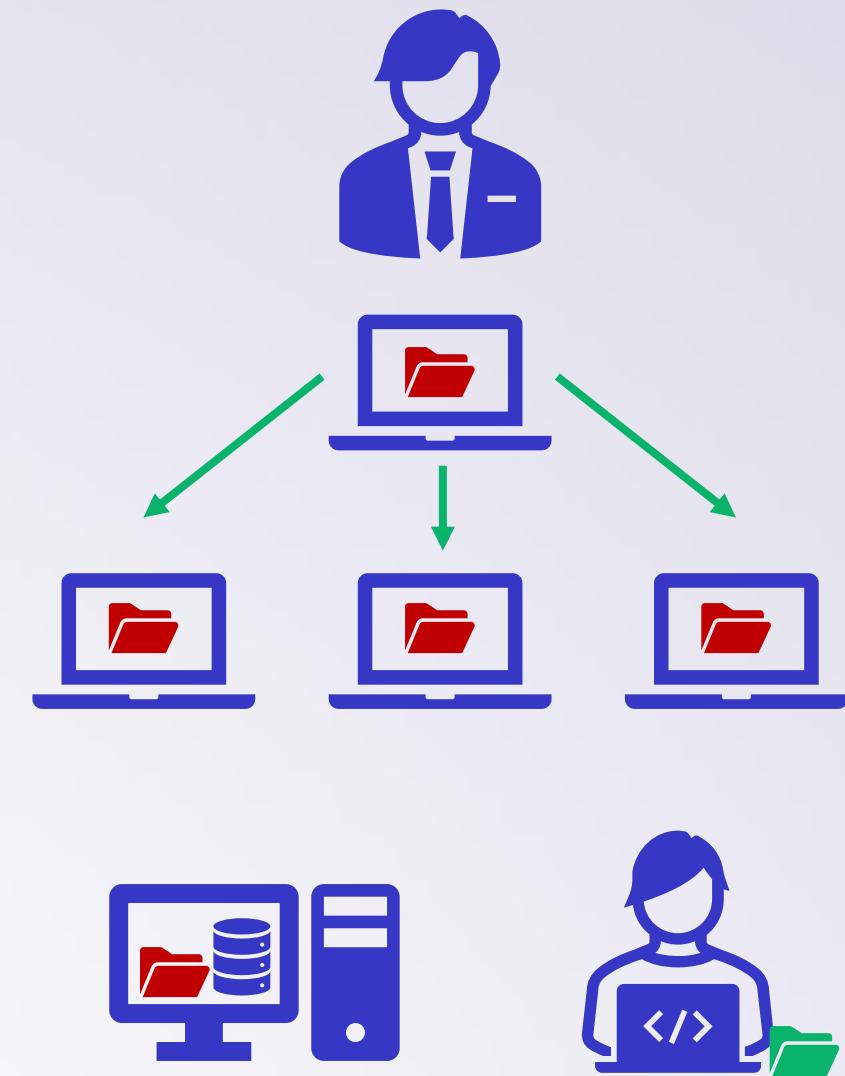
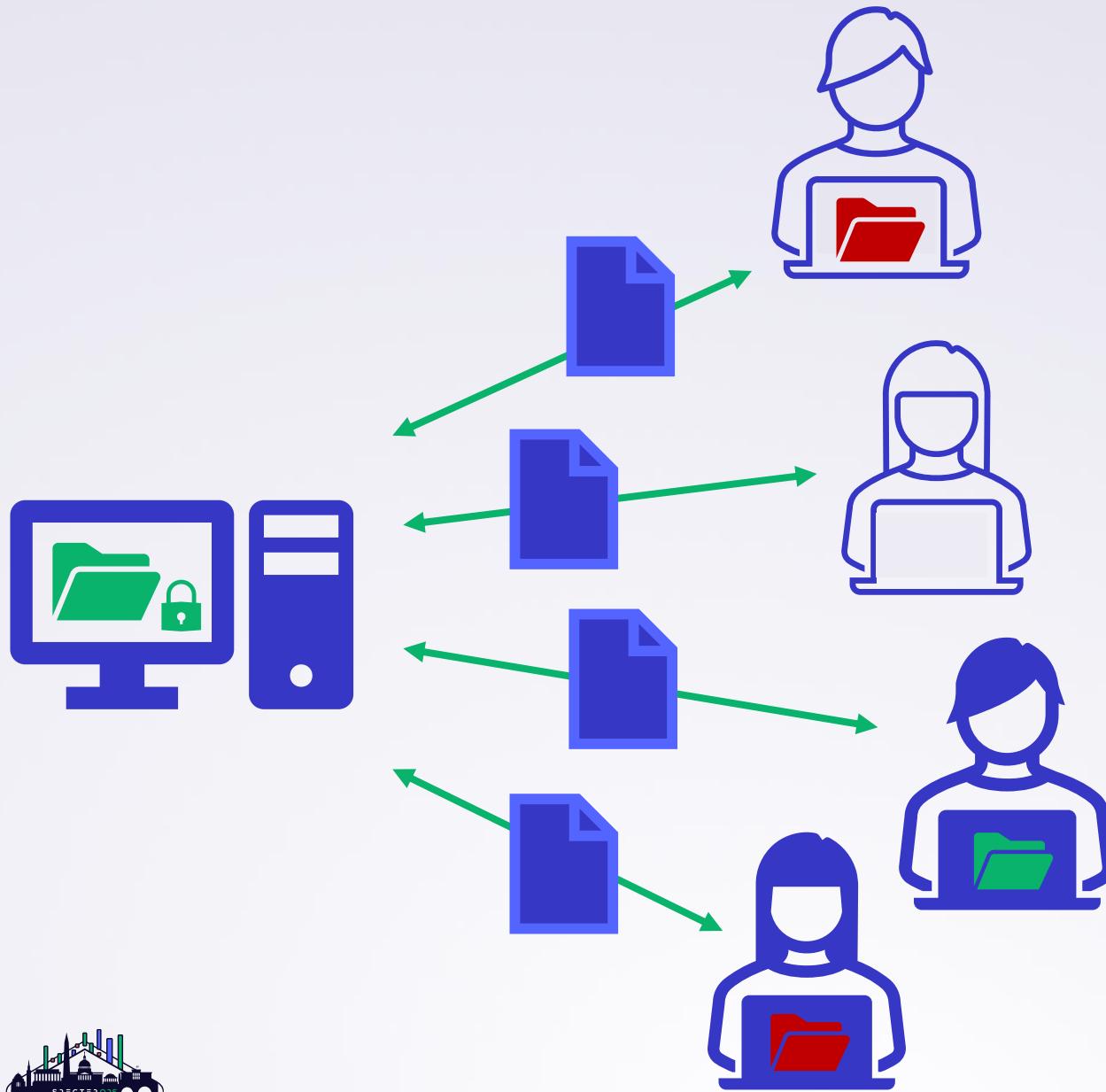
Story Time



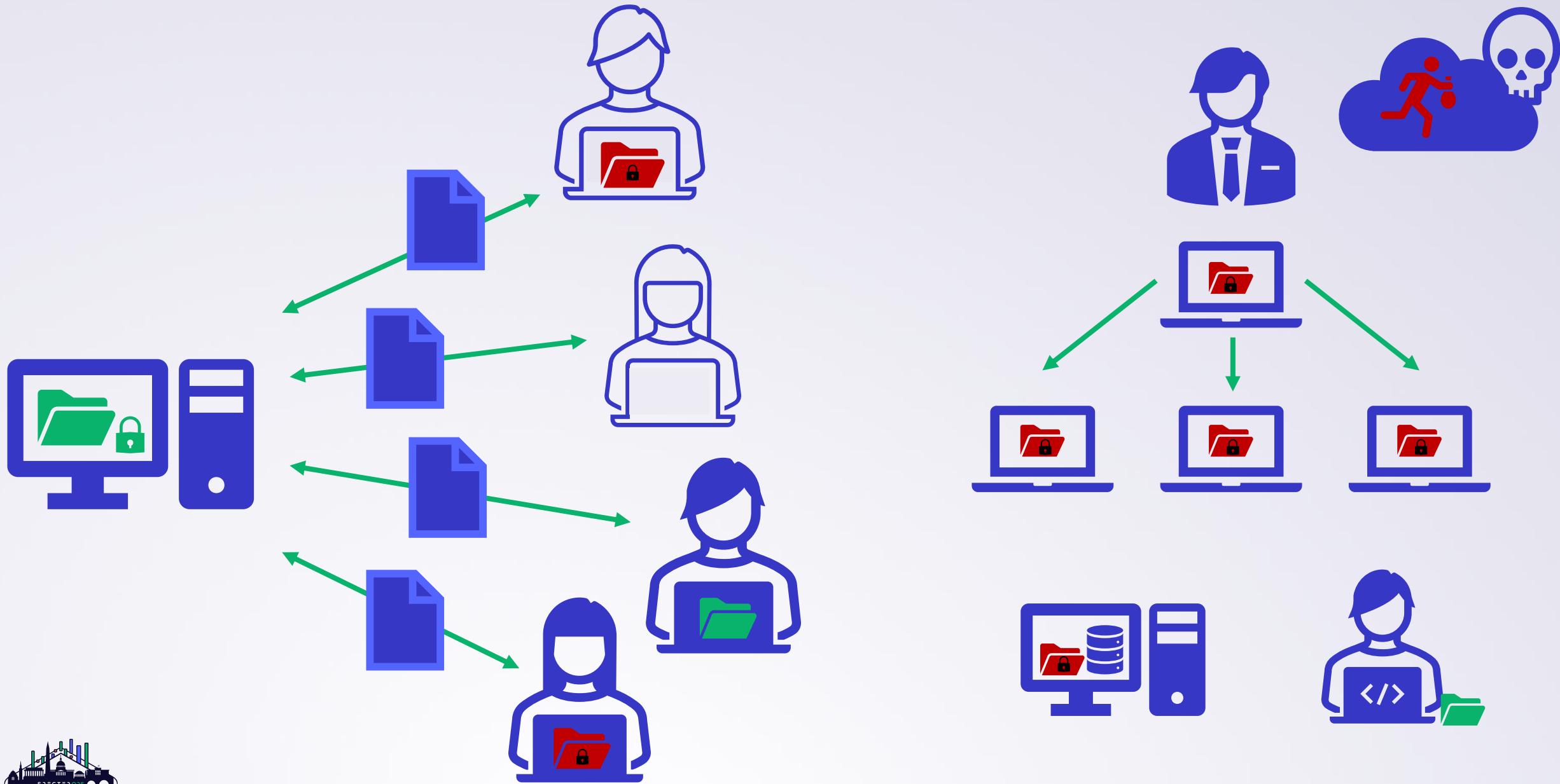
Story Time



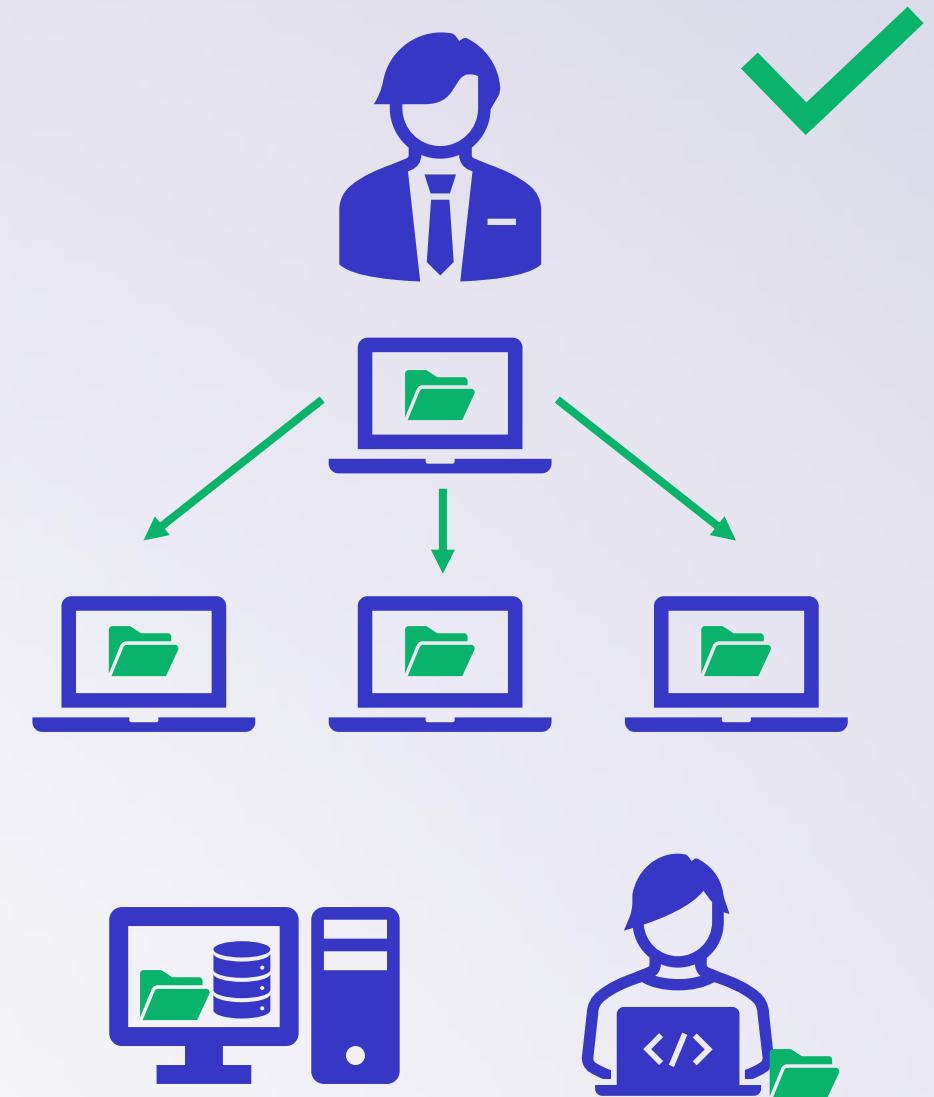
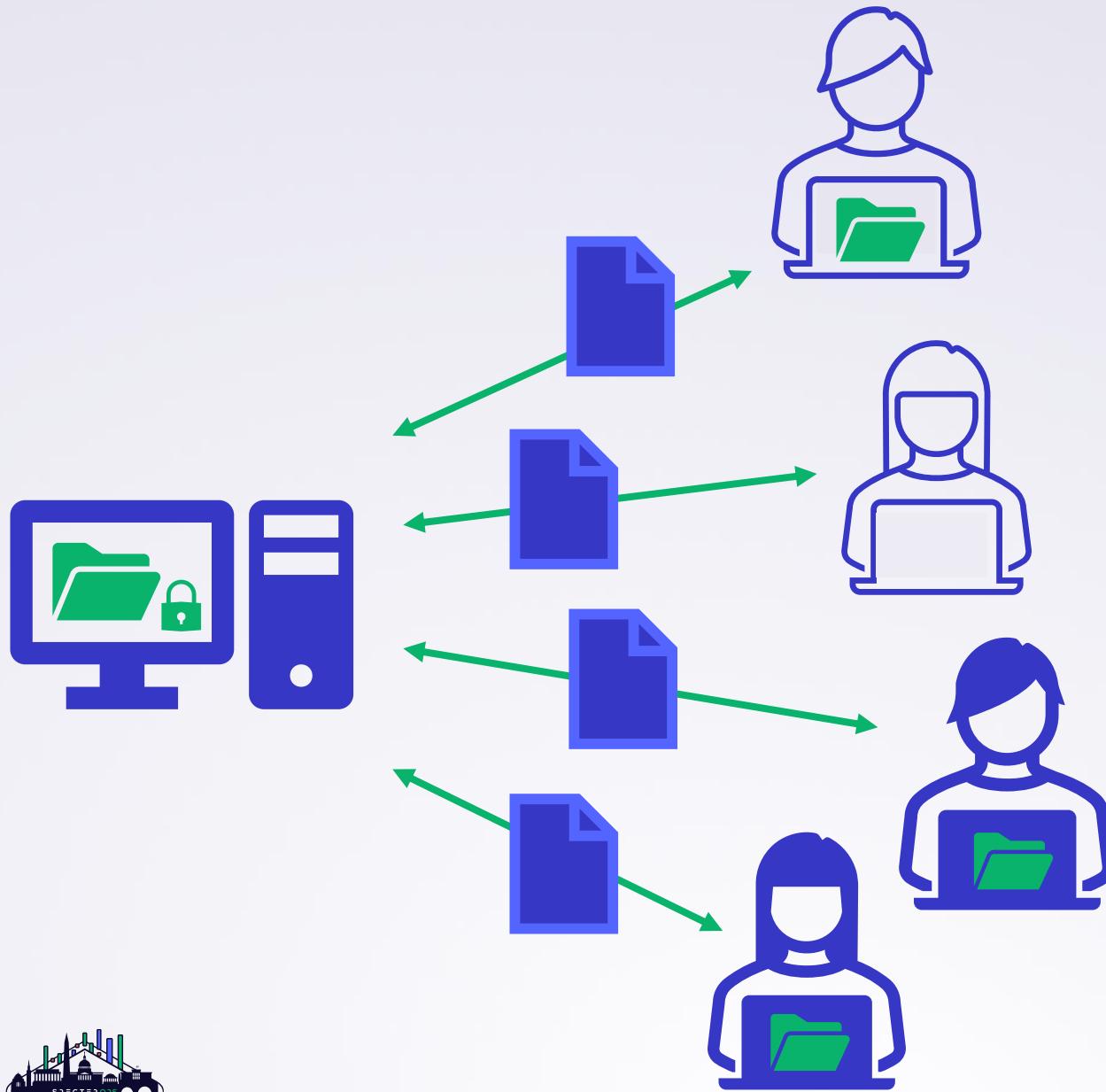
Story Time



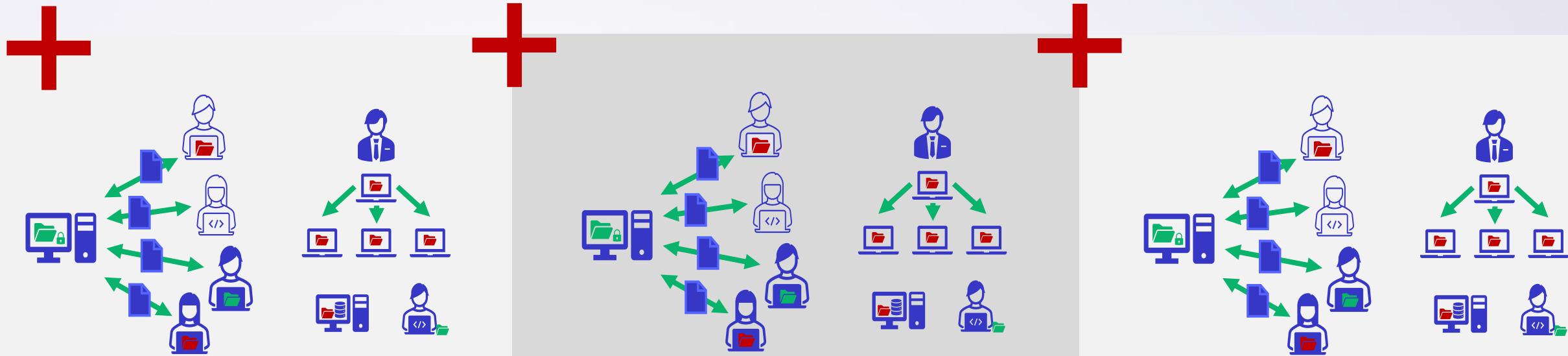
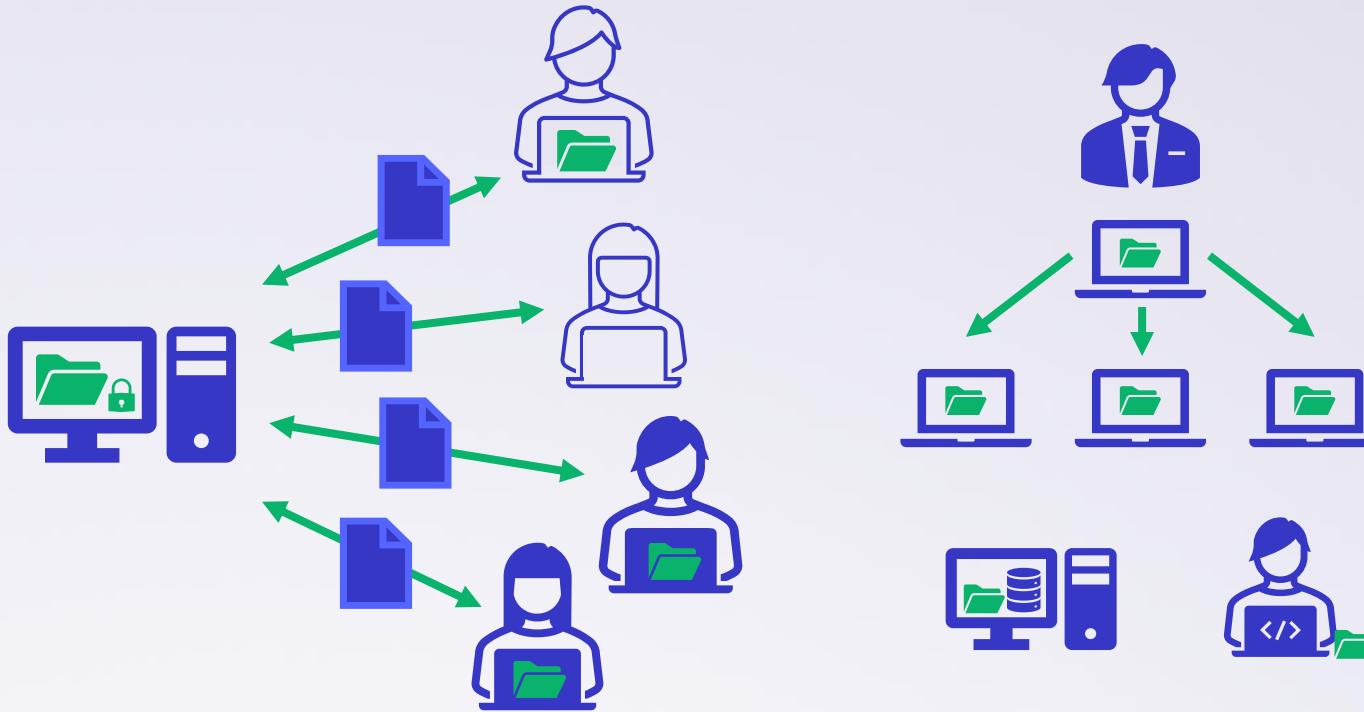
Story Time



Story Time



Story Time



**This is a reality that a lot of businesses  
are trying to manage. Still.**



# **What's broken, why are we missing so much now?**



## What's broken, why are we missing so much now?

- Incomplete inventory
- Insufficient vulnerability scanning
- Privilege inheritance and nested groups
- Generally understanding share context
- **Managing permissions at scale is hard!**

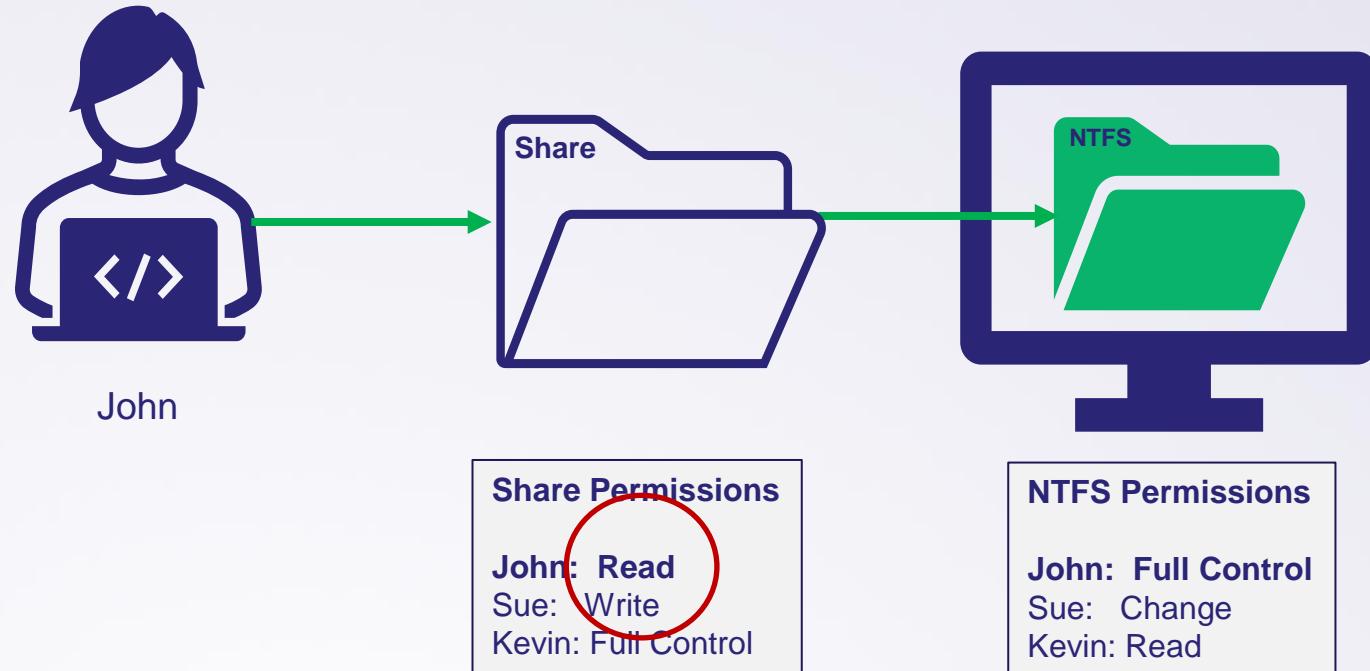


# So How do SMB Share Permissions Work?



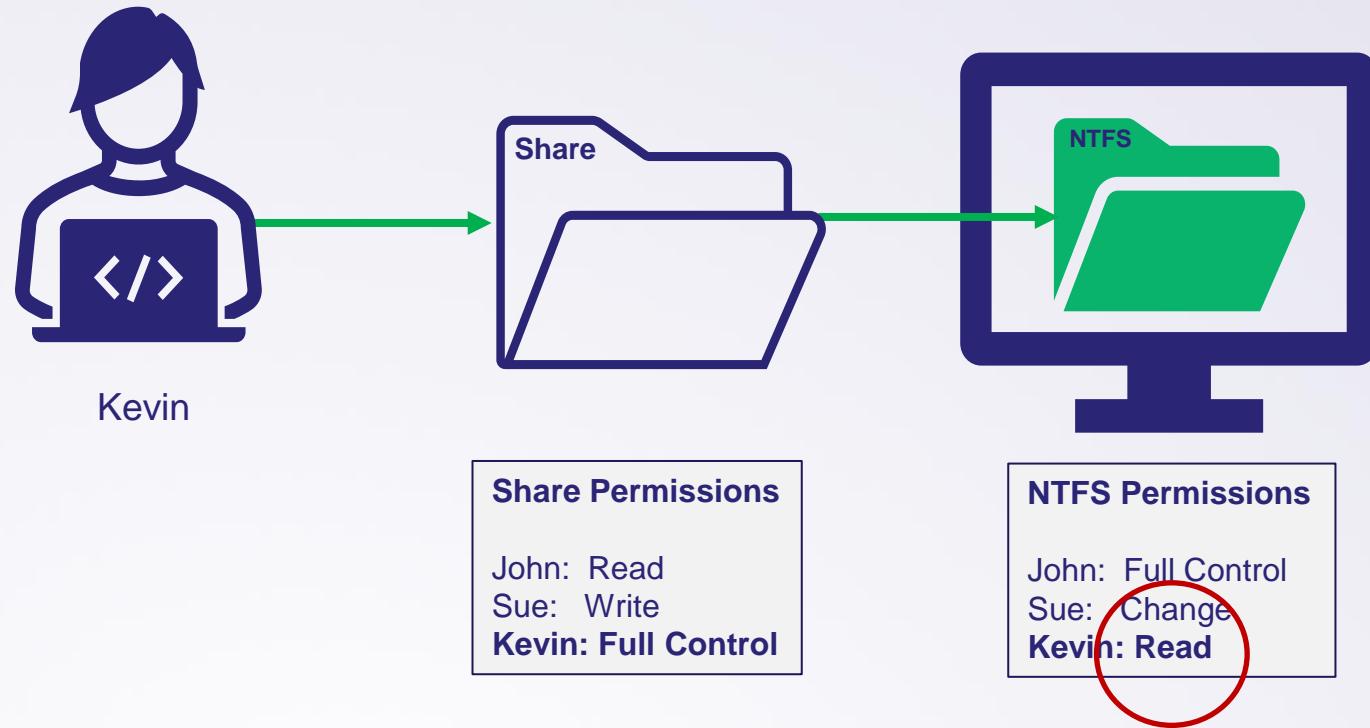
# NTFS & Share Permission

## Most Restrictive Wins



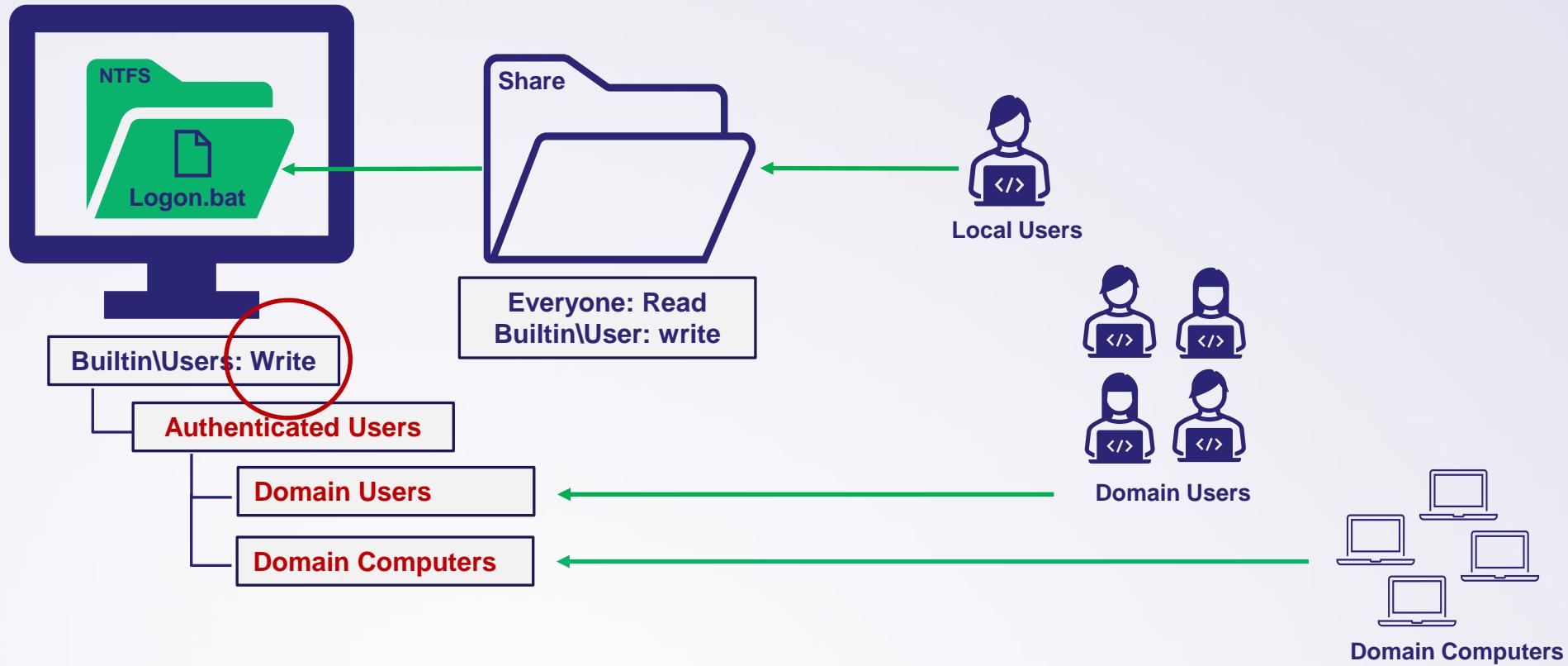
# NTFS & Share Permission

## Most Restrictive Wins



# Default Inherited Permissions

## Are. The. Worst. ...Best?

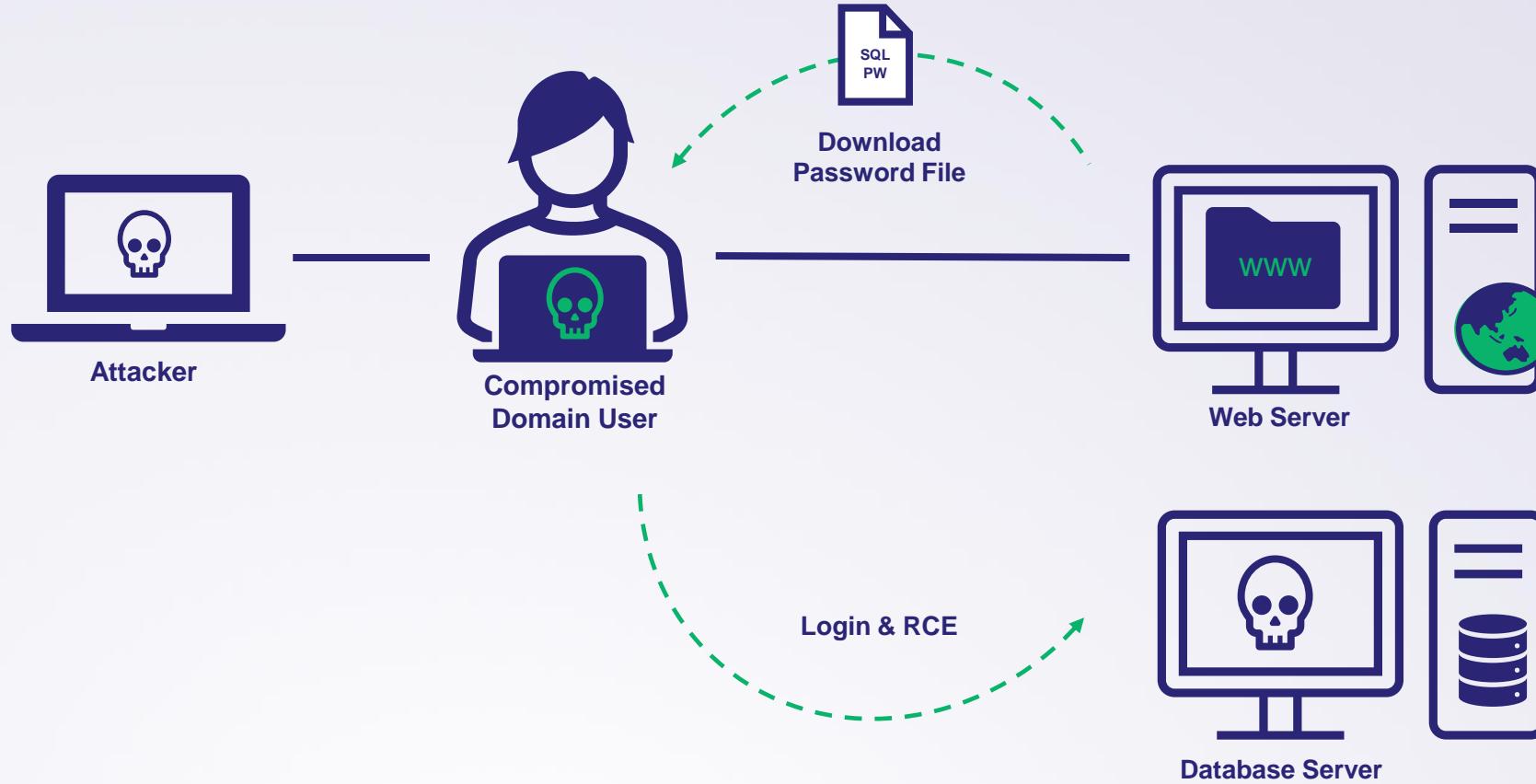


## What's the impact, what can attackers do?

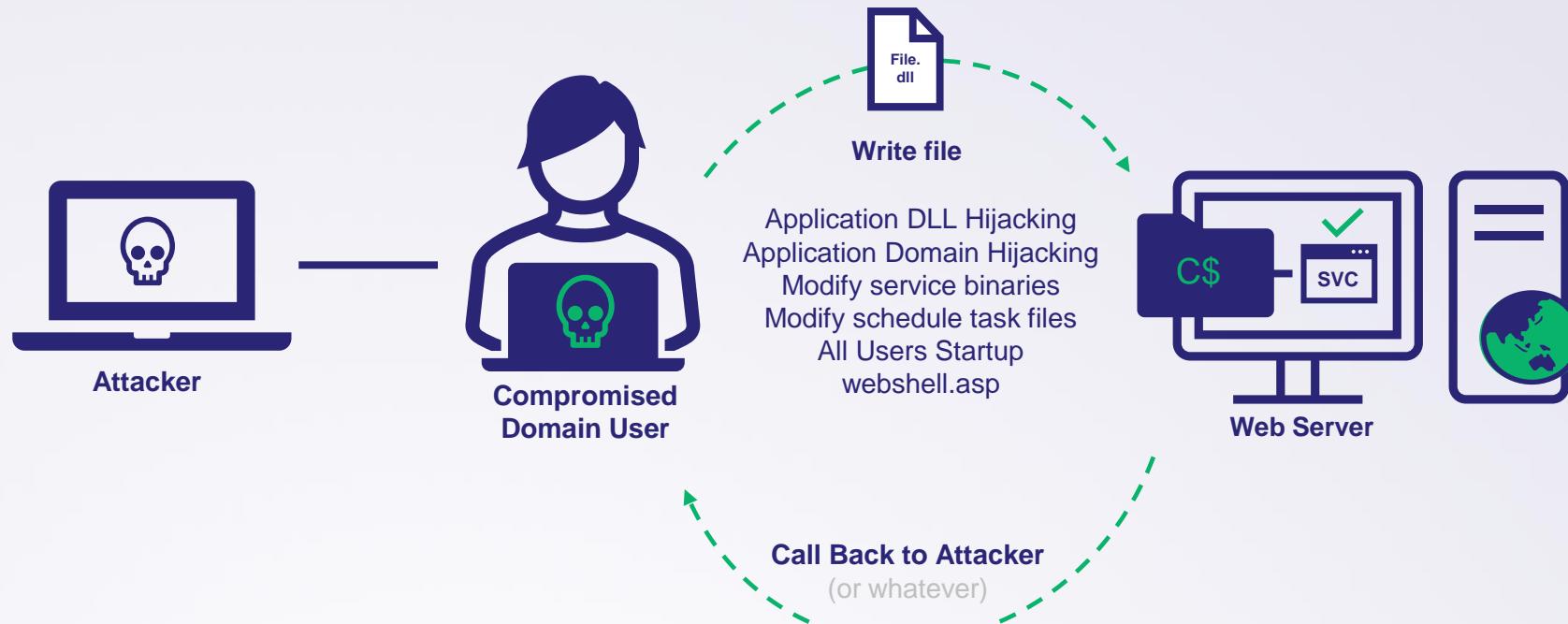
- **Read** data they shouldn't be able to
- **Write, Modify, Delete** data they shouldn't be able to
- **Execute Code Remotely...**



# Attacking Shares Read Access.



# Attacking Shares Write Access.



**Shares are one of the  
MOST abused attack surfaces but require the  
LEAST amount of knowledge to attack**

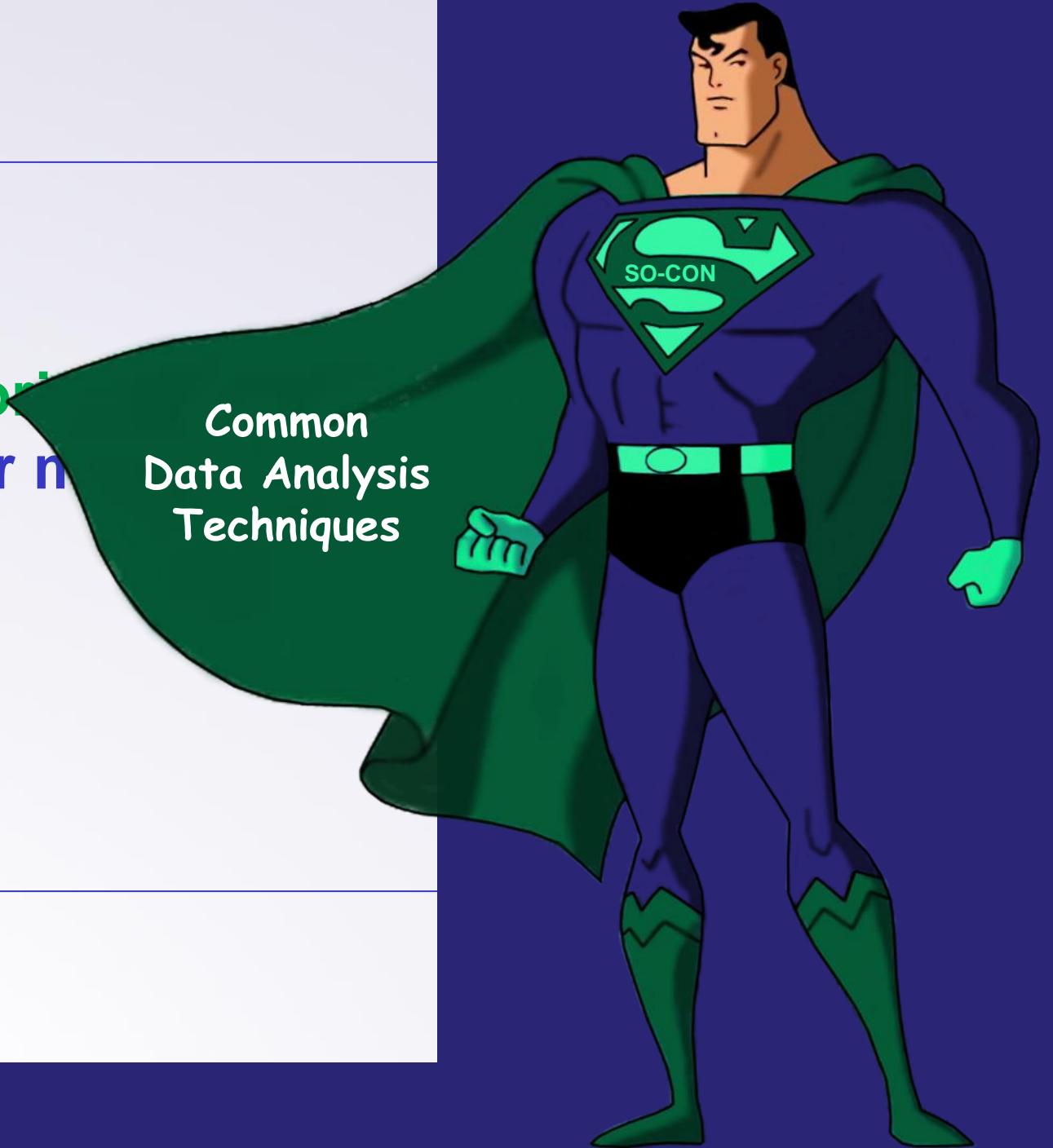


**How do we determine  
which share exposures represent actual risk?**



**How do we prioritize  
100,000 or more?**

Common  
Data Analysis  
Techniques



# Hunting for context in a sea of share data

*...while building PowerHuntShares v2*



# What is PowerHuntShares?

<https://github.com/NetSPI/PowerHuntShares>

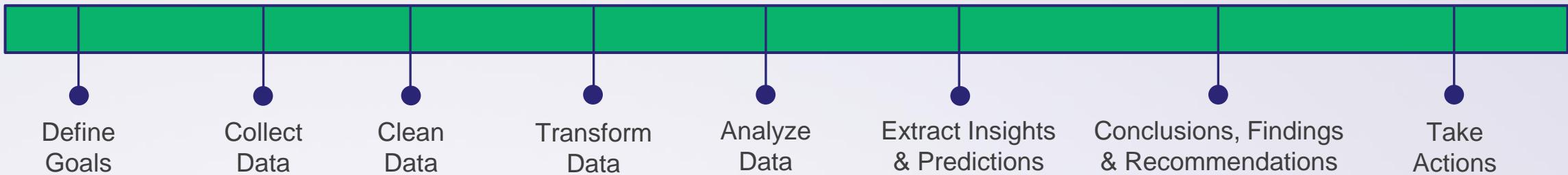
*“PowerHuntShares is PowerShell tool designed to help cybersecurity teams and penetration testers better identify, understand, attack, and remediate SMB shares in the Active Directory environments they protect.”*

## Key Features

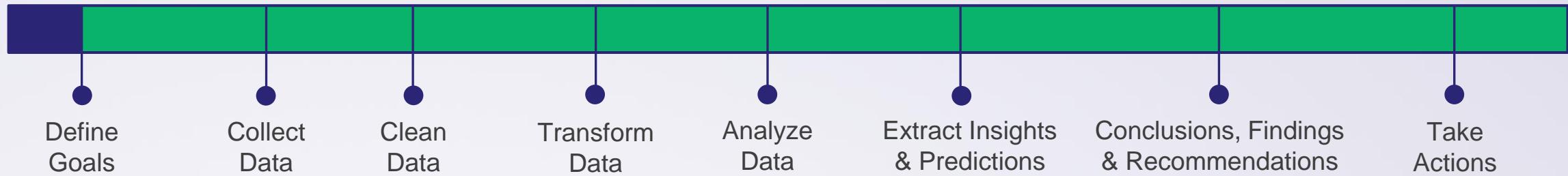
- Find Shares with Excessive Privileges
- Find RCE
- Find Data Exposures
- Find & Extract Secrets
- Add context through data enrichment
- Gain insights to prioritize and drive action!



# PowerHuntShares Process

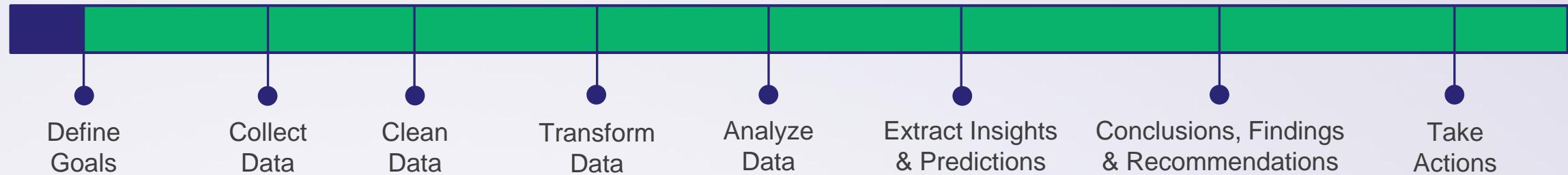


# PowerHuntShares Process



*“Alice created the ‘MyApp\$’ share on 200 systems to support the SuperPOS3k application on 4/1/2025. The shares were configured excessive read/write privileges which exposed sensitive data and provided a means to execute remote code.”*

# PowerHuntShares Process

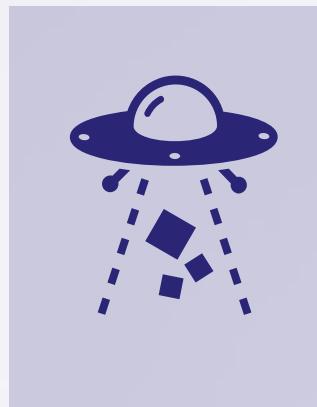
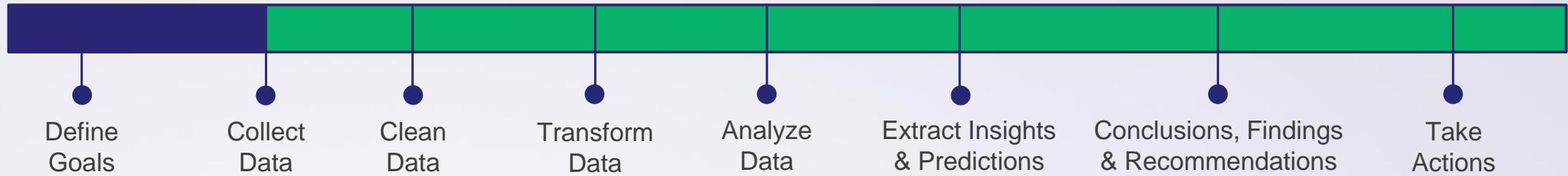


## Goals: Who, What, When, Where, Why, How

- **What Happened?**  
Descriptive Analysis
- **Why did it happen?**  
Diagnostic Analysis
- **What will happen?**  
Predictive Analysis
- **What should I do?**  
Prescriptive Analysis



# PowerHuntShares Process



## Data Collection

- **Asset Coverage**  
Active directory query + port connectivity tests + optional ping test
- **Data Visibility**  
Names, dates creation, last modified, and last accessed dates  
Directory listings, hashes of directory listings, file counts



# Bypass. Download. Run.

<https://github.com/NetSPI/powerhuntshares>

```
# Bypass execution policy restrictions
Set-ExecutionPolicy -Scope Process Bypass

# Import module that exists in the current directory
Import-Module .\PowerHuntShares.psm1

or

# Reduce SSL operating level to support connection to github
[System.Net.ServicePointManager]::ServerCertificateValidationCallback = {$true}
[Net.ServicePointManager]::SecurityProtocol =[Net.SecurityProtocolType]::Tls12

# Download and load PowerHuntShares.psm1 into memory
IEX(New-Object
System.Net.WebClient).DownloadString("https://raw.githubusercontent.com/NetSPI/PowerHuntShares/main/PowerHuntShares.psm1")
```



# Discovery Output

<https://github.com/NetSPI/powerhuntshares>

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## SHARE DISCOVERY

---

```
[*][03/01/2021 09:35] Scan Start
[*][03/01/2021 09:35] Output Directory: c:\temp\smbshares\SmbShareHunt-03012021093504
[*][03/01/2021 09:35] Successful connection to domain controller: dc1.demo.local
[*][03/01/2021 09:35] Performing LDAP query for computers associated with the demo.local domain
[*][03/01/2021 09:35] - 245 computers found
[*][03/01/2021 09:35] Pinging 245 computers
[*][03/01/2021 09:35] - 55 computers responded to ping requests.
[*][03/01/2021 09:35] Checking if TCP Port 445 is open on 55 computers
[*][03/01/2021 09:36] - 49 computers have TCP port 445 open.
[*][03/01/2021 09:36] Getting a list of SMB shares from 49 computers
[*][03/01/2021 09:36] - 217 SMB shares were found.
[*][03/01/2021 09:36] Getting share permissions from 217 SMB shares
[*][03/01/2021 09:37] - 374 share permissions were enumerated.
[*][03/01/2021 09:37] Getting directory listings from 33 SMB shares
[*][03/01/2021 09:37] - Targeting up to 3 nested directory levels
[*][03/01/2021 09:37] - 563 files and folders were enumerated.
[*][03/01/2021 09:37] Identifying potentially excessive share permissions
[*][03/01/2021 09:37] - 33 potentially excessive privileges were found across 12 systems..
[*][03/01/2021 09:37] Scan Complete
```



# Analysis Output

<https://github.com/NetSPI/powerhuntshares>

## SHARE ANALYSIS

- [\*][03/01/2021 09:37] Analysis Start
- [\*][03/01/2021 09:37] - 14 shares can be read across 12 systems.
- [\*][03/01/2021 09:37] - 1 shares can be written to across 1 systems.
- [\*][03/01/2021 09:37] - 46 shares are considered non-default across 32 systems.
- [\*][03/01/2021 09:37] - 0 shares are considered high risk across 0 systems
- [\*][03/01/2021 09:37] - Identified top 5 owners of excessive shares.
- [\*][03/01/2021 09:37] - Identified top 5 share groups.
- [\*][03/01/2021 09:37] - Identified top 5 share names.
- [\*][03/01/2021 09:37] - Identified shares created in last 90 days.
- [\*][03/01/2021 09:37] - Identified shares accessed in last 90 days.
- [\*][03/01/2021 09:37] - Identified shares modified in last 90 days.
- [\*][03/01/2021 09:37] Analysis Complete



# Share Report Output

<https://github.com/NetSPI/powerhuntshares>

---

## SHARE REPORT SUMMARY

---

```
[*][03/01/2021 09:37] Domain: demo.local  
[*][03/01/2021 09:37] Start time: 03/01/2021 09:35:04  
[*][03/01/2021 09:37] End time: 03/01/2021 09:37:27  
[*][03/01/2021 09:37] Run time: 00:02:23.2759086
```

....

```
[*][03/01/2021 09:37] SHARE ACL SUMMARY  
[*][03/01/2021 09:37] - 374 ACLs were found.  
[*][03/01/2021 09:37] - 374 (100.00%) ACLs were associated with non-default shares.  
[*][03/01/2021 09:37] - 33 (8.82%) ACLs were found to be potentially excessive.  
[*][03/01/2021 09:37] - 32 (8.56%) ACLs were found that allowed READ access.  
[*][03/01/2021 09:37] - 1 (0.27%) ACLs were found that allowed WRITE access.  
[*][03/01/2021 09:37] - 1 (0.27%) ACLs were found that are associated with HIGH-RISK share names
```



# Share Report Output

<https://github.com/NetSPI/powerhuntshares>

## SHARE REPORT SUMMARY

```
[*][03/01/2021 09:37] Domain: demo.local  
[*][03/01/2021 09:37] Start time: 03/01/2021 09:35:04  
[*][03/01/2021 09:37] End time: 03/01/2021 09:37:27  
[*][03/01/2021 09:37] Run time: 00:02:23.275
```

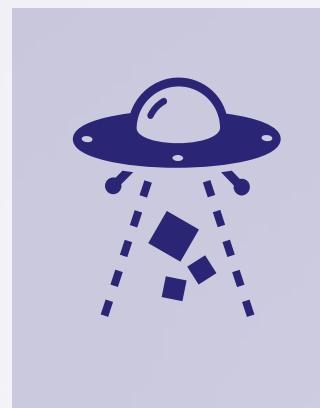
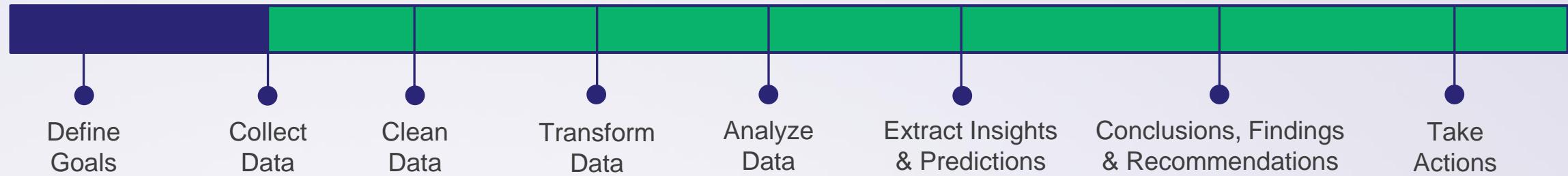
```
[*][03/01/2021 09:37] - 33 (8.82%) ACLs were found to be potentially excessive.  
[*][03/01/2021 09:37] - 32 (8.56%) ACLs were found that allowed READ access.
```

```
[*][03/01/2021 09:37] - 1 (0.27%) ACLs were found that allowed WRITE access.  
[*][03/01/2021 09:37] - 1 (0.27%) ACLs were found that are associated with HIGH-RISK share names
```

**BORING! What's happening under the hood?**



# PowerHuntShares Process

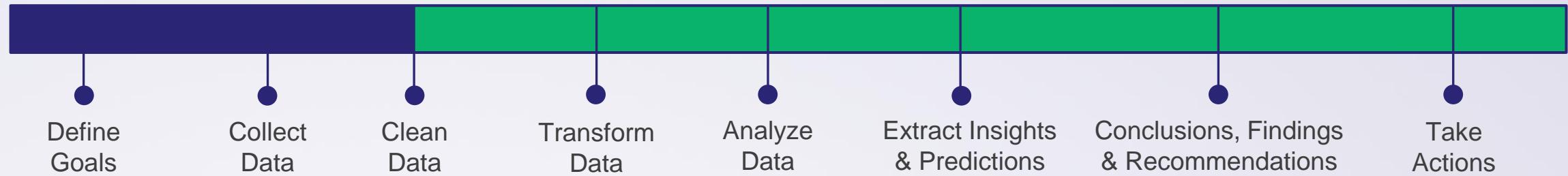


## Data Collection

- **Asset Coverage**  
Active directory query + port connectivity tests + optional ping test
- **Data Visibility**  
Names, dates creation, last modified, and last accessed dates  
Directory listings, hashes of directory listings, file counts



# PowerHuntShares Process



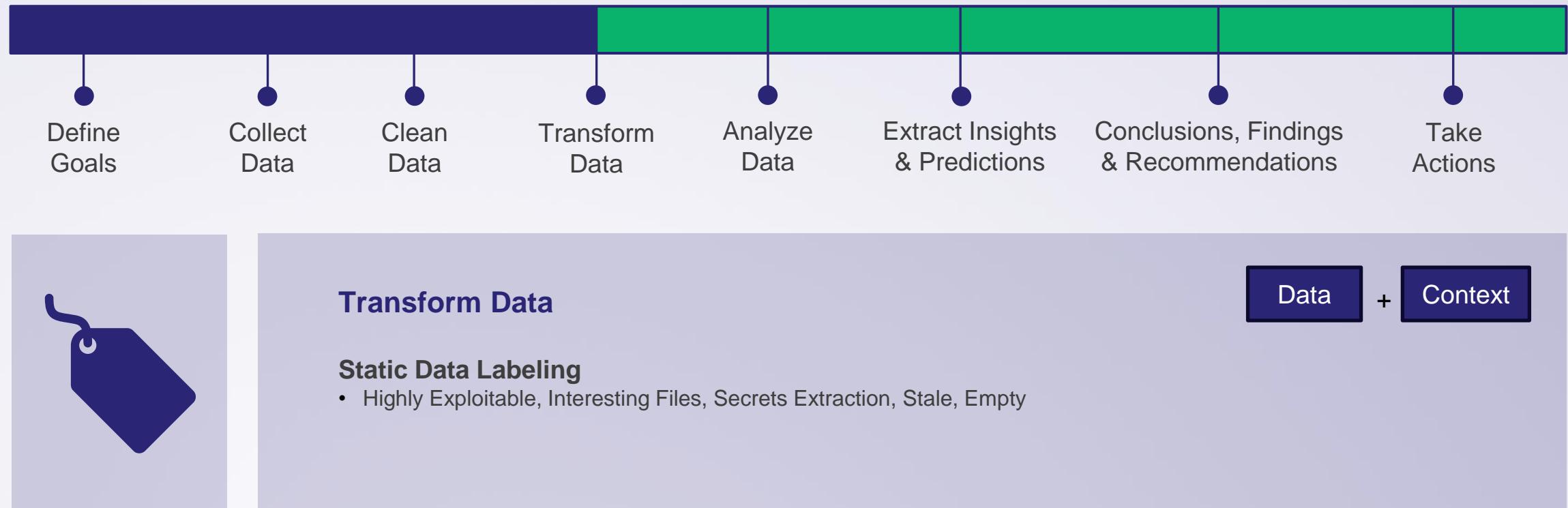
## Data Cleaning

- Parse data
- Normalize data structures
- Fix data type errors
- Remove records with errors
- Filter out unneeded data

Data



# PowerHuntShares Process



# Static Labeling

... mostly :)

- **Highly Exploitable**
- Interesting Files (data and secrets)
- Extracting Secrets
- Stale (last modified date > 1yr)
- Empty (no files)

## Summary

Share folder names that have historically provide attackers with the means to execute code on the system remotely.

Examples:

- C\$
- ADMIN\$
- WWWROOT
- INETPUB



# Static Labeling

... mostly :)

- High Risk Shares
- **Interesting Files** (data and secrets)
- Extracting Secrets
- Stale (last modified date > 1yr)
- Empty (no files)

## Summary

~ 200 file names, keywords and extensions used to label files and folders that may be used to execute remote code execution or expose sensitive data.

Examples:

- Known password files.
- Known data files.
- Interesting keywords in file name.
- Interesting file extensions.

Note: The list can be extended at run time using a file template.

## RESULTS

 Summary Report Scan Information

## EXPLORE

 Networks Computers Share Names Folder Groups Insecure ACEs Identities ShareGraph

## TARGET

 Interesting Files Extracted Secrets

## ACT

 Exploit Detect Remediate

# Interesting Files

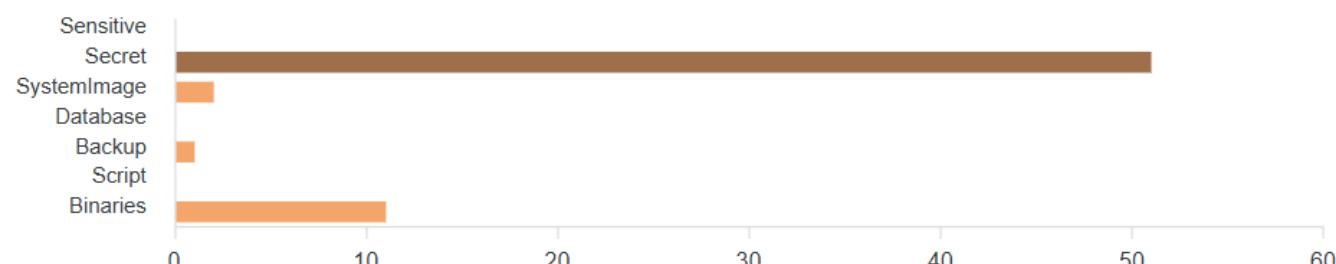
This section provides a list of files that may contain passwords or sensitive data, or may be abused for remote code execution.

## Interesting Files Found

**83**

(65 unique file names)

## Interesting File Exposure

65 matches found [Export](#) | [Clear](#)

## Search

File Count	File Name	Category	File Paths
5	program files	Binaries	5 Files
3	program files (x86)	Binaries	3 Files
3	system	Secret	3 Files
2	backup	Backup	2 Files
2	bfsvc.exe	Binaries	2 Files

# Static Labeling

... mostly :)

- High Risk Shares
- Interesting Files (data and secrets)
- **Extracting Secrets**
- Stale (last modified date > 1yr)
- Empty (no files)

# Summary

**50 functions** to automatically extract passwords from known configuration files.

## Examples

- Web.config
- App.config
- Machine.config
- Unattend.xml
- My.cnf
- Tomcat-users.xml
- Cisco Startup/Run Configs – Type 7 decoding
- Smb.conf
- Krb5.conf
- Shadow



# Static Labeling

... mostly :)

- High Risk Shares
- Interesting Files (data and secrets)
- **Extracting Secrets**
- Stale (last modified date > 1yr)
- Empty (no files)

## Summary

**~1 day of development using LLM prompt**

### Process Summary

1. Ask for top ten applications that store credentials in common categories.
2. Ask for links to sample configuration files and download them.
3. Create prompt to generate PowerShell functions to parse passwords based on a provided configuration file.
4. **Submit prompt with configuration file**
5. **~30% required small modifications.**
6. **Repeat.**



# Static Labeling

... mostly :)

- High Risk Shares
- Interesting Files (data and secrets)
- **Extracting Secrets**
- Stale (last modified date > 1yr)
- Empty (no files)

## Sample Prompt

1. Create a PowerShell function that parses usernames and passwords from the provided example file.
2. Ensure the PowerShell function supports an input parameter named “FilePath” that accepts a path to the configuration file so it can be read and parsed.
3. Ensure all output is provided as a PSObject. Ensure each parsed username and password pair is returned as a separate record. Output parameters should include “username” and “password”. If their values are empty in the file, then return “EMPTY” for their values in the PSObject.

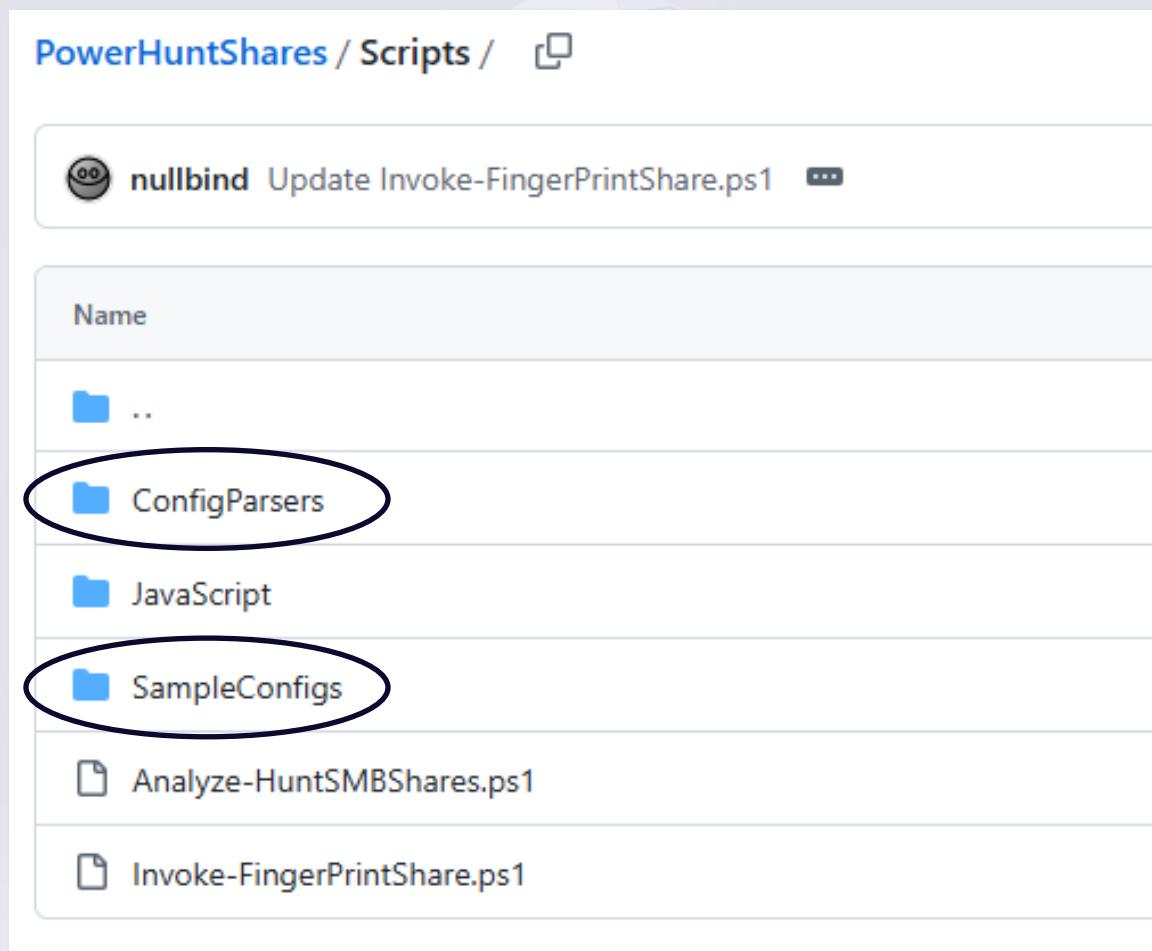
Example Configuration File:  
[Content Here](#)



# Static Labeling

... mostly :)

- High Risk Shares
- Interesting Files (data and secrets)
- **Extracting Secrets**
- Stale (last modified date > 1yr)
- Empty (no files)





## Extracted Secrets

This section includes a list of the credentials that were recovered during data collection. 143 credentials were recovered from 50 of the discovered 53 secrets files.

### Extracted Secrets Found

**143**143 matches found [Export](#) | [Clear](#)

Search

ComputerName	ShareName	FileName	FilePath	Username	Password	PasswordEnc	KeyfilePath	Details
2012SERVERSCCM. demo.local	files	bootstrap.ini	\\\2012SERVERS CCM.demo.loc al\files\bootstr ap.ini	adminUser	P@ssw0rd123	NA	NA	<a href="#">Details</a>
2012SERVERSCCM. demo.local	files	bootstrap.ini	\\\2012SERVERS CCM.demo.loc al\files\bootstr ap.ini	NA	public	NA	NA	<a href="#">Details</a>
2012SERVERSCCM. demo.local	files	bootstrap.ini	\\\2012SERVERS CCM.demo.loc al\files\bootstr ap.ini	NA	mysecret	NA	NA	<a href="#">Details</a>
2012SERVERSCCM. demo.local	files	bootstrap.ini	\\\2012SERVERS CCM.demo.loc al\files\bootstr ap.ini	NA	mysecret	NA	NA	<a href="#">Details</a>
2012SERVERSCCM. demo.local	files	bootstrap.ini	\\\2012SERVERS CCM.demo.loc al\files\bootstr ap.ini	NA	mykey	NA	NA	<a href="#">Details</a>

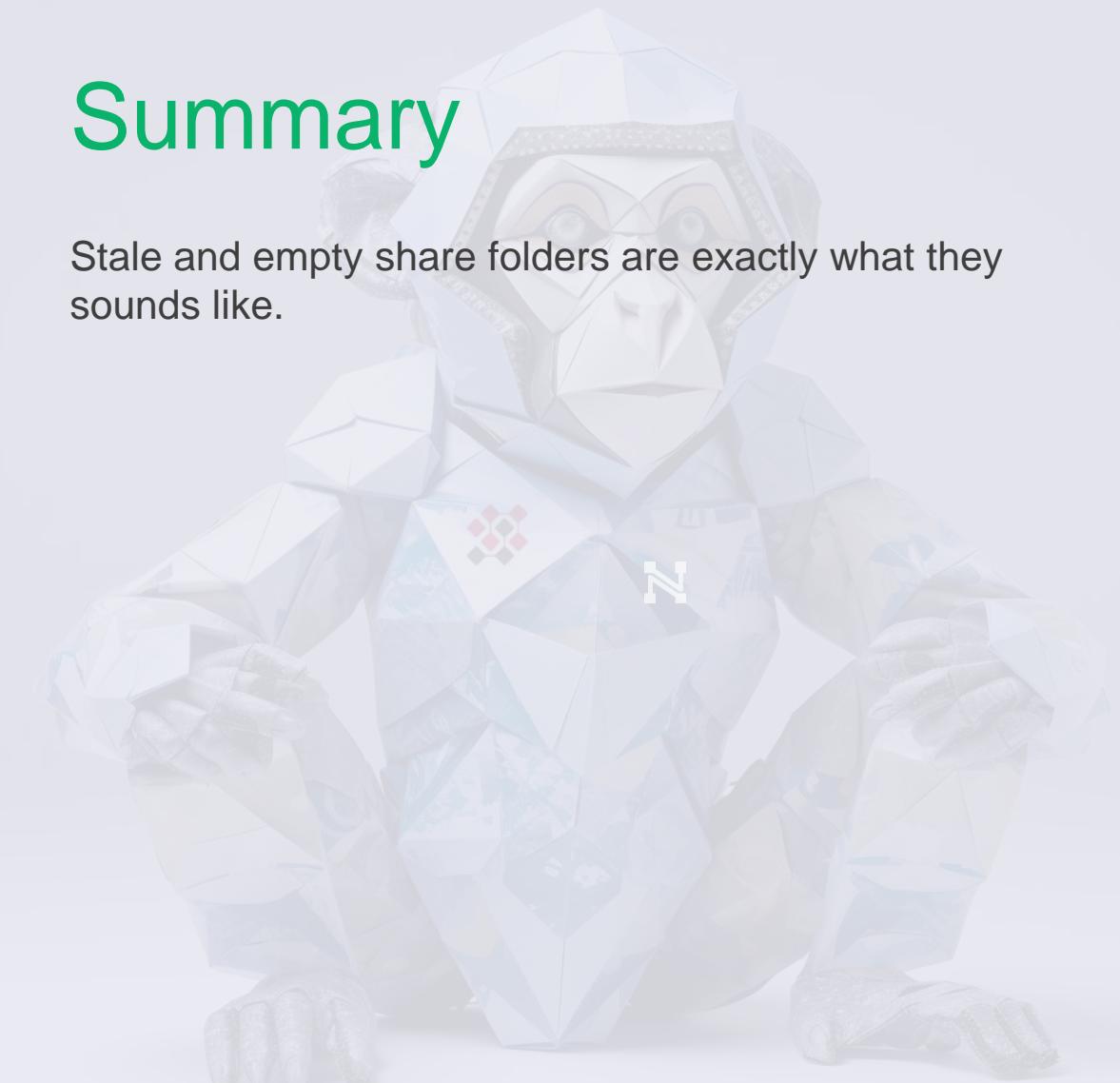
# Static Labeling

... mostly :)

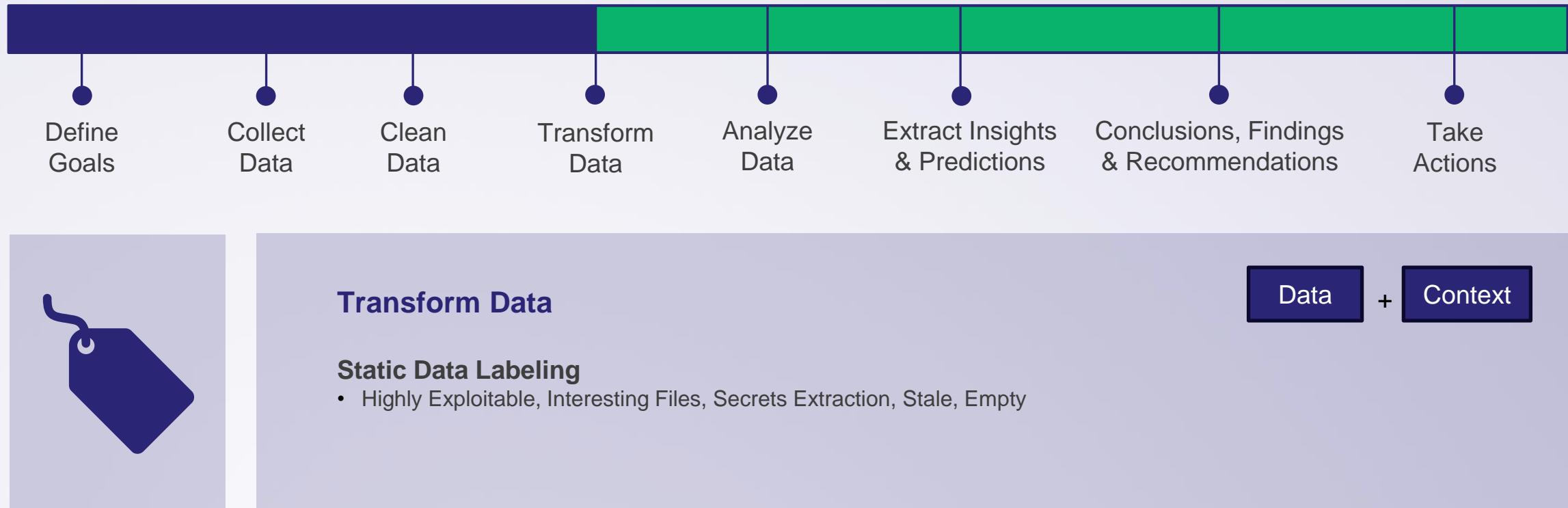
- High Risk Shares
- Interesting Files (data and secrets)
- Extracting Secrets
- **Stale (last modified date > 1yr)**
- **Empty (no files)**

## Summary

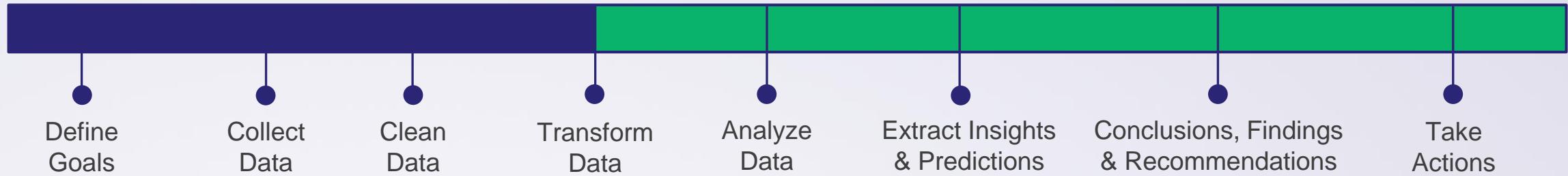
Stale and empty share folders are exactly what they sounds like.



# PowerHuntShares Process



# PowerHuntShares Process



## Transform Data

### Static Data Labeling

- Highly Exploitable, Interesting Files, Secrets Extraction, Stale, Empty

### Dynamic Data Enrichment

- Fingerprinting

Data + Context  
+  
Context



# Share Fingerprinting



*“What is this share used for?”*



# Share Fingerprinting



*"What is this share used for?"*

## Why Fingerprint Shares?

### Improve Offensive Context

- Increase confidence that a share contains specific files with stored secrets, sensitive data or can be used for remote code execution.

### Improve Defensive Context

- Better understand the impact of removing potentially excessive privilege.
- Increase confidence the share or group of shares are related to a specific application or process that can be remediated at the same time.

# Share Fingerprinting



*"What is this share used for?"*

- Static Hardcoded Application Fingerprint Library

## Summary

- ~ **100** environments manually analyzed
- ~ **80** share names mapped to common applications and operating systems

## Pros

- Better than what I had, which was nothing. 😊
- Includes descriptions for the shares and related apps.

## Cons

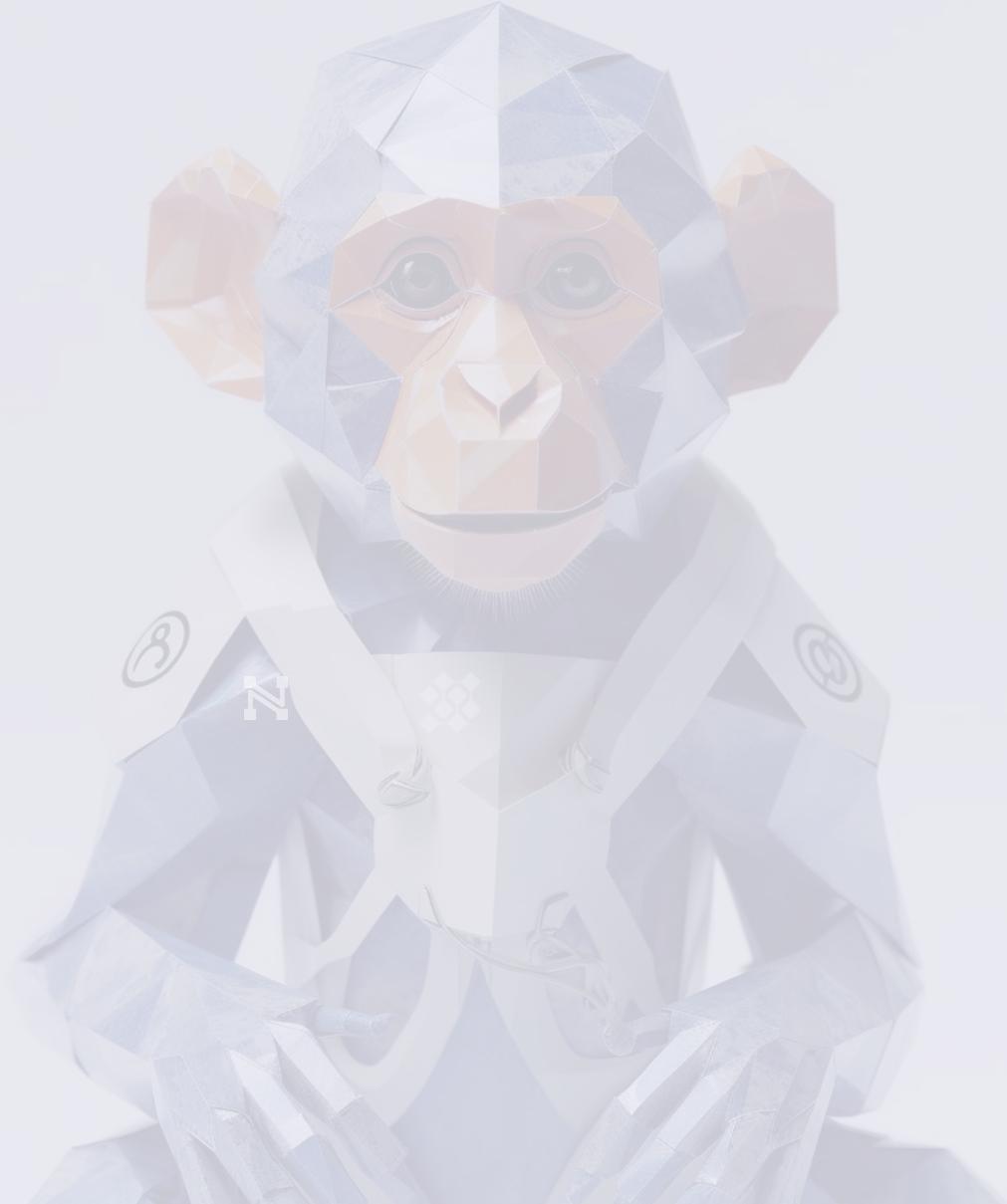
- Doesn't consider file listings which can lead to false positives.
- Doesn't include any fuzzy logic to account for share name variations which can lead to false negatives.
- Currently doesn't output CPE.

# Share Fingerprinting

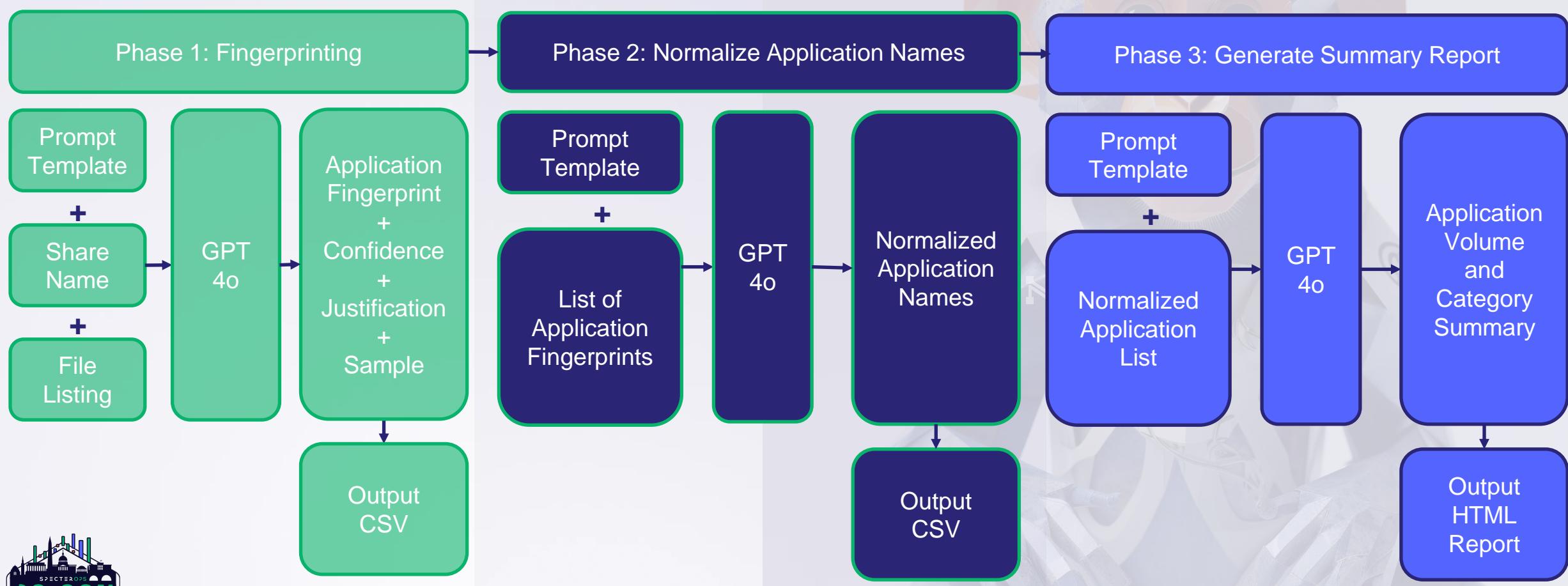


*"What is this share used for?"*

- Static Hardcoded Application Fingerprint Library
- **Dynamic LLM-Based Application Fingerprinting**



# Share Fingerprinting LLM-Based Process



# Share Fingerprinting LLM-Based



## Asset Exposure Summary

47 ACL entries, on 16 shares, hosted by 2 computers were found configured with excessive privileges on the demo.local domain. In this environment, we observed a total of 19 application instances, with 4 unique application names primarily focused on operating systems, configuration management, virtualization, and security. The Windows Operating System had the highest count with 10 instances (52.63% of the total), followed by Microsoft System Center Configuration Manager with 3 instances (15.79% of the total).

### Networks

**1**

affected

### Computers

**2**

affected

### Shares

**16**

affected

### ACEs

**47**

affected

Note: Application fingerprints were generated using an experimental version of the LLM-based application fingerprinting function. As a result, some application classifications may not be accurate.



## Phase 1: Fingerprinting

### Prompt Template



Share Name

GPT 4o



File Listing

# Share Fingerprinting

LLM-Based



Phase 1: Fingerprinting

Prompt  
Template

+  
Share  
Name

+  
File  
Listing

GPT  
4o

## Asset Exposure Summary

47 ACL entries, on 16 shares, hosted by 2 computers in the demo.local domain. In this environment, we observed application names primarily focused on operational security. The Windows Operating System had the most shares followed by Microsoft System Center Configuration Manager.

### Networks

1  
affected

### Shares

16  
affected

47  
affected

PowerHuntShares / Scripts /



nullbind Update Invoke-FingerPrintShare.ps1

Name



...



ConfigParsers



JavaScript



SampleConfigs



Analyze-HuntSMBShares.ps1



Invoke-FingerPrintShare.ps1



Note: Application fingerprints were generated using an experimental version of the LLM-based application fingerprinting function. As a result, some application classifications may not be accurate.

# Share Fingerprinting



*"What is this share used for?"*

- Static Hardcoded Application Fingerprint Library
- **Dynamic LLM-Based Application Fingerprinting**

## Lessons Learnd

- Large context windows != Accuracy
- Break problem into smaller parts
- Use explicit instructions
- Run multiple iterations
- Generate confidence scores
- Generate justification
- XML > JSON

# Share Fingerprinting



*"What is this share used for?"*

- Static Hardcoded Application Fingerprint Library
- **Dynamic LLM-Based Application Fingerprinting**

## Summary

### Pros

- Can account for things I've never seen before.

### Cons

- We still have some hallucinations.
- Does not include vendor name is a separate field.
- Does not output CPE in the current version.

**RESULTS**[Summary Report](#)[Scan Information](#)**EXPLORE**[Networks](#)[Computers](#)[Share Names](#)[Folder Groups](#)[Insecure ACES](#)[Identities](#)[ShareGraph](#)**TARGET**[Interesting Files](#)[Extracted Secrets](#)**ACT**[Exploit](#)[Detect](#)[Remediate](#)14 matches found [Export](#) | [Clear](#)

Search

Quick Filters:  Exploitable  Write  Read  Interesting  Empty  Stale  Default

Share Count	Share Name
2	C\$

[H](#) [W](#) [R](#) [I](#) [S](#)Sample Description  
Default share

## Share Context Guess

The C\$ may be associated with the Windows Admin Share. An administrative share for remote management. C\$ is a default administrative share in Windows. C:\Windows\System32 is the expected local path.

## LLM Application Guess

Windows Operating System, Microsoft System Center Configuration Manager

[View in ShareGraph](#)

## Affected Assets

Computers: 2 of 13 (15.38%)

Shares: 2 of 21 (9.52%)

ACLs: 6 of 127 (4.72%)

## Timeline Context

First Created: 07/26/2012

Last Created: 07/26/2012

Last Mod: 11/06/2024

## Owners (1)

NT SERVICE\TrustedInstaller

2 ADMIN\$

[H](#) [R](#) [I](#) [S](#)

1 backup

[W](#) [R](#) [E](#) [S](#)

1 inetpub

[H](#) [W](#) [R](#) [E](#) [S](#)

1 sccm

[W](#) [R](#) [E](#) [S](#)**C\$**[H](#) [W](#) [R](#) [I](#) [S](#)**24 Critical****Sample Description**

Default share

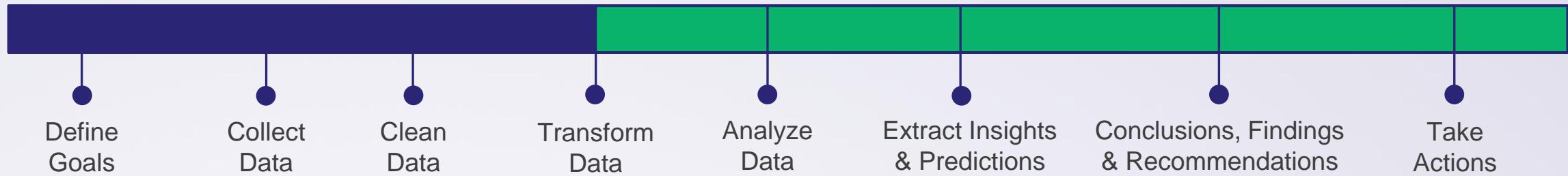
**Share Context Guess**

The C\$ may be associated with the Windows Admin Share. An administrative share for remote management. C\$ is a default administrative share in Windows. C:\Windows\System32 is the expected local path.

**LLM Application Guess**

Windows Operating System, Microsoft System Center Configuration Manager

# PowerHuntShares Process



## Transform Data

### Static Data Labeling

- Highly Exploitable, Interesting Files, Secrets Extraction, Stale, Empty

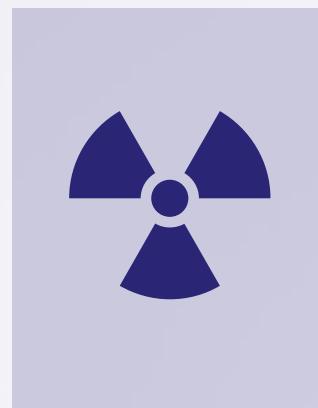
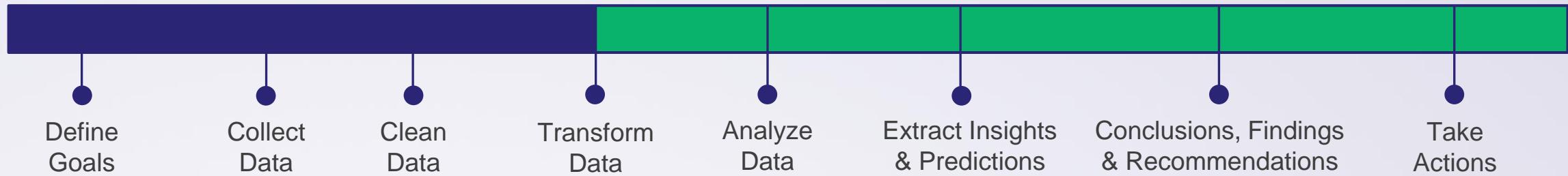
### Dynamic Data Enrichment

- **Fingerprinting**

Data + Context  
+  
Context



# PowerHuntShares Process



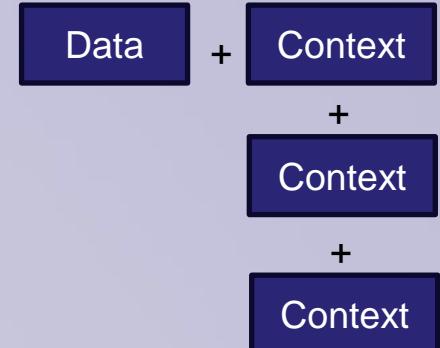
## Transform Data

### Static Data Labeling

- Highly Exploitable, Interesting Files, Secrets Extraction, Stale, Empty

### Dynamic Data Enrichment

- Fingerprinting, **Risk Scoring**



# Risk Scoring

*"Be honest, how bad is it?"*



# Risk Scoring

*"Be honest, how bad is it?"*

- Summary

## Summary

The PowerHuntShares **risk score** is a simple formula that helps evaluate and rank risk associated with shares based simple questions.



# Risk Scoring

*"Be honest, how bad is it?"*

- Summary
- Why Risk Scores?

## Summary

The PowerHuntShares **risk score** is a simple formula that helps evaluate and rank risk associated with shares based on simple questions.

## Why Risk Scores?

- Help prioritize exploitation
- Help prioritize remediation
- Add context related to abuse impact

## Why Another Risk Rating?

- **CVSS** didn't provide the data context and volume in the way I wanted.



# Risk Scoring

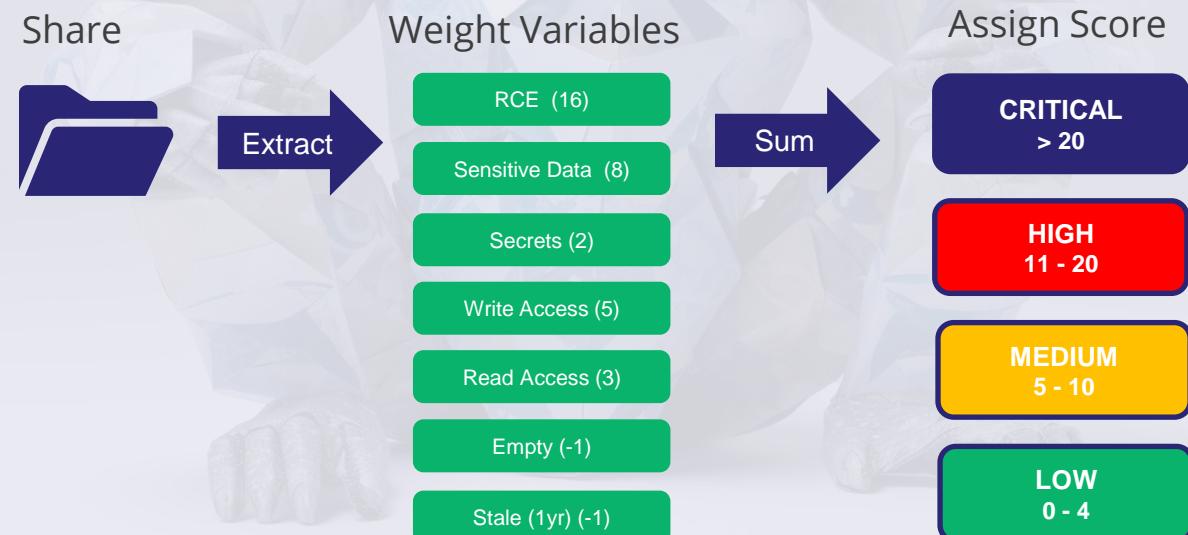
*"Be honest, how bad is it?"*

- Summary
- Why Risk Scores
- **Formula Abstract**

## Summary

The PowerHuntShares **risk score** is a simple formula that helps evaluate and rank risk associated with shares based on simple questions.

## Formula Abstract



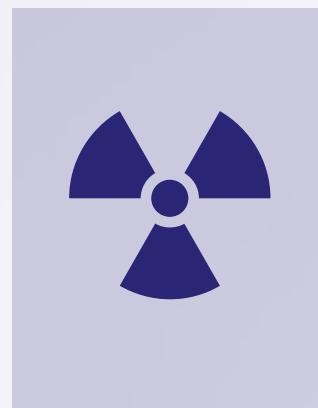
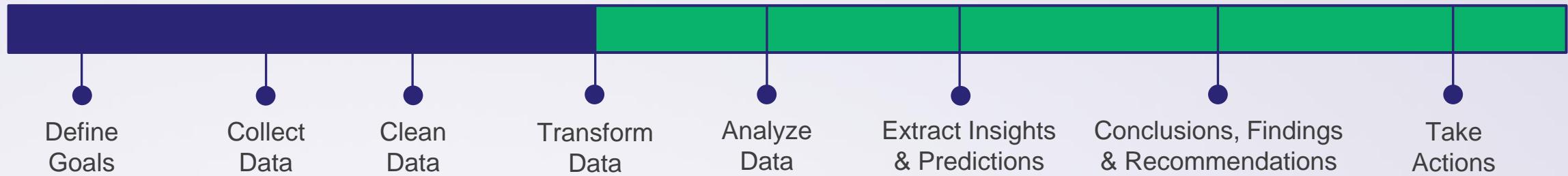


Search

Quick Filters:  Exploitable  Write  Read  Interesting  Empty  Stale  Default

Share Count ⓘ	Share Name ⓘ	Risk Level ⓘ	Share Similarity ⓘ	Folder Groups ⓘ	Common Files ⓘ	Interesting Files ⓘ
2	C\$ 	24 Critical	84% High	2	6 Files	6 Files
2	ADMIN\$ 	20 Critical	84% High	2	74 Files	11 Files
1	backup 	3 Low	100% Very High	1	0 Files	0 Files
1	inetpub 	21 Critical	100% Very High	1	0 Files	0 Files
1	sccm 	3 Low	100% Very High	1	0 Files	0 Files
1	logs 	3 Low	100% Very High	1	0 Files	0 Files
1	sql 	3 Low	100% Very High	1	0 Files	0 Files
1	C 	22 Critical	100% Very High	1	12 Files	3 Files
1	apps 	3 Low	100% Very High	1	0 Files	0 Files
1	wwwroot 	21 Critical	100% Very High	1	0 Files	0 Files

# PowerHuntShares Process



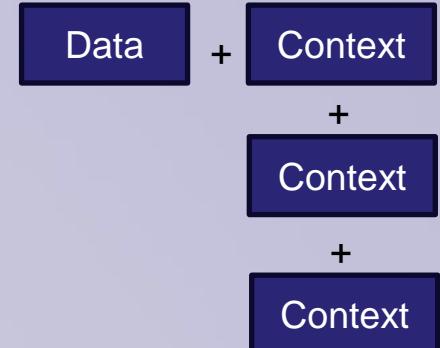
## Transform Data

### Static Data Labeling

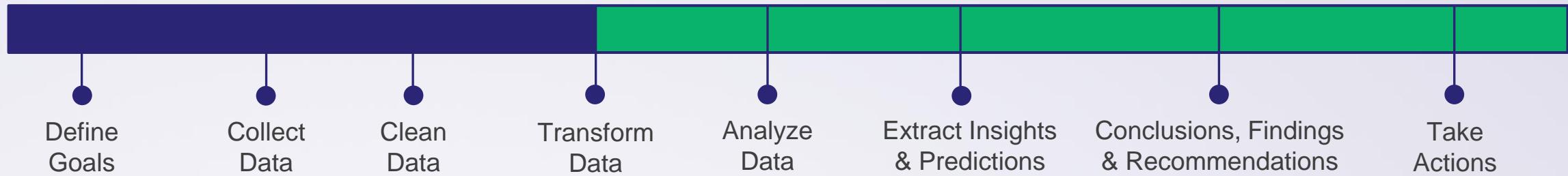
- Highly Exploitable, Interesting Files, Password, Extraction, Stale, Empty

### Dynamic Data Enrichment

- Fingerprinting, **Risk Scoring**



# PowerHuntShares Process



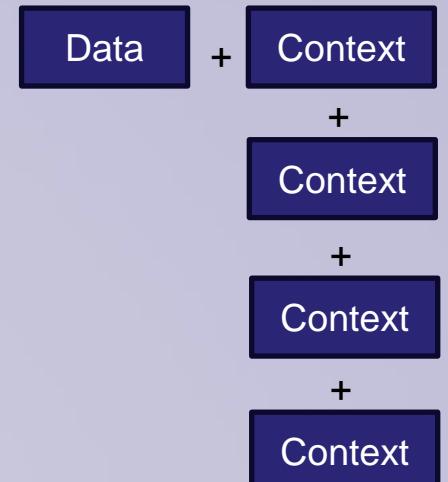
## Transform Data

### Static Data Labeling

- Highly Exploitable, Interesting Files, Password Extraction, Stale, Empty

### Dynamic Data Enrichment

- Fingerprinting, Risk Scoring, **Peer Comparison**



# Peer Comparison

“ So, we have 1,000 critical risk shares, really?...

...Good to know, but how do we compare to our peers? ”



## Summary

Companies want to understand what's normal and where they fall short and when they are overachieving.

## Use Cases

1. Acquire Budget.
2. Use as KPI.

## Tested Approaches

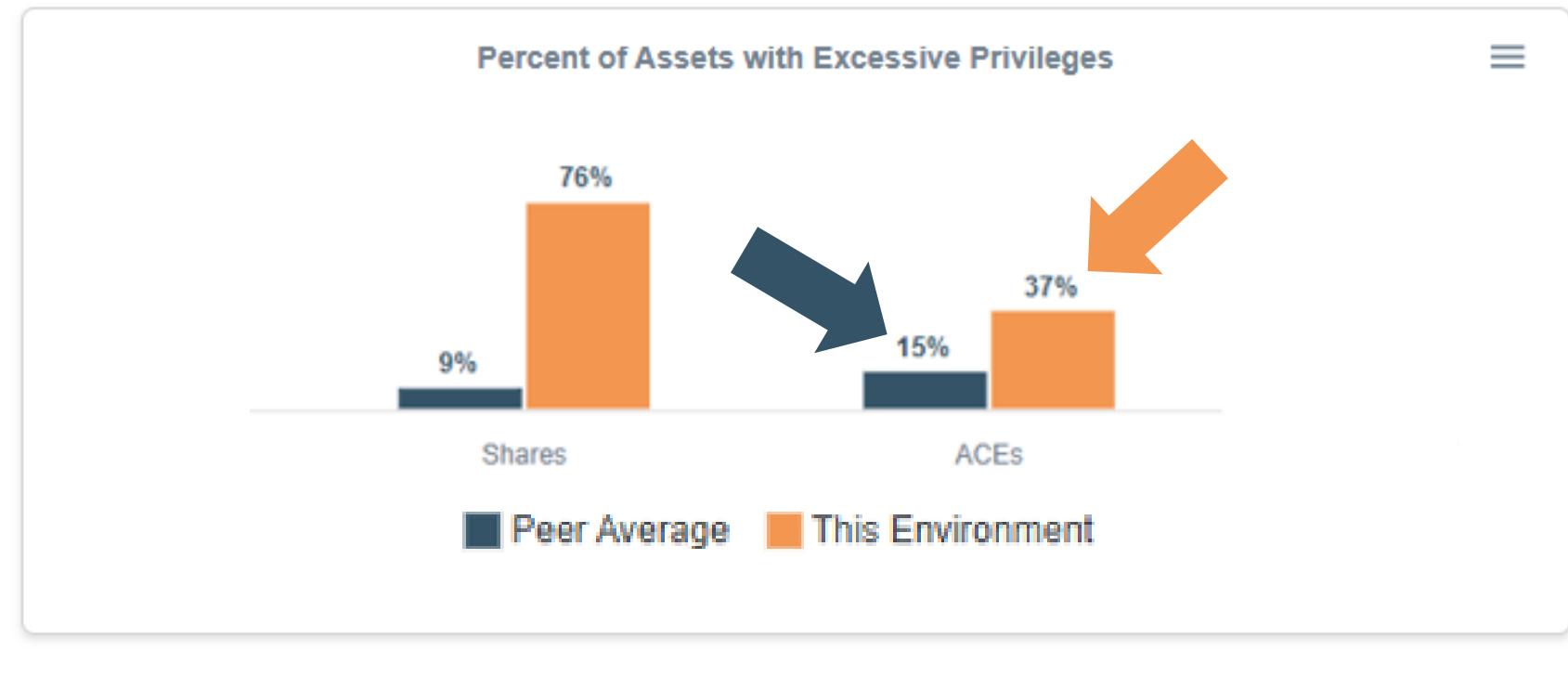
- **Do nothing.** PowerHuntShares v1
- **Historical Averages.** PowerHuntShares v2
- **Predictive Models.** PowerHuntShares v3?

# Peer Comparison

## Historical Average

### Affected Asset Peer Comparison

Below is a comparison between the percent of affected assets in this environment and the average percent of affected assets observed in other environments. The percentage is calculated based on the total number of live assets discovered for each asset type. Based on the volume of ACEs configured with excessive privileges, this environment was less secure compared to the average.



# Peer Comparison

## Predictive Models

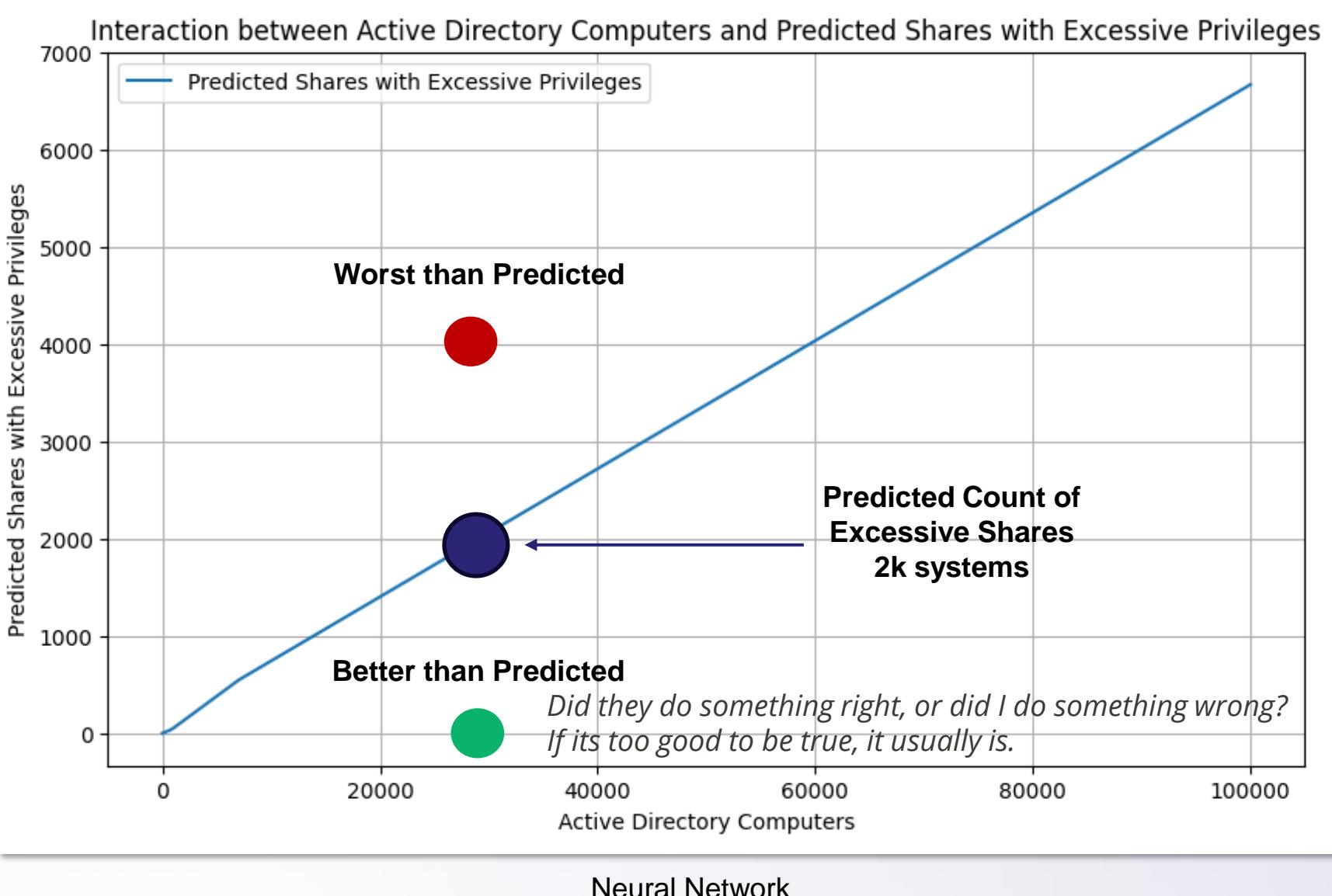
Linear Regression

Randomforest

Neural Network

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import mean_squared_error
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
import matplotlib.pyplot as plt
import shap
```

```
# Load data
file_path = r"C:\tools\data\share_privilege.csv"
df = pd.read_csv(file_path)
```



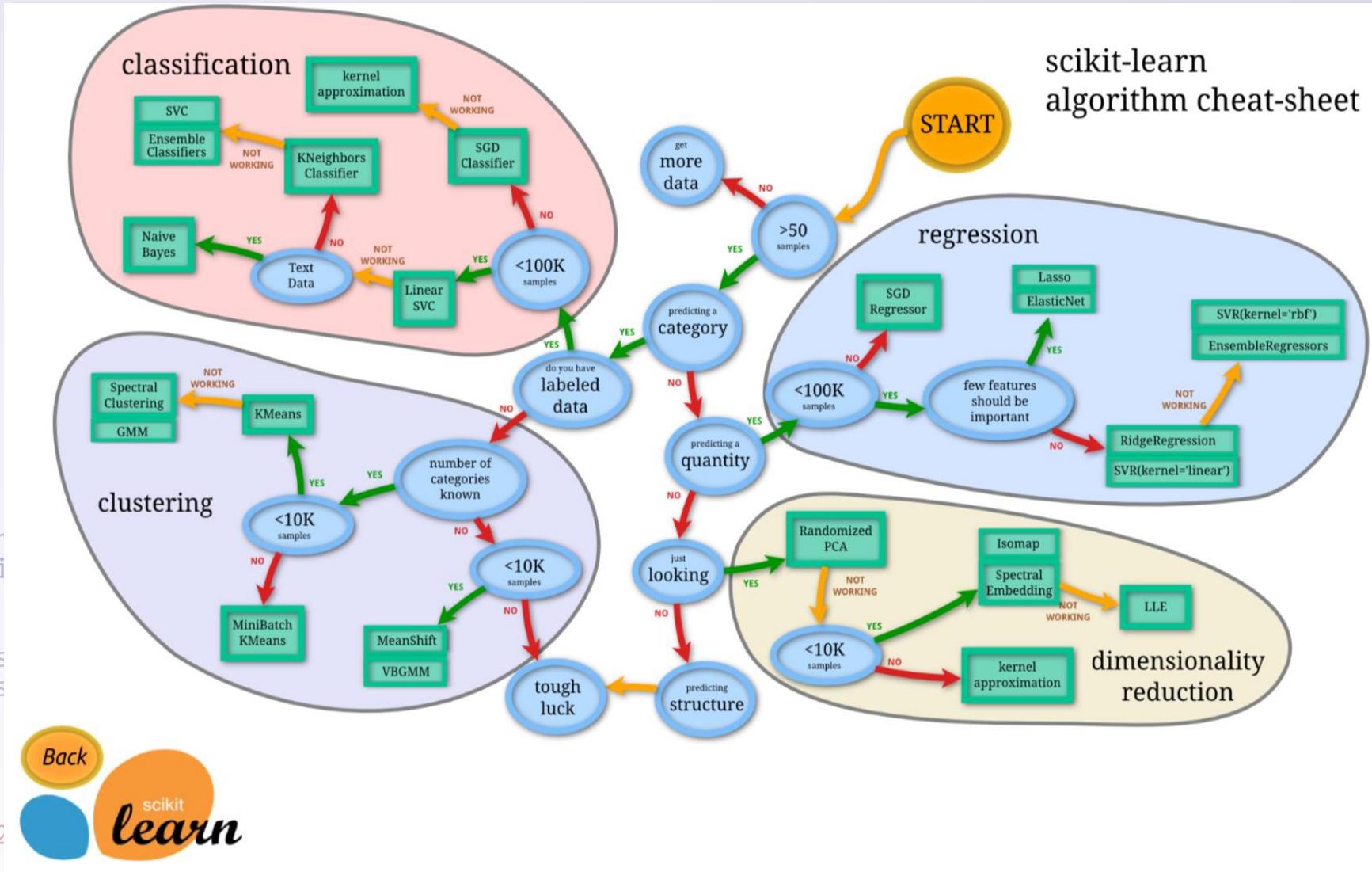
# Peer Comparison

## Predictive Models

Linear Regression  
Randomforest  
Neural Network

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Load data
file_path = r"C:\tools\data\iris.csv"
df = pd.read_csv(file_path)
```



# Peer Comparison

## Predictive Models

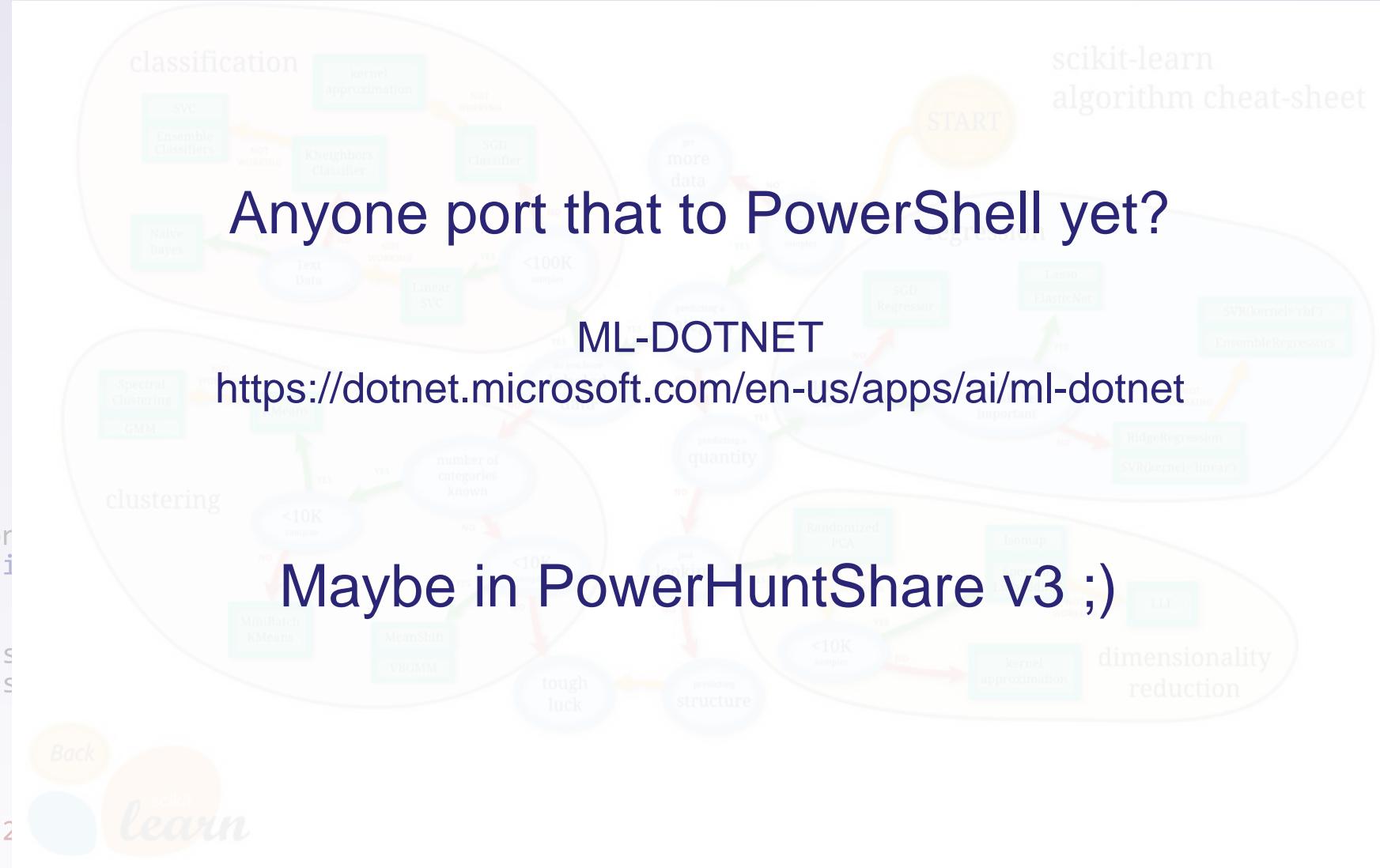
Linear Regression

Randomforest

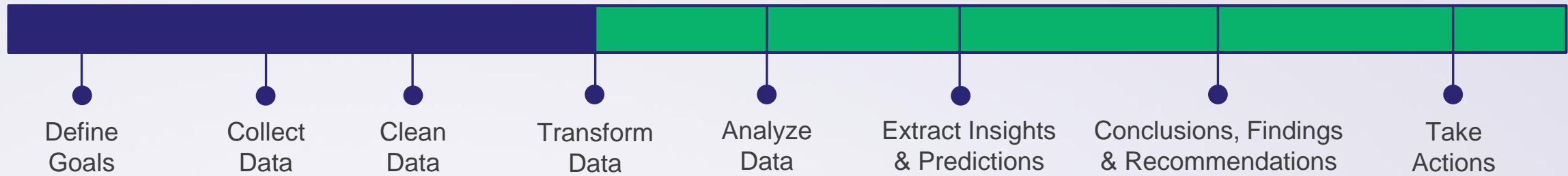
Neural Network

```
import pandas as pd
import numpy as np
from sklearn.model_selection
from sklearn.preprocessing i
from sklearn.metrics import
import tensorflow as tf
from tensorflow.keras.models
from tensorflow.keras.layers
import matplotlib.pyplot as
import shap
```

```
# Load data
file_path = r"C:\tools\data2
df = pd.read_csv(file_path)
```



# PowerHuntShares Process



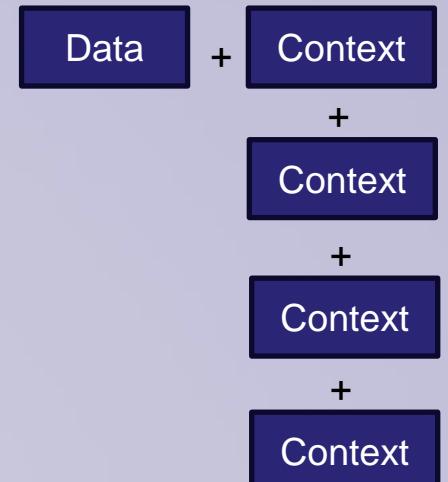
## Transform Data

### Static Data Labeling

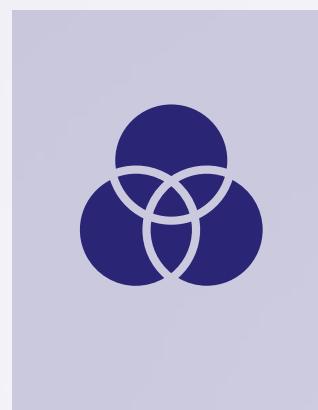
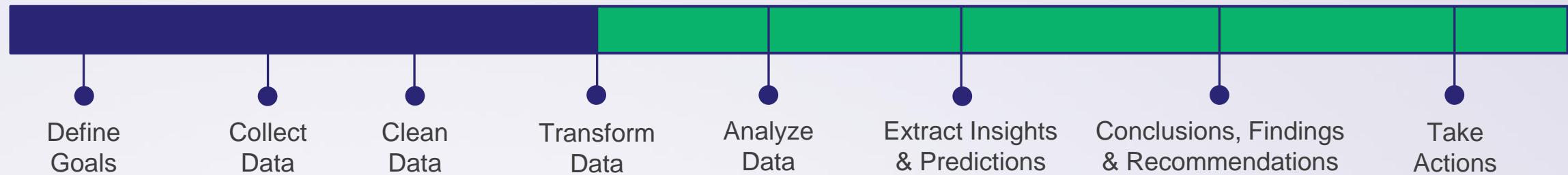
- Highly Exploitable, Interesting Files, Secrets Extraction, Stale, Empty

### Dynamic Data Enrichment

- Fingerprinting, Risk Scoring, **Peer Comparison**



# PowerHuntShares Process



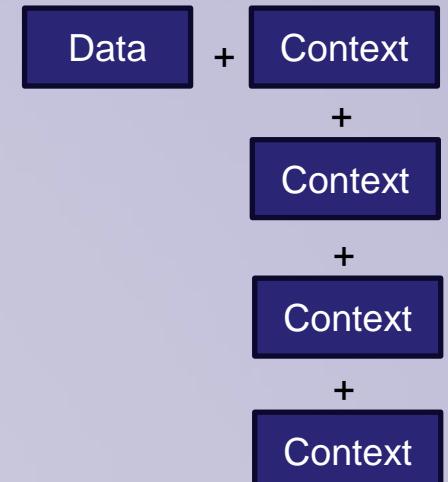
## Transform Data

### Static Data Labeling

- Highly Exploitable, Interesting Files, Secrets Extraction, Stale, Empty

### Dynamic Data Enrichment

- Fingerprinting, Risk Scoring, Peer Comparison, **Grouping & Similarity Scoring**



# Grouping & Similarity

*"How can I group similar shares so I can take fewer targeted actions?"*



# Grouping & Similarity

*"How can I group similar shares so I can take fewer targeted actions?"*

## Why Group Shares?

### Offensive Action Target Consolidation

- Secrets extraction
- Sensitive data extractions
- Remote code execution

### Defensive Action Target Consolidation

- Groups assets part of the same process or application with confidence
- Prioritize large groups of vulnerable assets at once
- Remediate groups of similar assets at the same time



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by **Share Name**

## Summary

Group shares together by their name as the sole means of determining similarity.



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by **Share Name**

## Summary

Group shares together by their name as the sole means of determining similarity.

Share Name

Apps

Logs

Logs

C\$

The shares named “logs”  
get grouped together.



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by **Share Name**

## Summary

Group shares together by their name as the sole means of determining similarity.

## Pros

- Fast and easy to execute via common query syntax.
- Works great if the shares were created to support the same process or application at the same time.

## Cons

- Works poorly if shares have the same name but they are NOT related. **Which happens a lot.**
- Works poorly when you want to consider other factors like, *who owns the shares, data exposure risk, rce risk, or when shares were created, modified, or accessed*.



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by **Share Name**

## Summary

Group shares together by their name as the sole means of determining similarity.

## Example Queries

### SQL QUERY

```
SELECT ShareName, COUNT(ShareName) AS ShareCount  
FROM Shares  
GROUP BY ShareName  
ORDER BY ShareCount DESC;
```

### PowerShell Example

```
$Shares | Group-Object | Sort-Object Count -Descending |  
Select-Object Count, Name
```



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by **Share Name**

## Summary

Group shares together by their name as the sole means of determining similarity.

## Example Output

ShareName	Count
Logs	2
Apps	1
C\$	1



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by **Folder Group (Dir Hash)**

## Summary

Folder groups are MD5 hashes of a share's file listing.



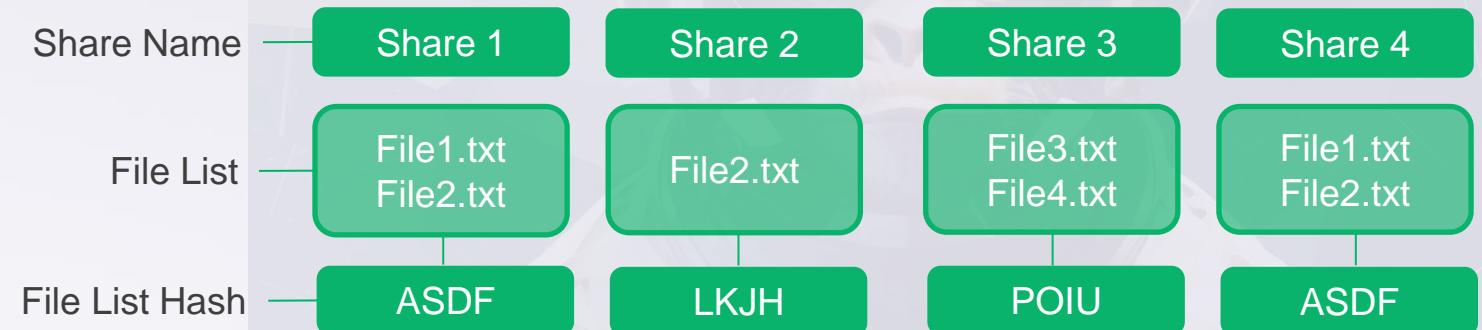
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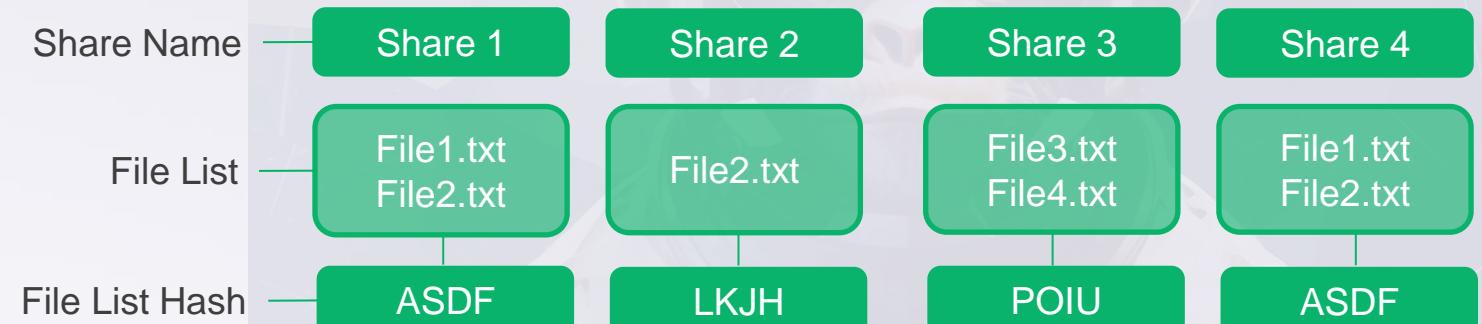
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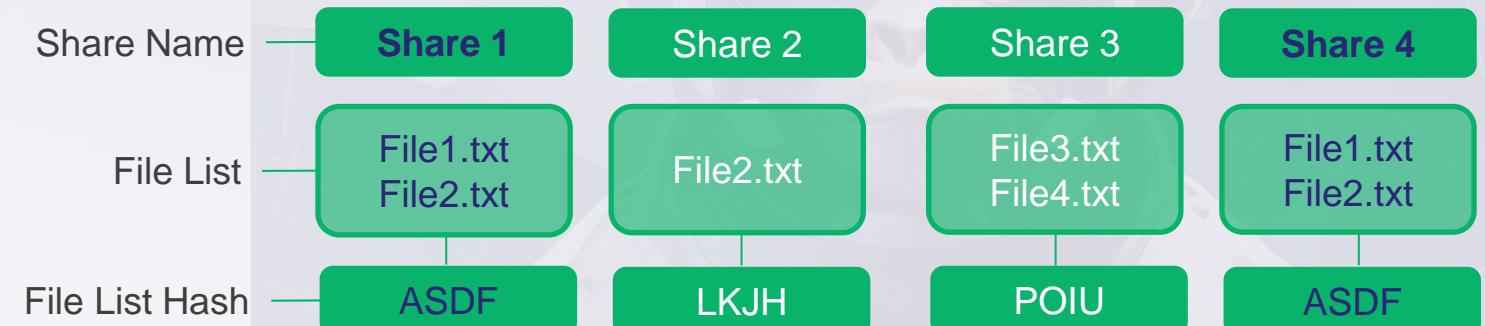


Share 1 & 4  
will have the same  
"Folder Group" hash.

# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by **Folder Group (Dir Hash)**



# Summary

Folder groups are MD5 hashes of a share's file listing.

Share 1 & 4  
are have EXACTLY  
the same file list

# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by **Folder Group (Dir Hash)**

## Summary

Folder groups are MD5 hashes of a share's file listing.

## Pros

- Condensed representation of fil list for quick display, filtering and sorting.
- Fast and easy to execute via common query syntax & functions.
- Great for finding shares that have the EXACT SAME list of files at the root level.

## Cons

- Works poorly when the shares DO NOT have the exact same list of files but are used by the same application.  
**Which happens a lot.**
- Folder groups functionality in PowerHuntShares does not currently include nested folder listings.





## RESULTS

[Summary Report](#)[Scan Information](#)

## EXPLORE

[Networks](#)[Computers](#)[Share Names](#)[Folder Groups](#)[Insecure ACEs](#)[Identities](#)[ShareGraph](#)

## TARGET

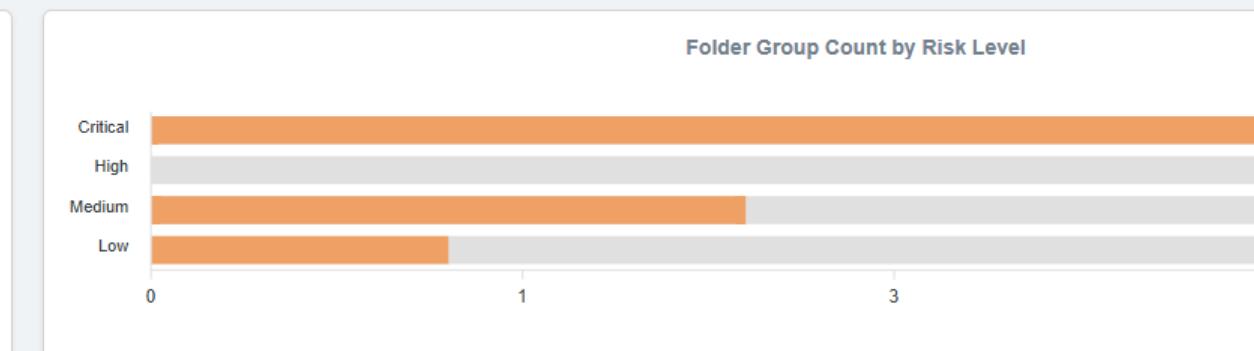
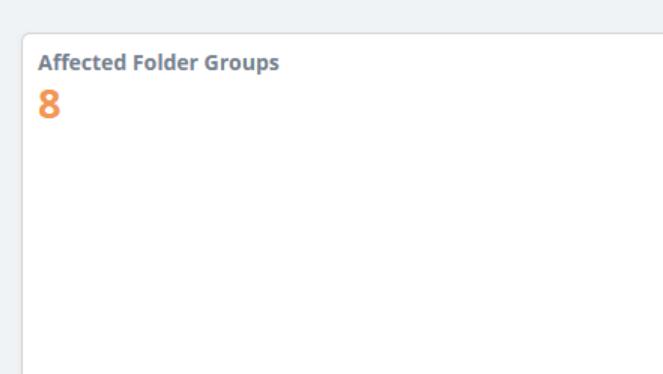
[Interesting Files](#)[Extracted Secrets](#)

## ACT

[Exploit](#)[Detect](#)[Remediate](#)

## Folder Groups

Folder groups are SMB shares that contain the exact same file listing. Each folder group has been hashed so they can be quickly correlated. In some cases, shares with the exact same file listing may be related to a single application process. This information can help identify the root cause associated with the excessive privileges and expedite remediation. Note: Application fingerprints were generated using an experimental version of the LLM-based application fingerprinting function. As a result, some application classifications may not be accurate.



Unique Share Names	Share Count	File Count	Risk Level	Folder Group	Related App
8	8	0 Files	21 Critical	d41d8cd98f00b204e9800998ecf8427e	
2 C\$ C	2 \demo.local\C\$ \demo.local\C	12 Files apps backup inetpub logs PerfLogs Program Files Program Files (x86) sccm sql Users Windows wwwroot	22 Critical	003fe65715d4b71b68e7e42d2cbfd11f	Windows Operating System
1	1	52 Files	8 Medium	608fe6cb11c8dd935745fdfbce83c5be	
1	1	14 Files	24 Critical	f910ff7451dc52f16511bc1858288a7b	Microsoft System Center Configuration Manager

# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by Folder Group (Dir Hash)
- Group by **Merkle Hash (Nested Dir Hash)**

## Summary

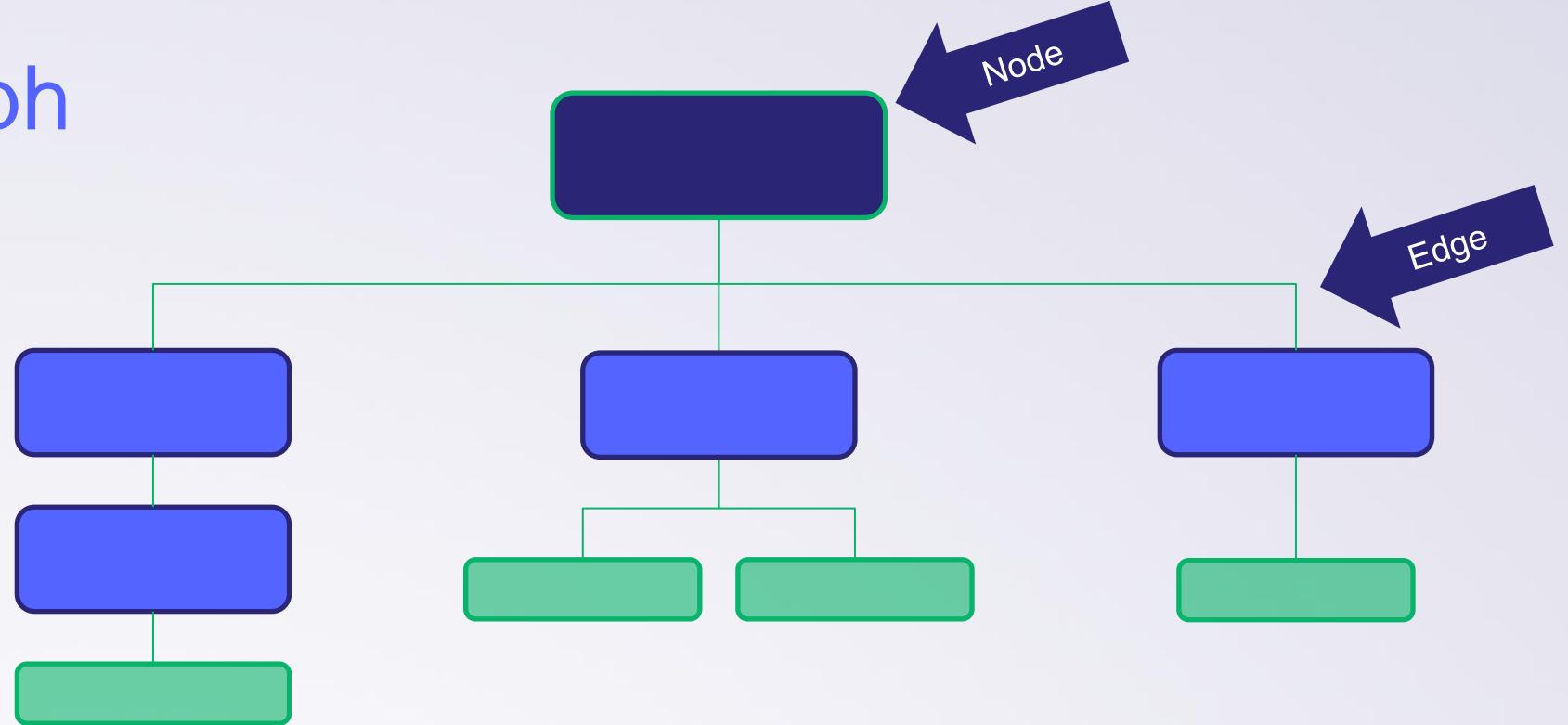
A Merkle Tree is hashing technique that can be applied to any hierachal structures and has been traditionally used for data integrity validation.



# Merkle Trees

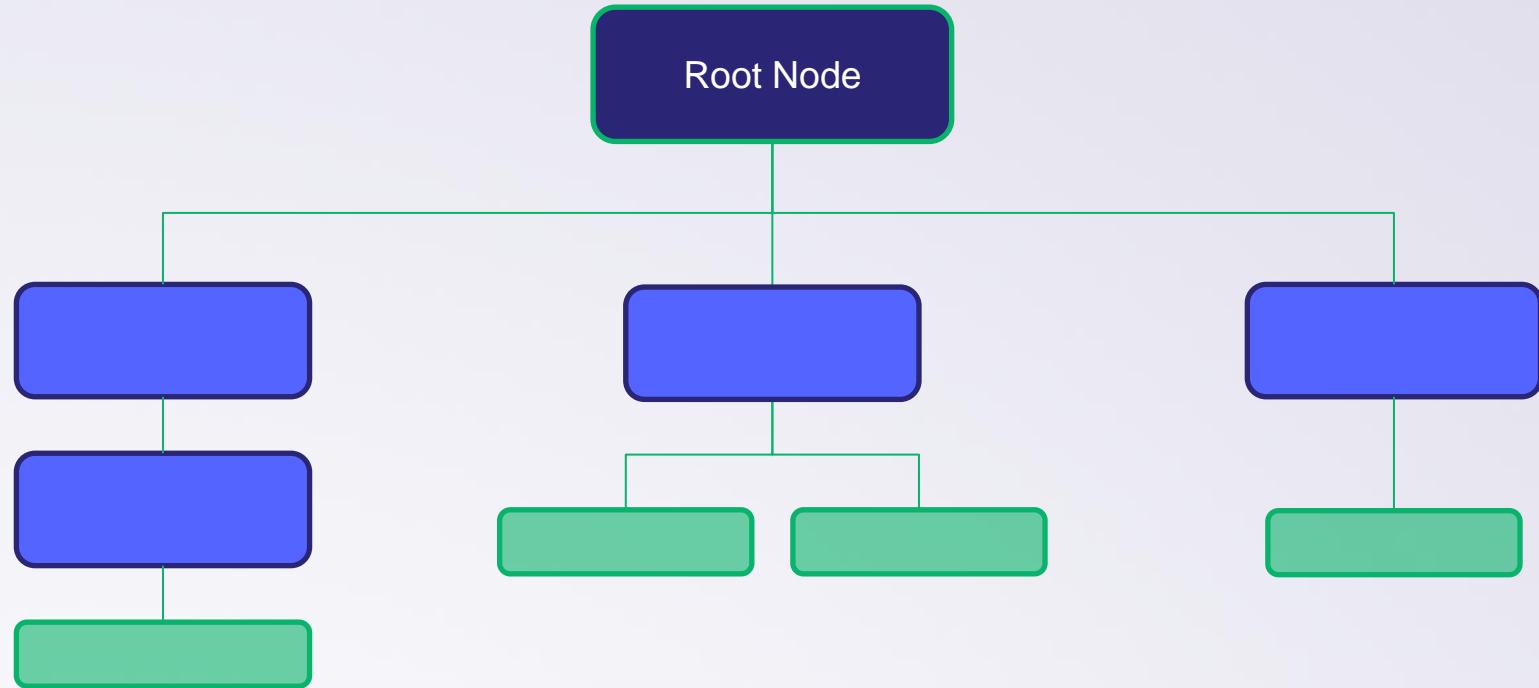
## Hierarchal Graph

A Merkle Tree is hashing technique that can be applied to any hierachal structures and has been traditionally used for data integrity validation.



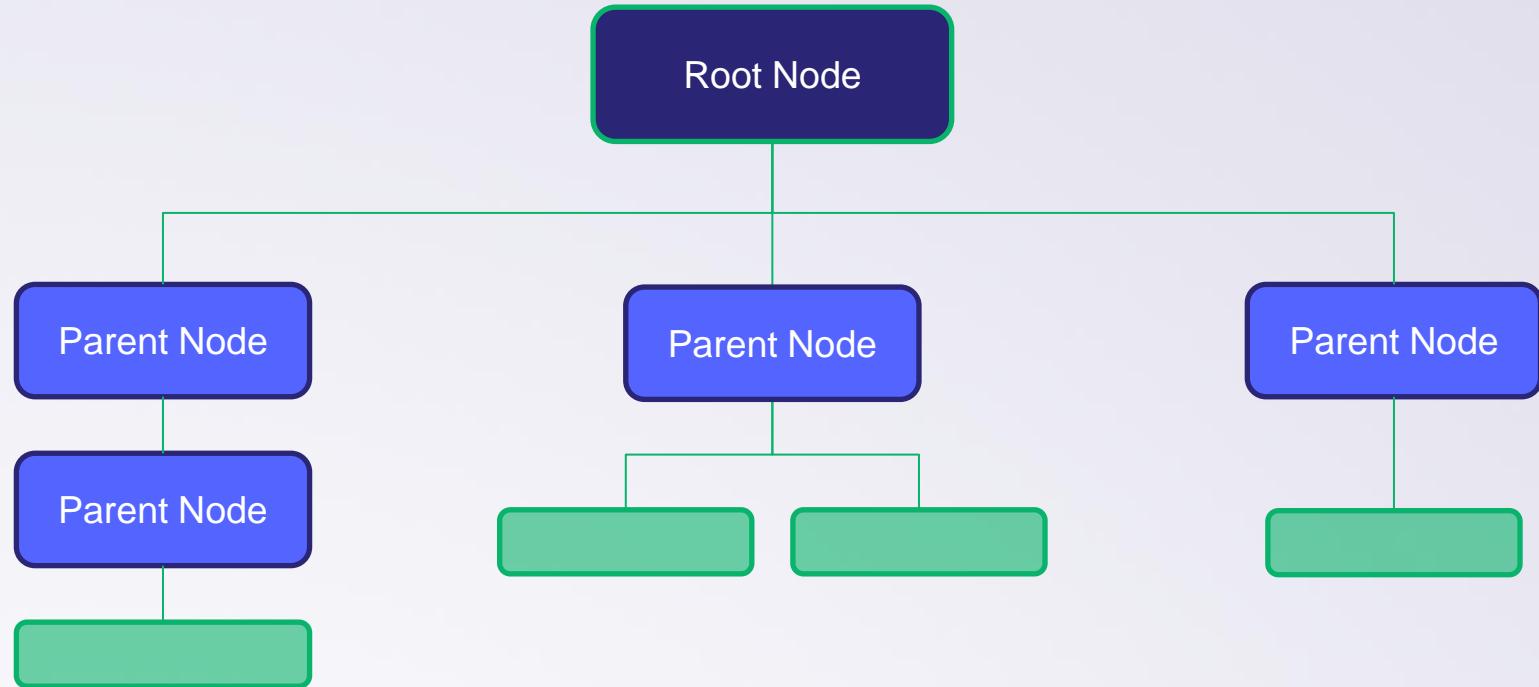
# Merkle Trees

## Root Nodes



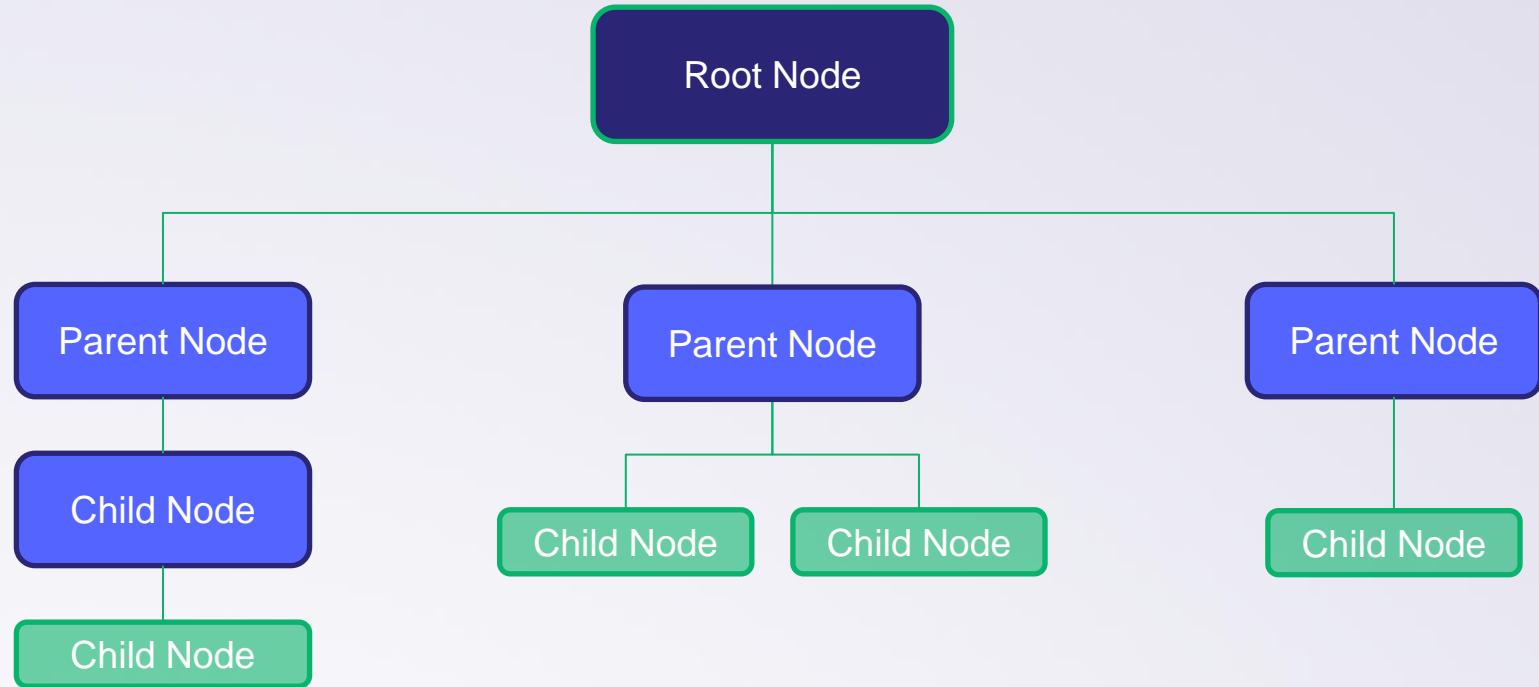
# Merkle Trees

## Parent Nodes



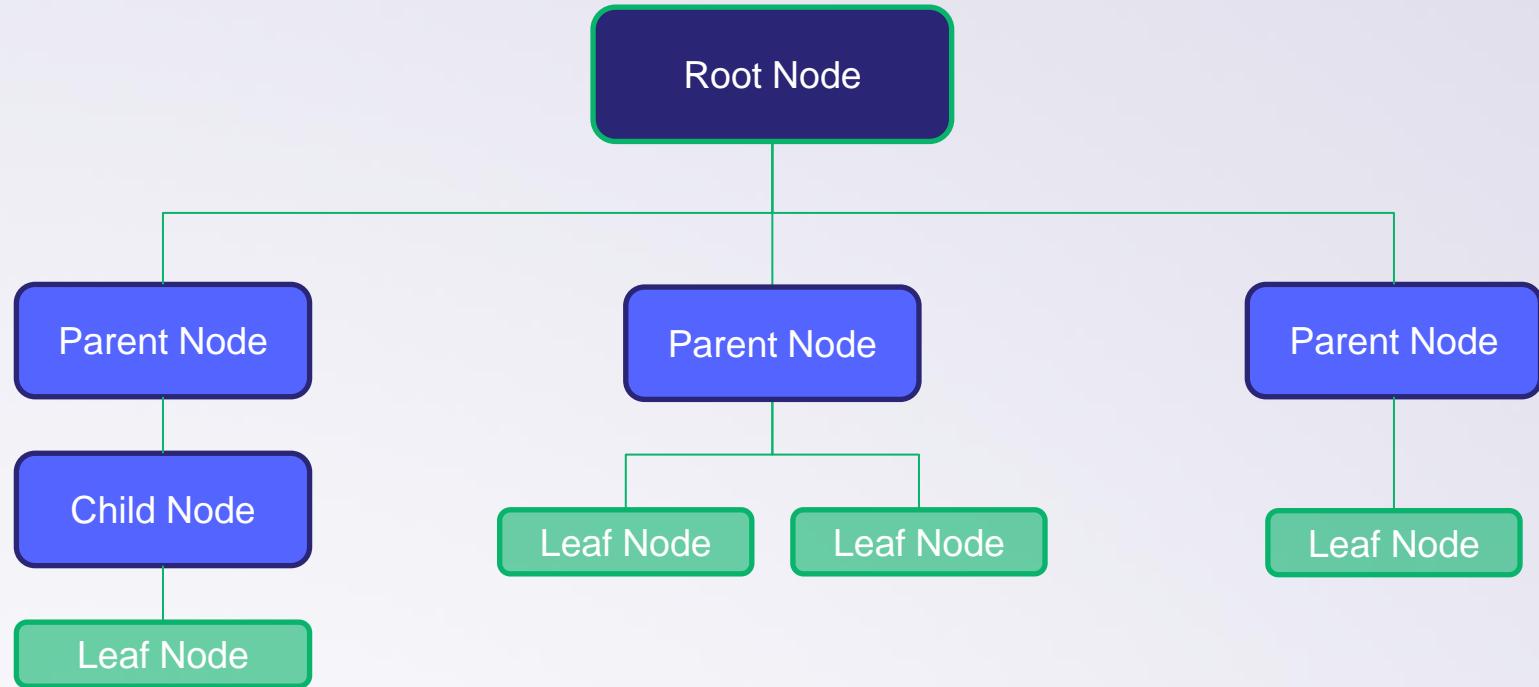
# Merkle Trees

## Child Nodes



# Merkle Trees

## Leaf Nodes



# Merkle Trees

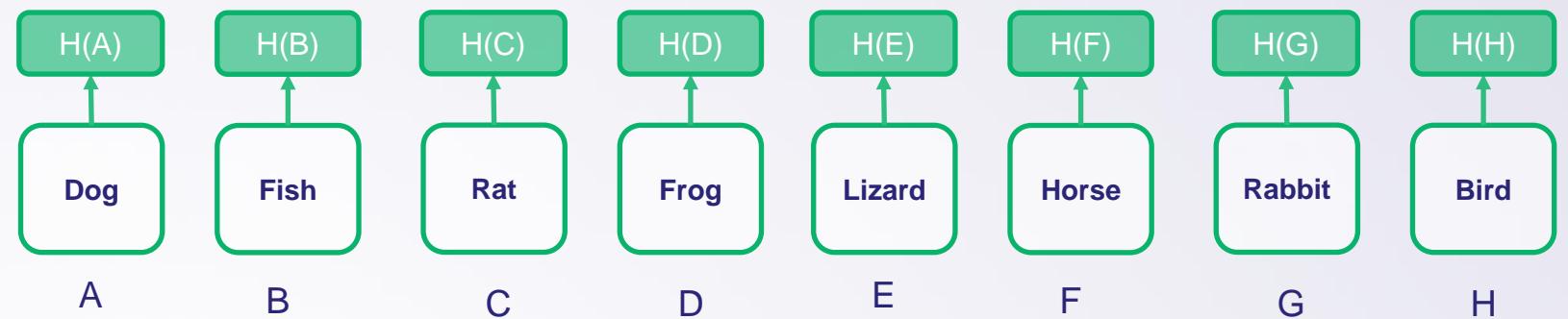
## Hashing Process



# Merkle Trees

## Hashing Process

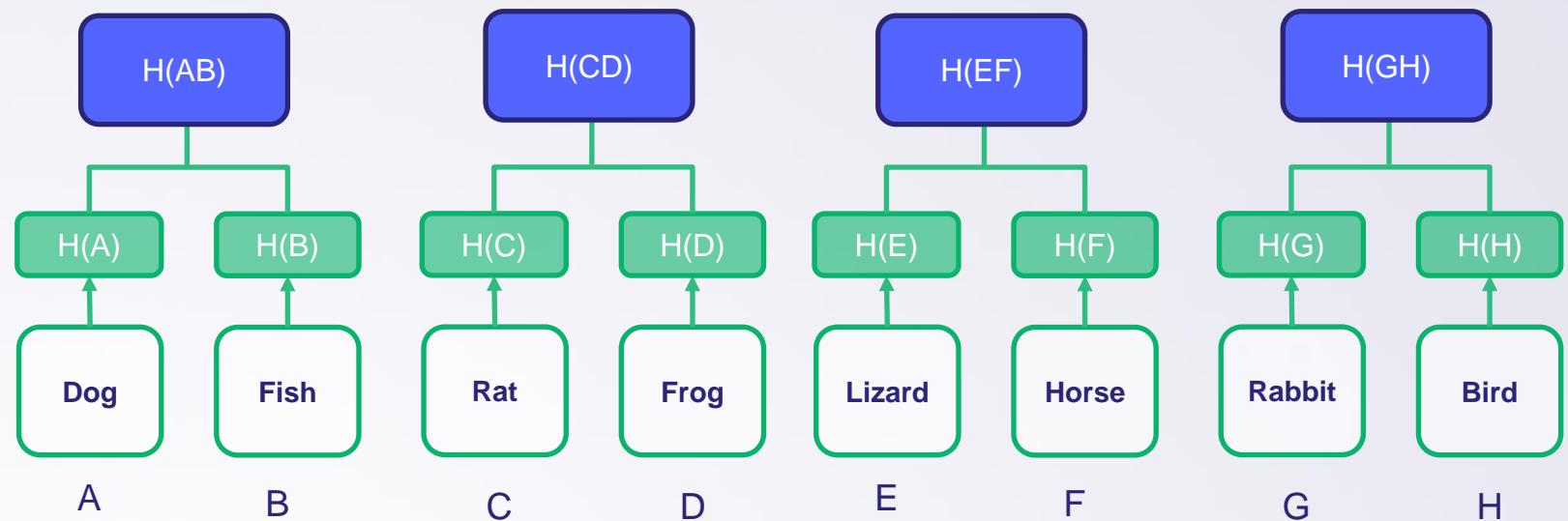
1. Hash the leaf node data.



# Merkle Trees

## Hashing Process

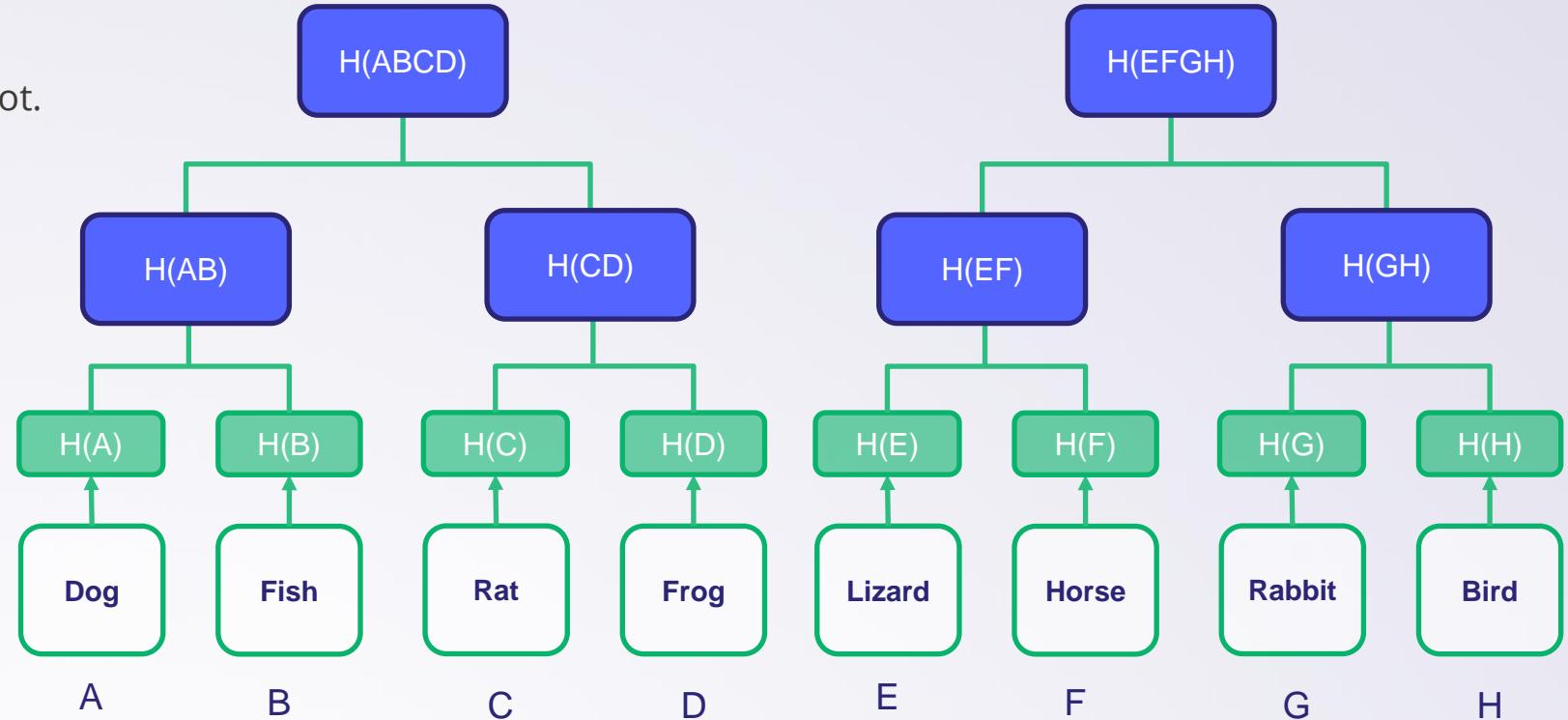
1. Hash the leaf node data.
2. Group the leaf nodes into pairs and hash their hashes.



# Merkle Trees

## Hashing Process

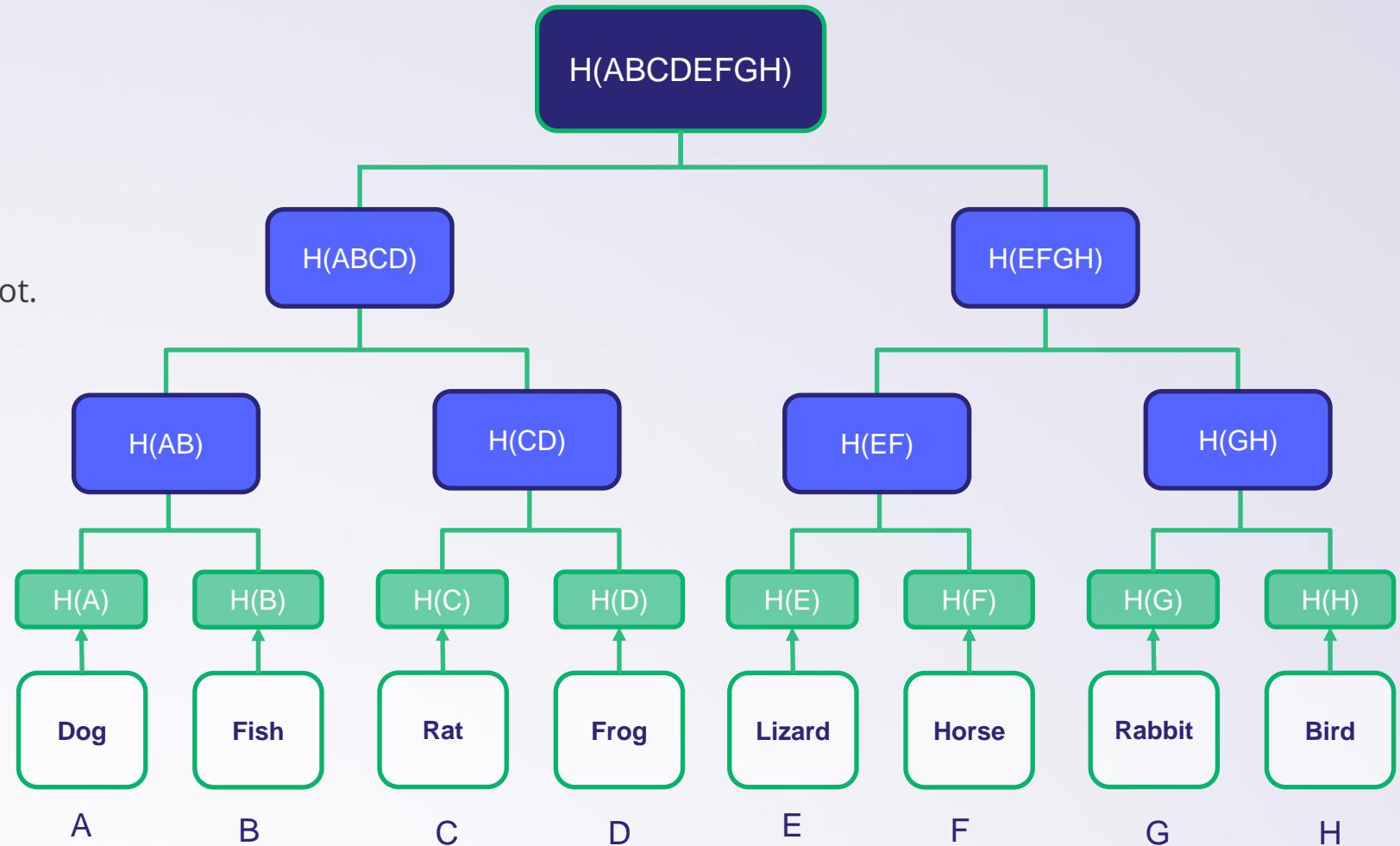
1. Hash the leaf node data.
2. Group the leaf nodes into pairs and hash their hashes.
3. Repeat with the parent nodes until root.



# Merkle Trees

## Hashing Process

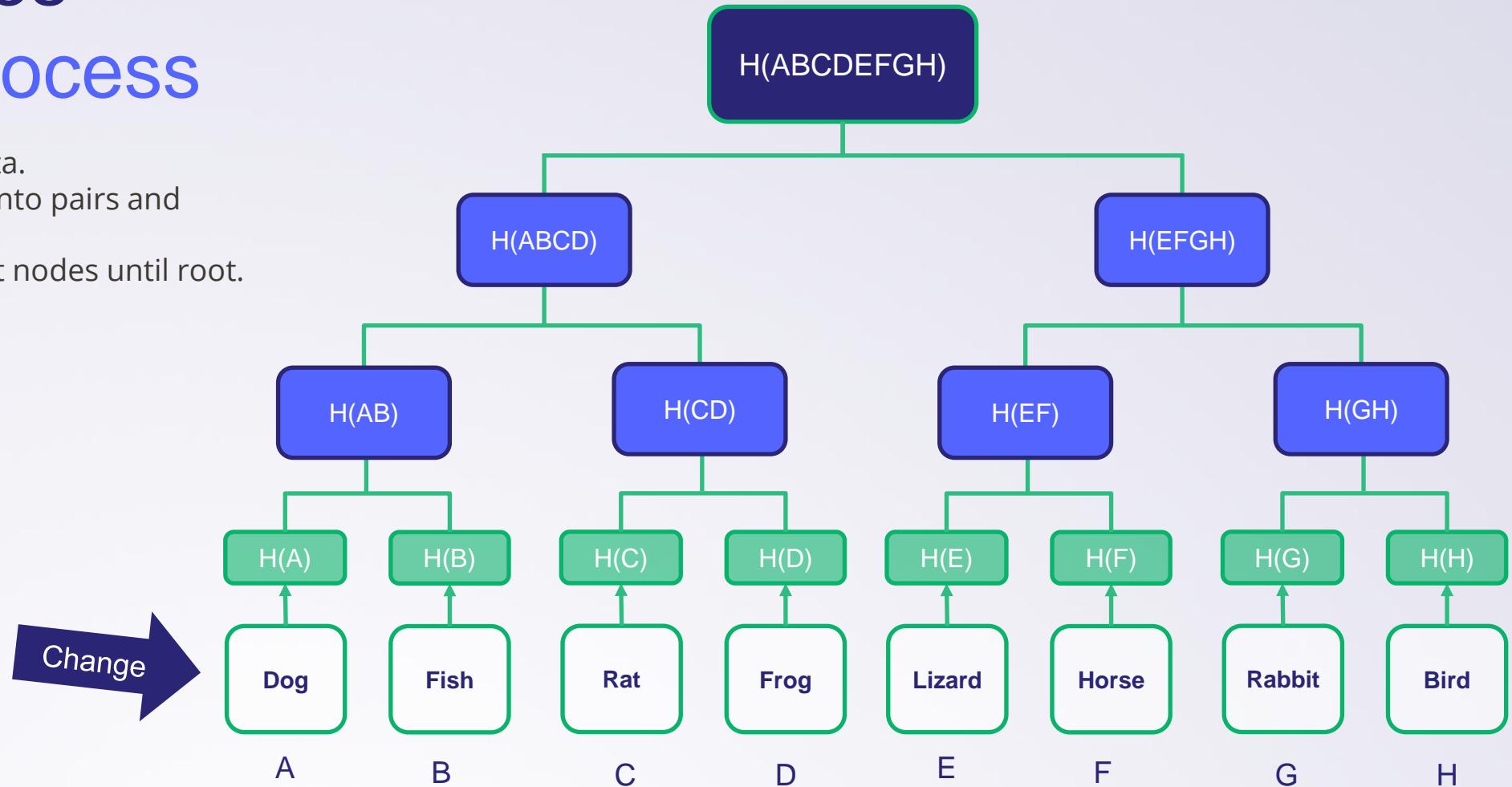
1. Hash the leaf node data.
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# Merkle Trees

## Hashing Process

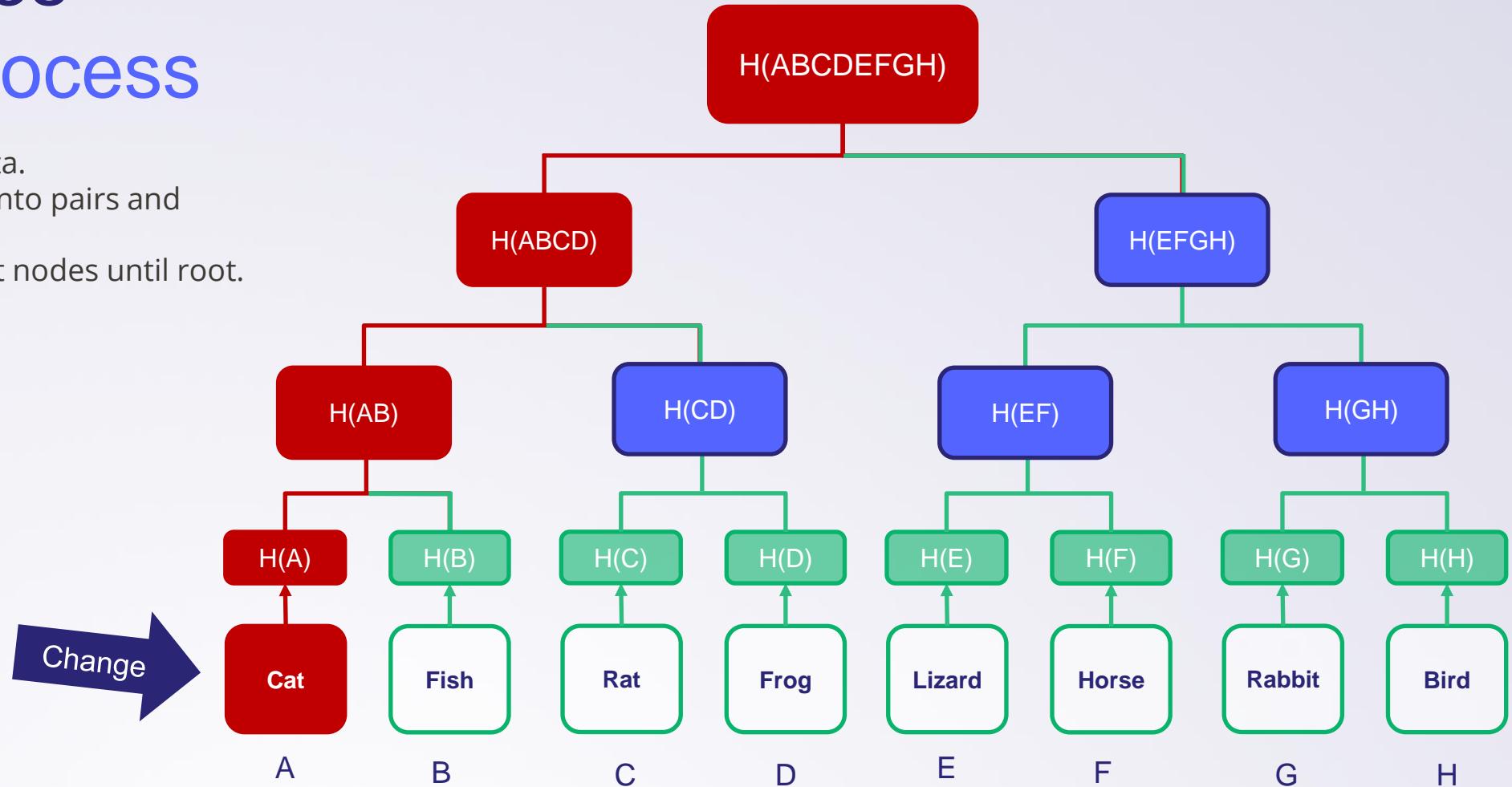
1. Hash the leaf node data.
2. Group the leaf nodes into pairs and hash their hashes.
3. Repeat with the parent nodes until root.



# Merkle Trees

## Hashing Process

1. Hash the leaf node data.
2. Group the leaf nodes into pairs and hash their hashes.
3. Repeat with the parent nodes until root.



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by Folder Group (Dir Hash)
- Group by **Merkle Hash (Nested Dir Hash)**

## Summary

A Merkle Tree is hashing technique that can be applied to any hierachal structures and has been traditionally used for data integrity validation.

## Common Use Cases

- Blockchain
- Certificate Transparency Logs
- P2P File Transfers
- Database indexing

## Share Use Case

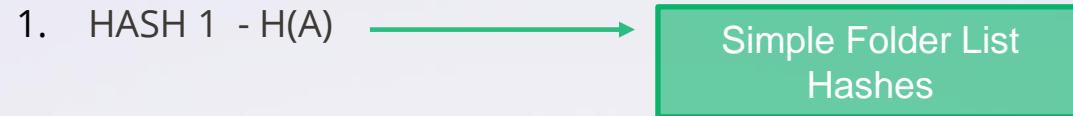
Merkle Trees can also be used to expand on the idea of the "Folder group" by hashing the file listings recursively so you can identify single folder matches as well as **hierarchical folder structure matches**



# Merkle Tree (Modified) Share Use Case



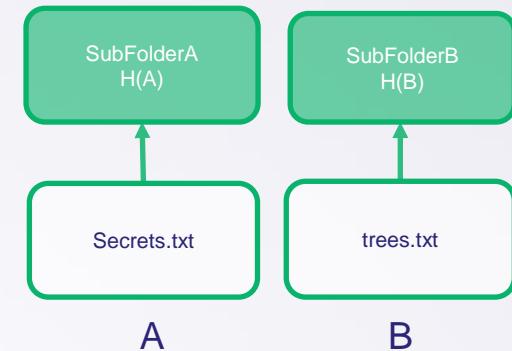
# Merkle Tree (Modified) Share Use Case



# Merkle Tree (Modified)

## Share Use Case

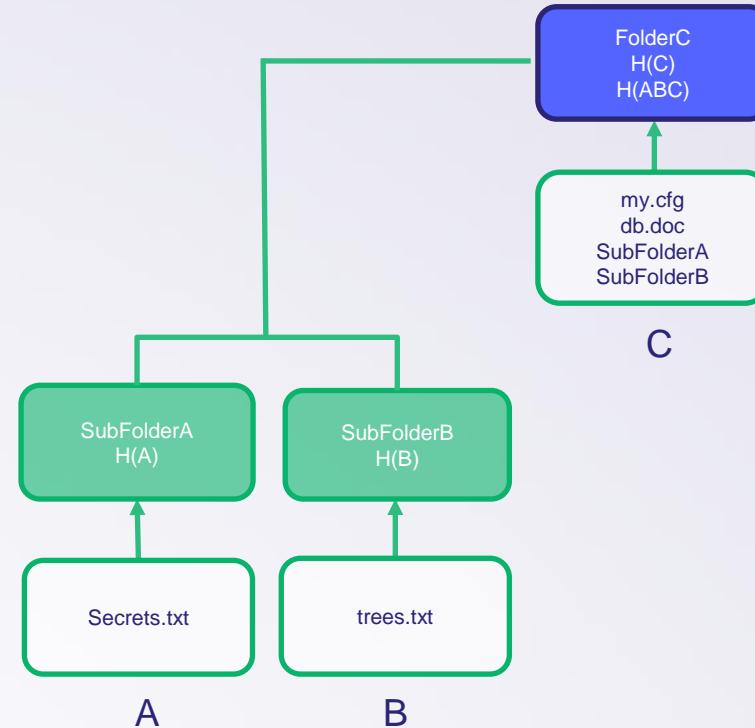
1. HASH 1 - H(A)
  2. HASH 2 - H(B)
- 
- Simple Folder List  
Hashes



# Merkle Tree (Modified)

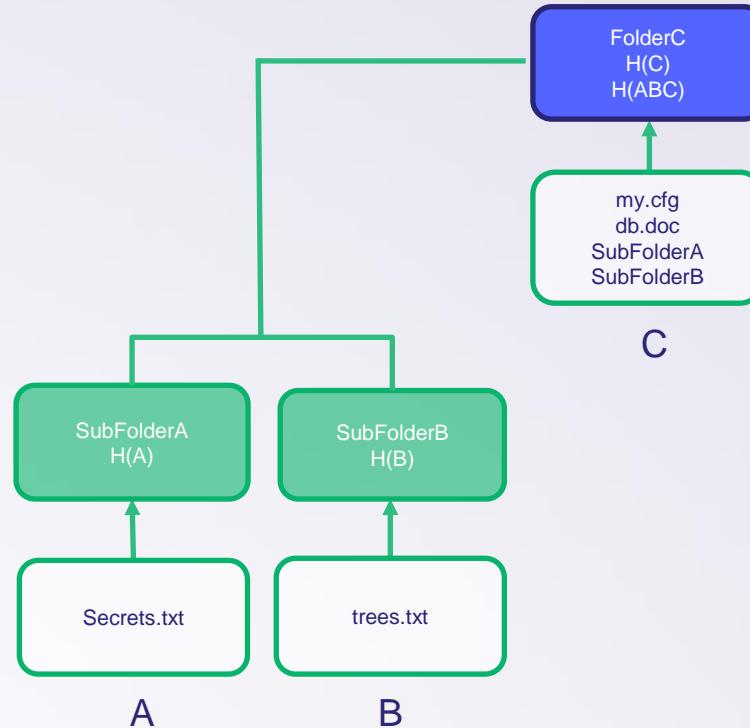
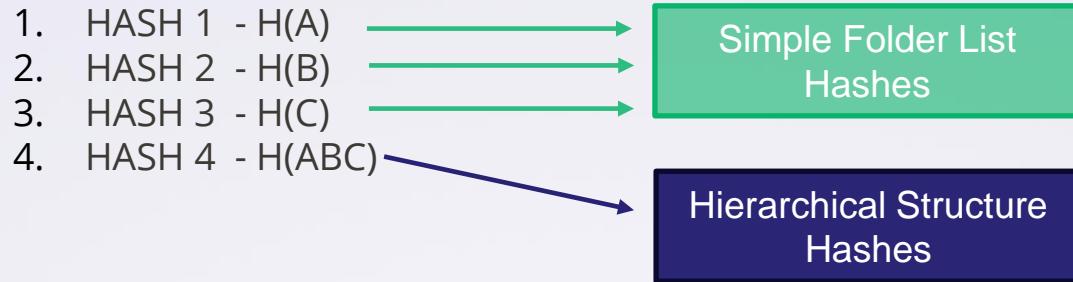
## Share Use Case

1. HASH 1 - H(A)
  2. HASH 2 - H(B)
  3. HASH 3 - H(C)
- Simple Folder List  
Hashes



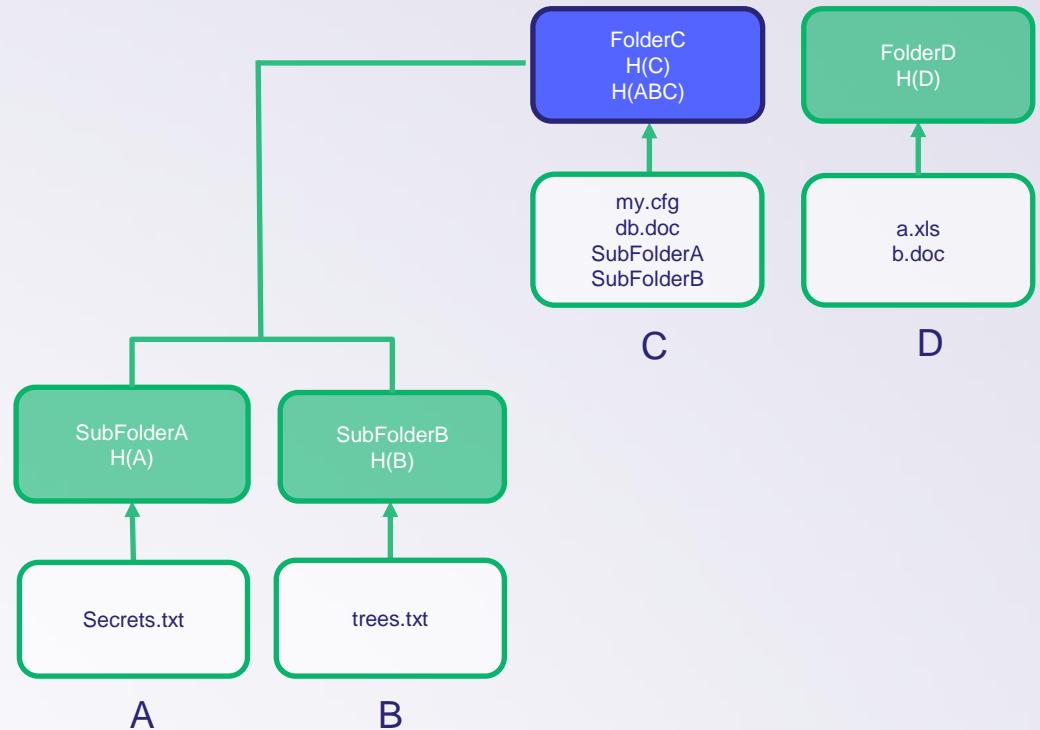
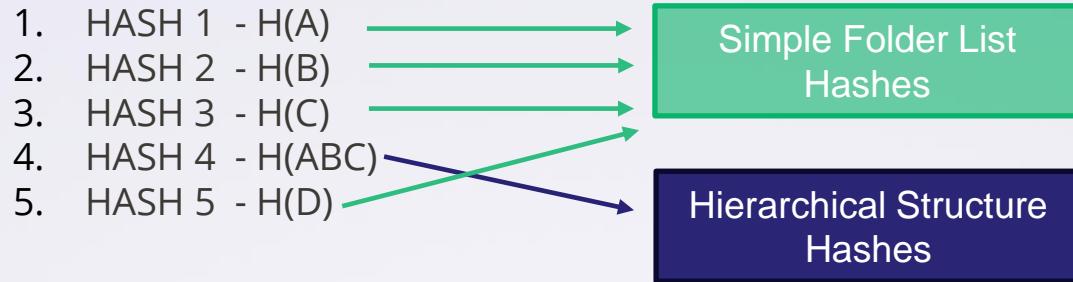
# Merkle Tree (Modified)

## Share Use Case



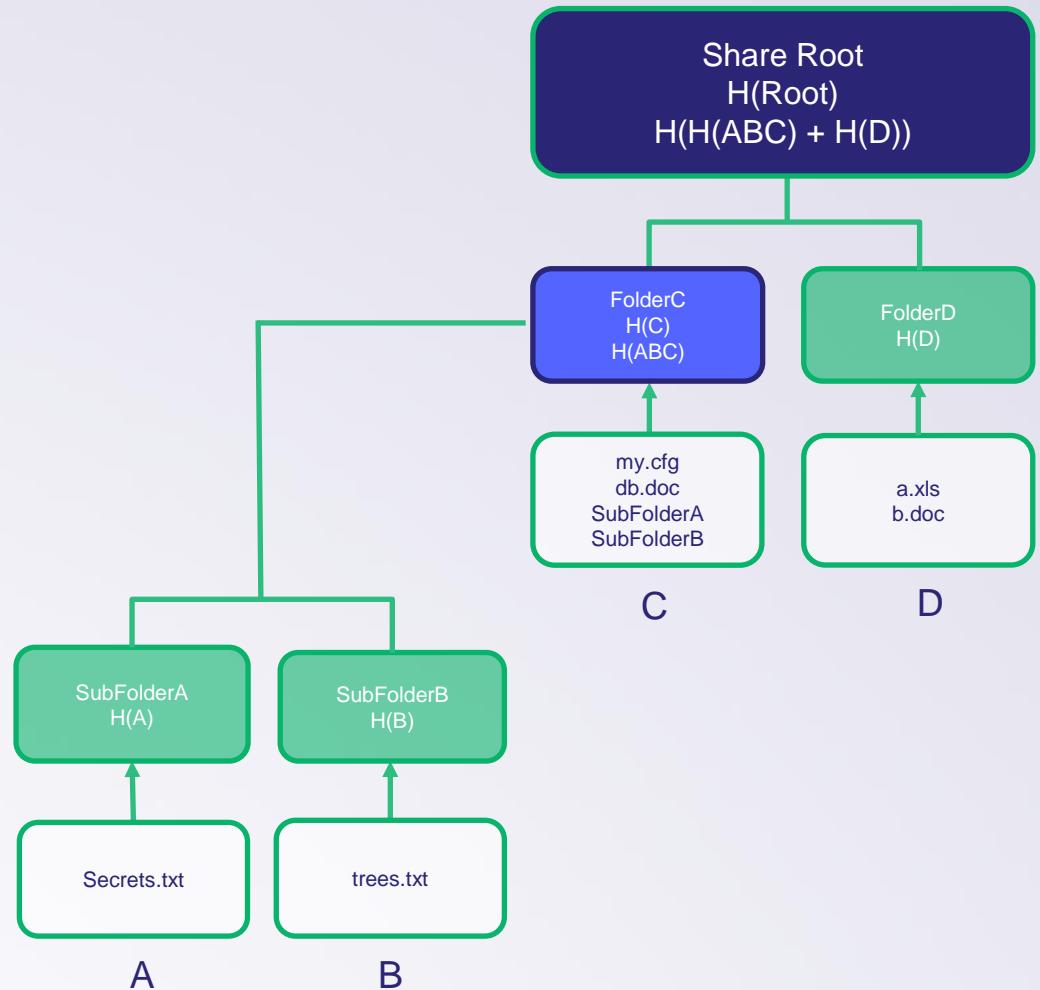
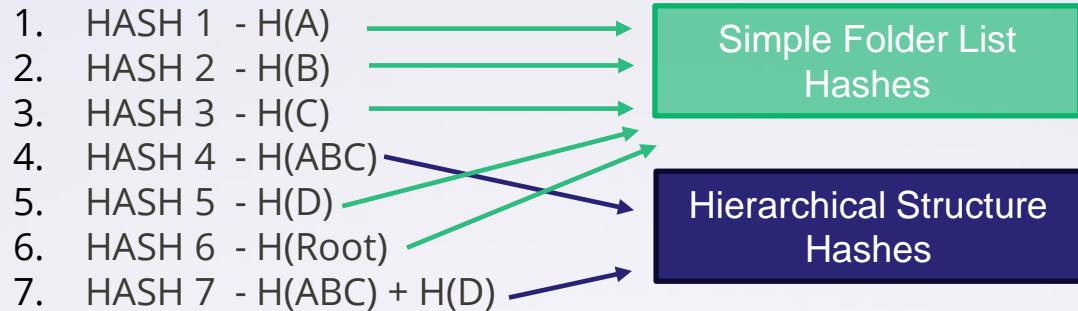
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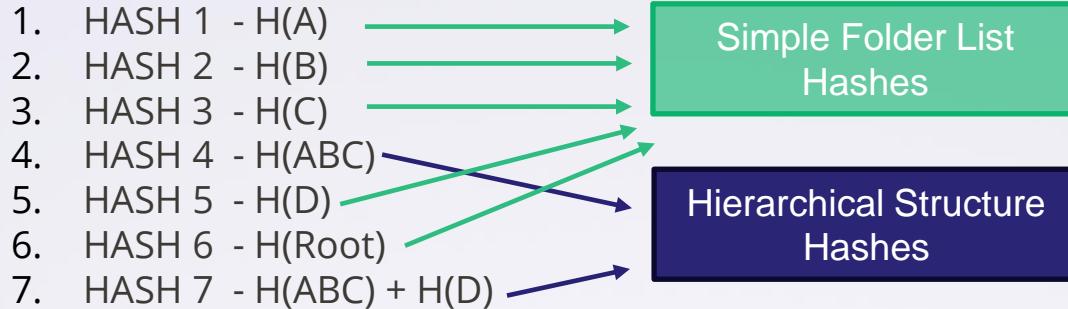
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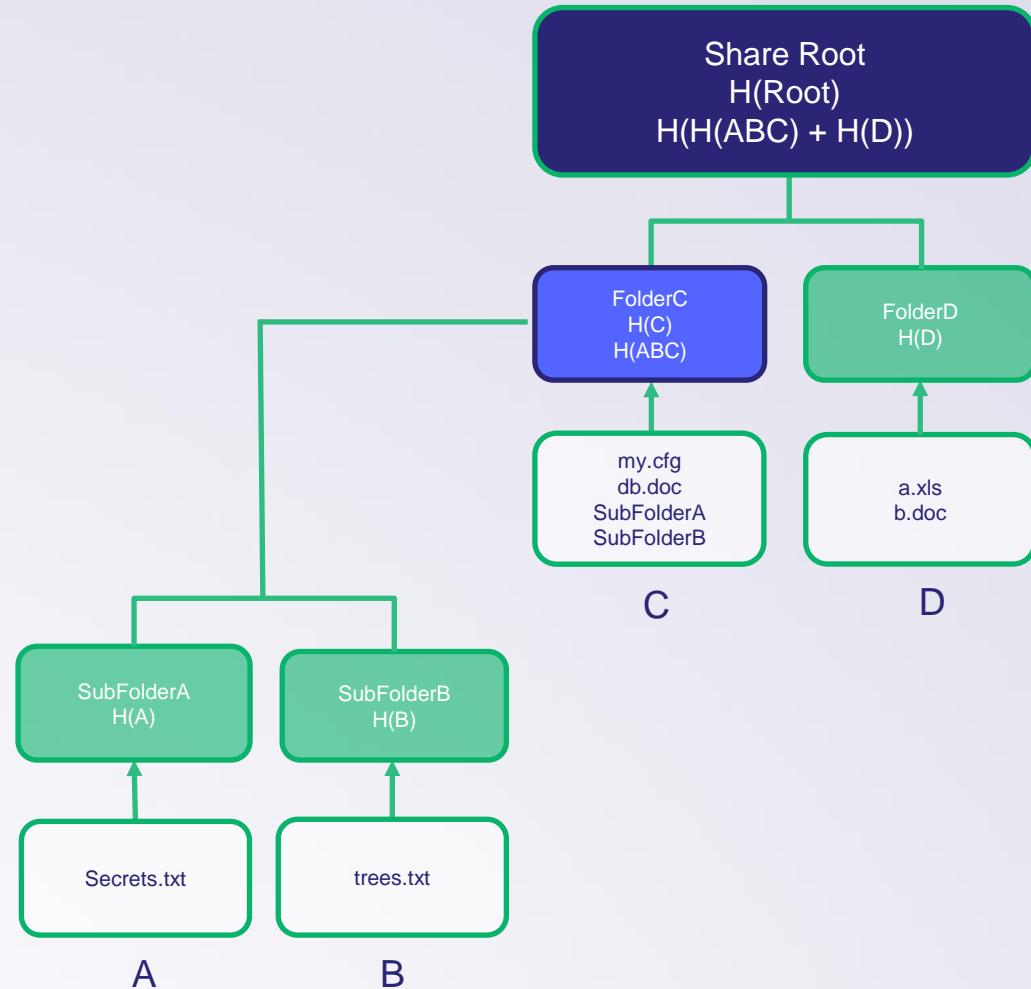


# Merkle Tree (Modified)

## Share Use Case

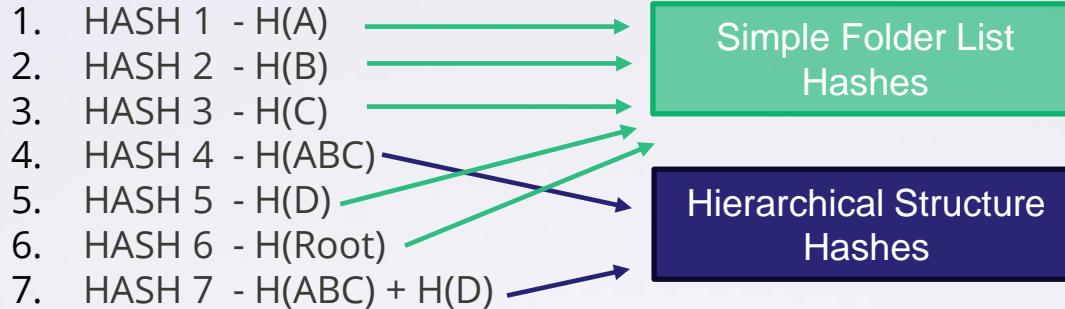


If we store all the hashes in a table, we can then perform simple SQL GROUP BY operations like the “Folder Groups”, but this time we can also see groups of folder hierarchies. ☺

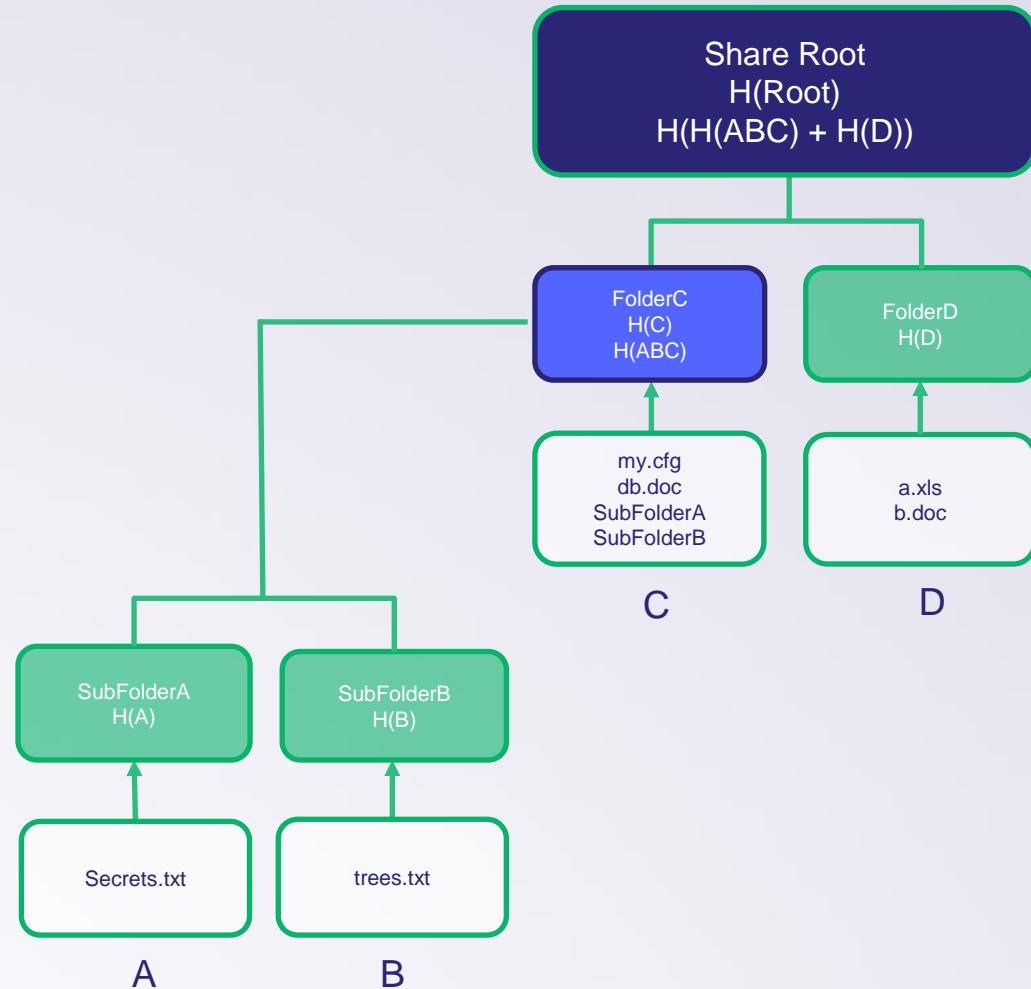


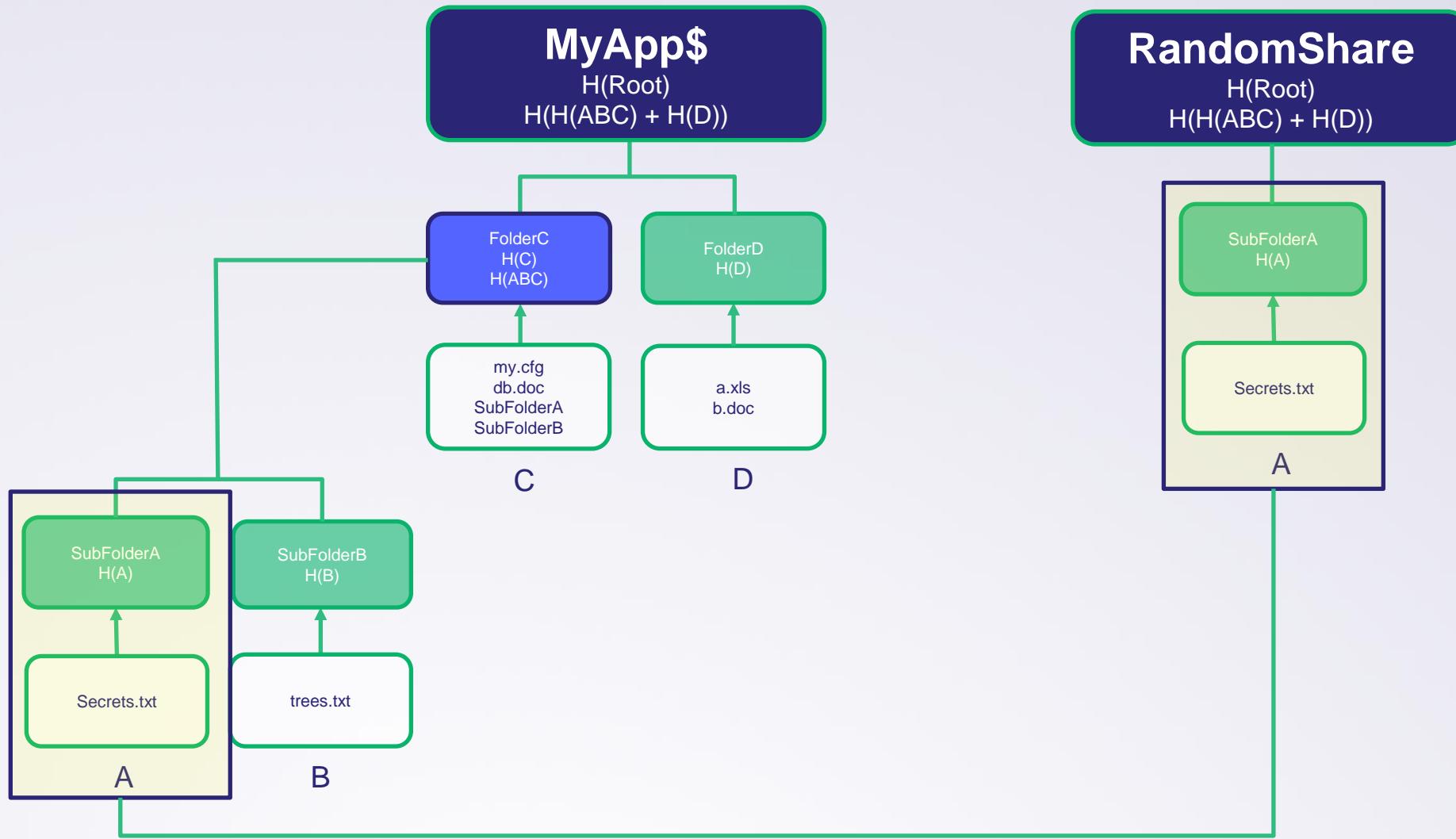
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## Share Use Case



If we store all the hashes in a table, we can then perform simple SQL GROUP BY operations like the “Folder Groups”, but this time we can also see groups of folder hierarchies. ☺





# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by Folder Group (Dir Hash)
- Group by **Merkle Hash (Nested Dir Hash)**

## Our Use Case

Merkle Trees can also be used to expand on the idea of the "Folder group" by hashing the file listings recursively so you can identify nested folder and file listing structure at any folder level.

### Pros

- Can surface relationships between shares.
- Works great for grouping hierarchies with EXACT structural match.
- Can be used **for hunting for threats and vulnerabilities based on folder, registry memory, database, code structures etc.**

### Cons

- Works poorly when the shares DO NOT have the exact same list of files but are used by the same application.  
**Which happens a lot.**
- Collecting recursive direct listings from shares deeper than 3 levels can take a long time in large environments.



# Grouping & Similarity

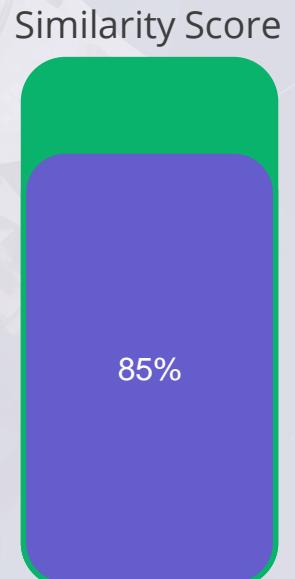
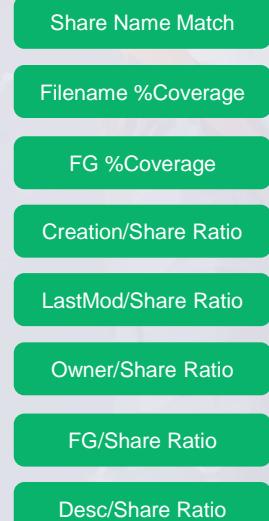
*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by Folder Group (Dir Hash)
- Group by Merkle Hash (Nested Dir Hash)
- **Calculate weighted similarity score**

## Summary

The weighted similarity score used to group shares in PowerHuntShares v2 is derived from multiple data points which are normalized to determine the percentage of similarity.

## Logic Abstract



# Grouping & Similarity

*"How can I group similar shares so I can take targeted actions?"*

- Group by Share Name
- Group by Folder Group (Dir Hash)
- Group by Merkle Hash (Nested Dir Hash)
- **Calculate weighted similarity score**

## Summary

The similarity score in PowerHuntShares v2 is derived from the following meta data:

## Pros

- More accurate than the other methods alone.
- More granular metrics provide more information for root cause analysis. Example: Date & owner differences can tell a story.

## Cons

- Does not take into account fingerprints.
- Does not take into account Merkle Hashes.
- Does not take into account file contents.

Note: The same approach could be applied to almost any file storage medium. For example: AWS s3, Azure blob, or GCP storage.



# Grouping & Similarity

## Summary

**POWERHUNTHASHES** demo.local

14 matches found [Export](#) | [Clear](#)

**RESULTS**

- [Summary Report](#)
- [Scan Information](#)

**EXPLORE**

- [Networks](#)
- [Computers](#)
- [Share Names](#)
- [Folder Groups](#)
- [Insecure ACEs](#)
- [Identities](#)
- [ShareGraph](#)

**TARGET**

- [Interesting Files](#)
- [Extracted Secrets](#)

**ACT**

- [Exploit](#)
- [Detect](#)
- [Remediate](#)

Search

Quick Filters:  Exploitable  Write  Read  Interesting  Empty  Stale  Default

Share Count	Share Name	Risk Level	Share Similarity	Folder Groups	Common Files
2	C\$	24 Critical	84% High	2	6 Files
			Risk Summary		
			HE: 100% (2)		
			Write: 100% (2)		
			Read: 100% (2)		
			Stale: 100% (2)		
			Empty: 0% (0)		
			Default: Yes		
			Sensitive: 0		
			Secrets: 1		
2	ADMIN\$	20 Critical	84% High	2	74 Files
1	backup	3 Low	100% Very High	1	0 Files

**Share Context Guess**  
The C\$ may be associated with the Windows Admin Share. An administrative share for remote management. C\$ is a default administrative share in Windows. C:\Windows\System32 is the expected local path.

**LLM Application Guess**  
Windows Operating System, Microsoft System Center Configuration Manager

[View in ShareGraph](#)

**Affected Assets**

- Computers: 2 of 13 (15.38%)
- Shares: 2 of 21 (9.52%)
- ACLs: 6 of 127 (4.72%)

**Timeline Context**

- First Created: 07/26/2012
- Last Created: 07/26/2012
- Last Mod: 11/06/2024

**Owners (1)**  
NT SERVICE\TrustedInstaller

**RESULTS**[Summary Report](#)[Scan Information](#)**EXPLORE**[Networks](#)[Computers](#)[Share Names](#)[Folder Groups](#)[Insecure ACEs](#)[Identities](#)[ShareGraph](#)**TARGET**[Interesting Files](#)[Extracted Secrets](#)**ACT**[Exploit](#)[Detect](#)[Remediate](#)

## Remediation & Prioritization Recommendations

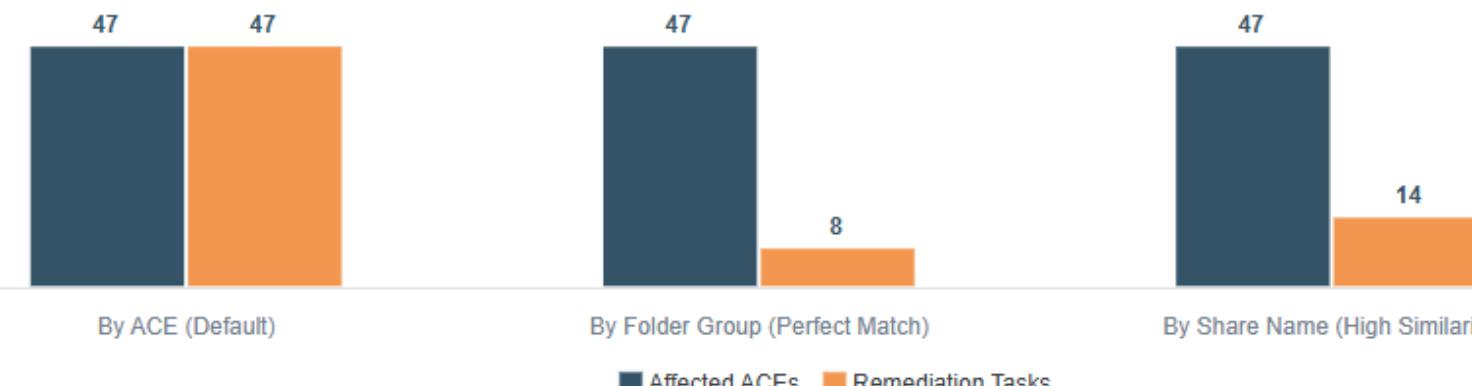
Remediate share ACEs by risk level, starting with critical and high risks. Review the share creation timeline and share name details from other sections for additional context. Consider remediating multiple ACEs at one time based on natural share groupings to reduce the number of remediation tasks.

Group Examples:

- Group ACE remediation tasks by *folder groups*, which contain exactly the same file listing.
- Group ACE remediation tasks by *share names* with a high similarity scores.

Remediating ACEs by group may reduce remediation tasks by as much as **83%** for this environment. The chart below shows the task savings.

Number of Remediation Tasks by Grouping Approach

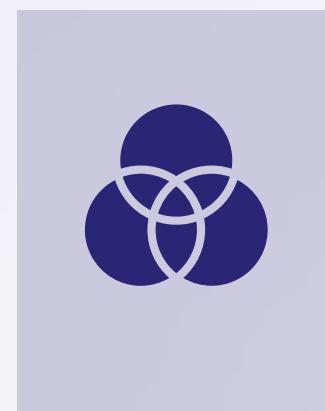
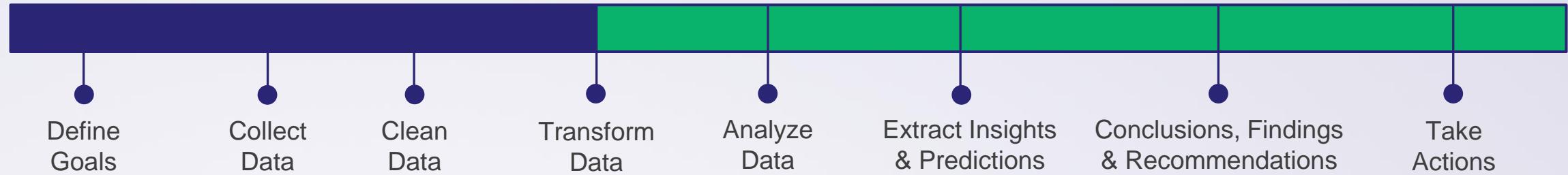


More details are available in the [Folder Group](#), and [Share Names](#) sections.

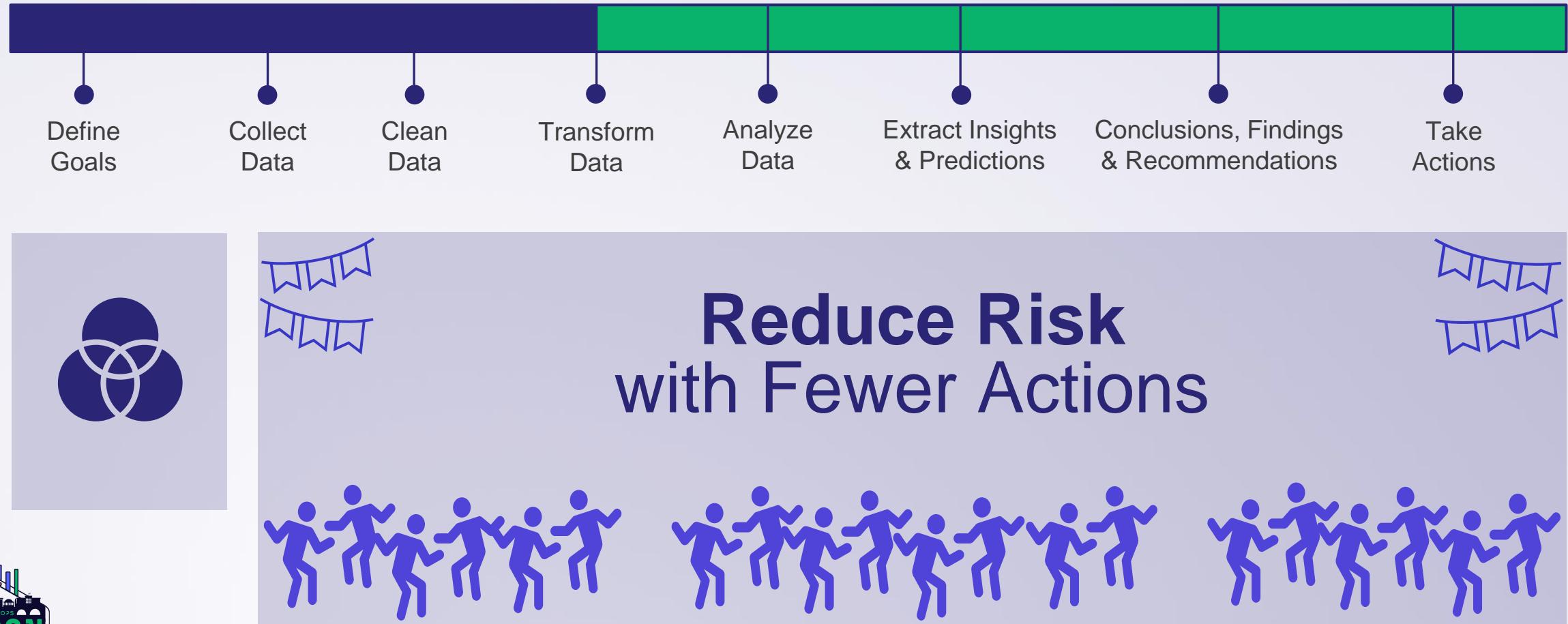
# PowerHuntShares Process



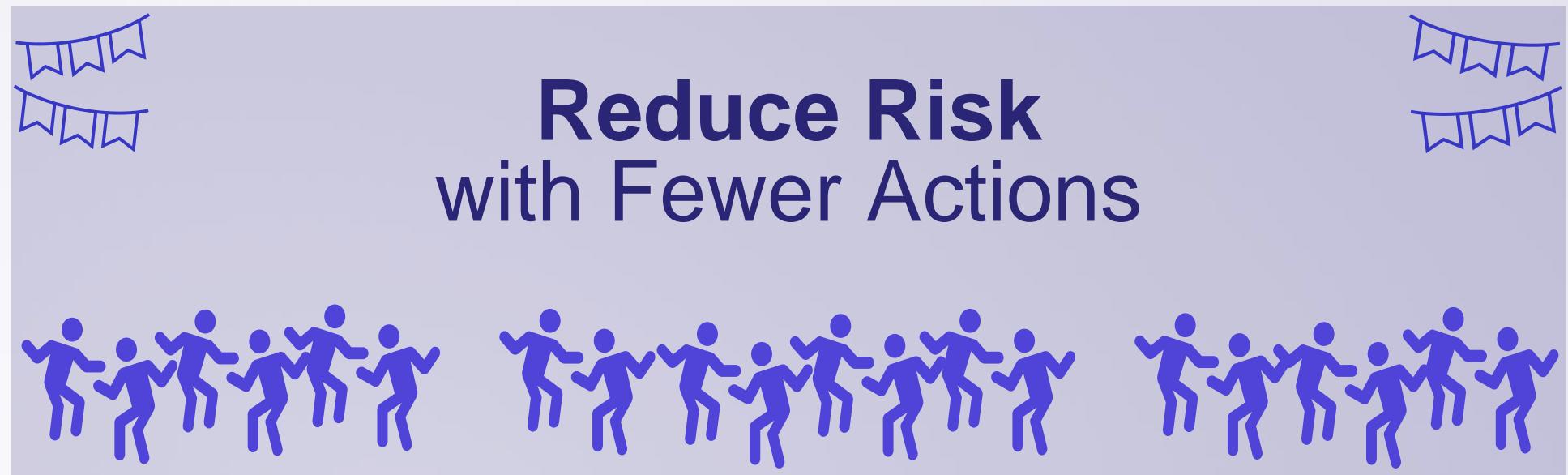
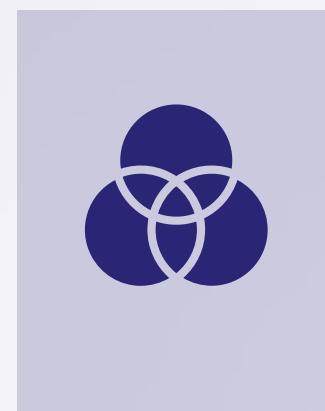
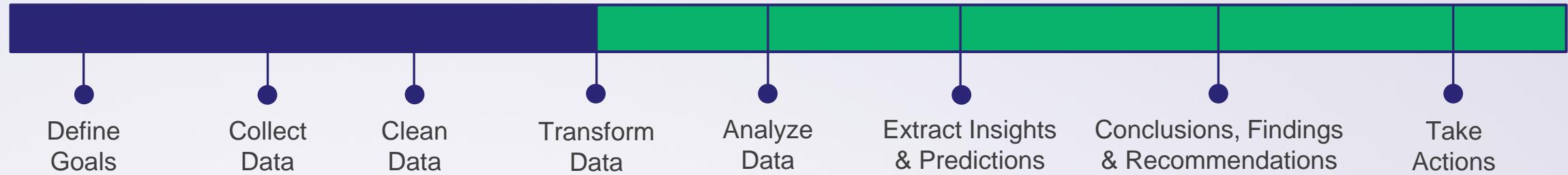
# PowerHuntShares Process



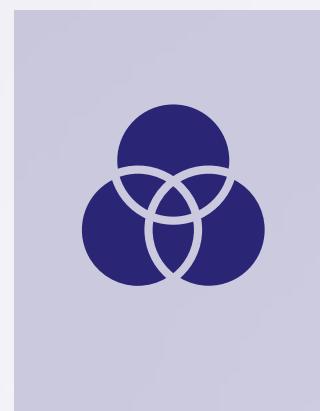
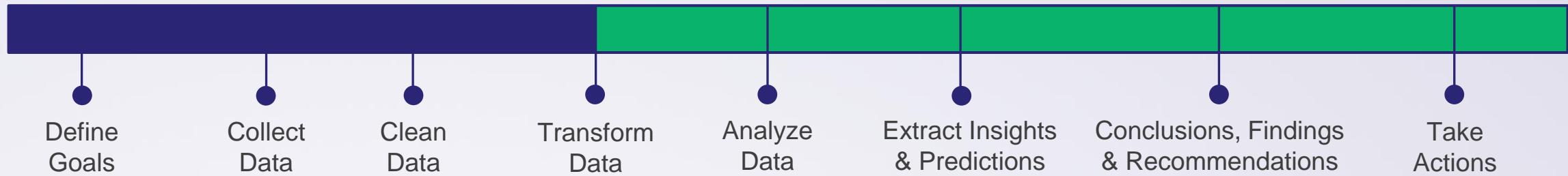
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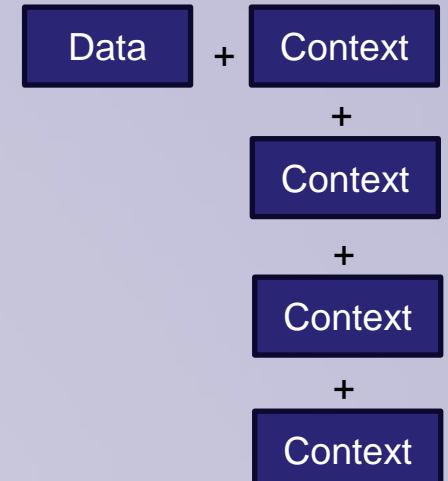
## Transform Data

### Static Data Labeling

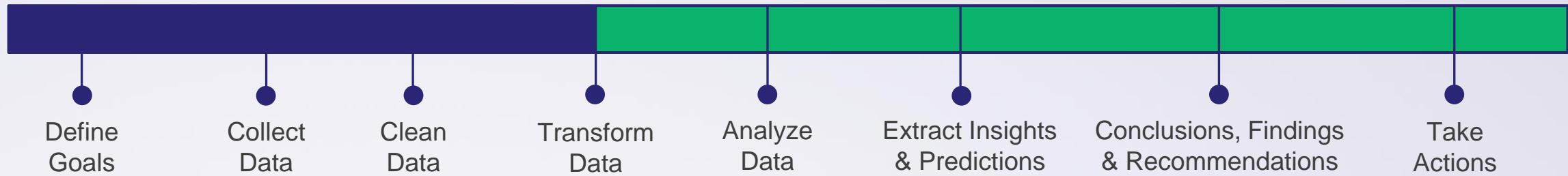
- Highly Exploitable, Interesting Files, Secrets Extraction, Stale, Empty

### Dynamic Data Enrichment

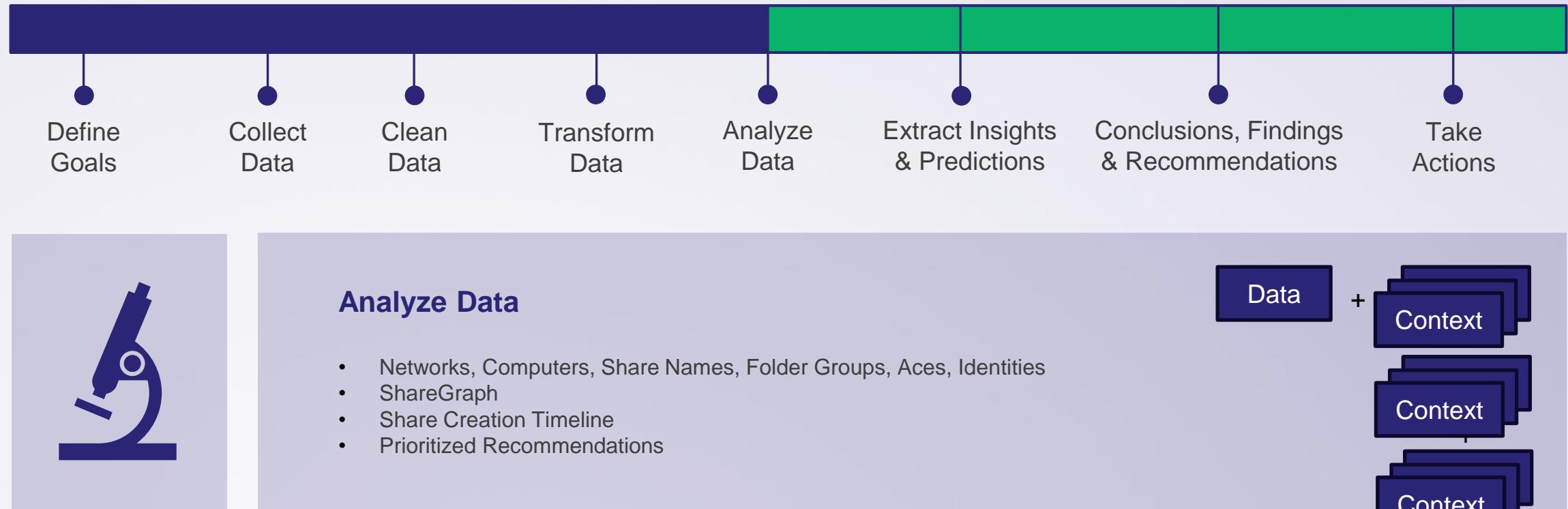
- Fingerprinting, Risk Scoring, Peer Comparison, **Grouping & Similarity Scoring**



# PowerHuntShares Process



# PowerHuntShares Process



# Exploring Data Chart & Graphs

*“How can I explore and visualize my data to gain insights and tell stories?”*



# Exploring Data Chart & Graphs

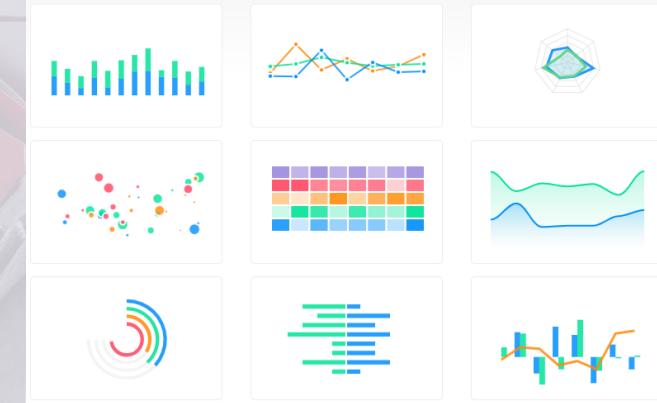
- Simple Charts with **ApexCharts.js**

## ApexCharts.js

*“Can you help me visualize this data in a chart?”*

### Quick Story

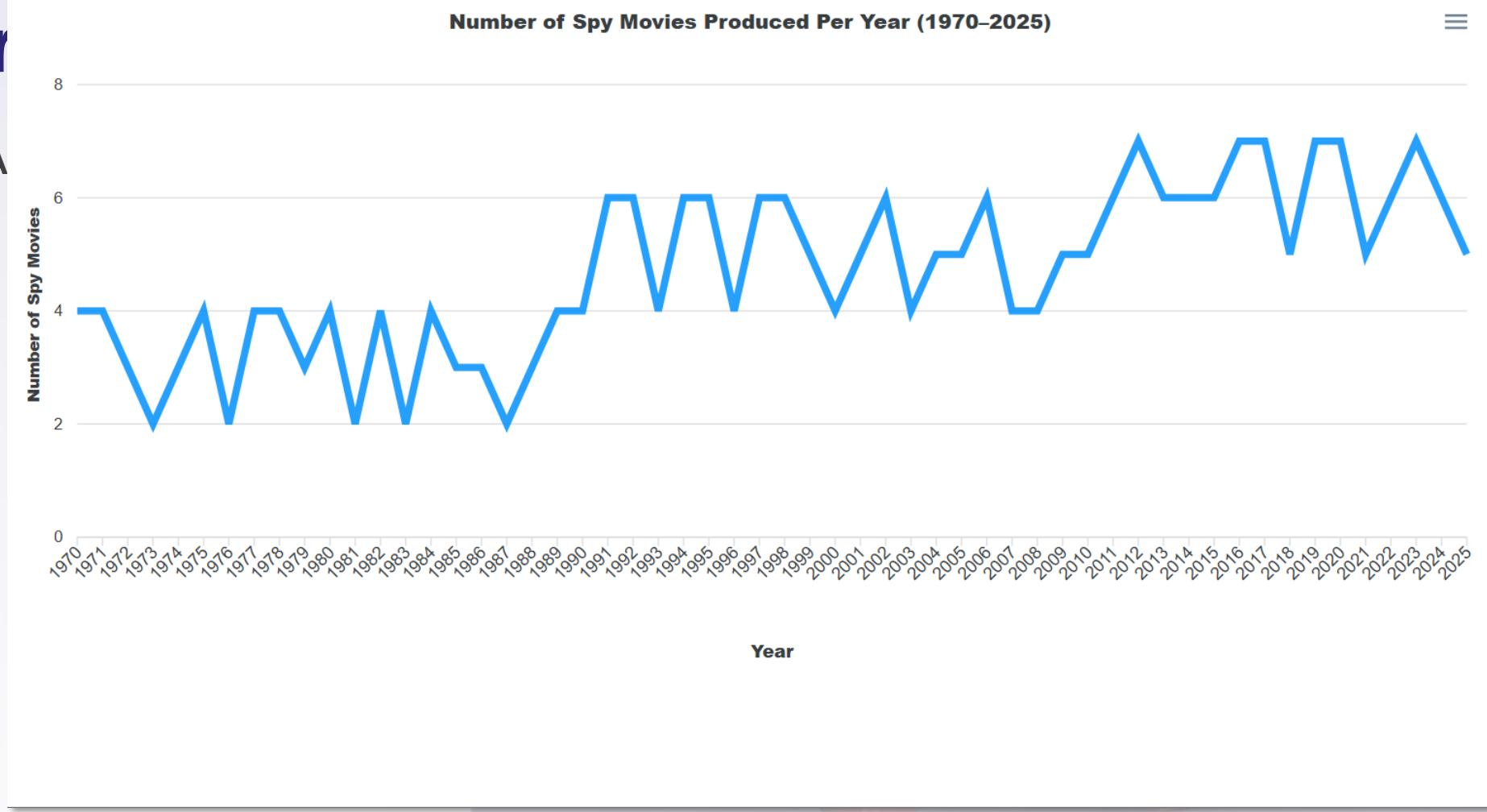
1. Asked ChatGPT for the top 5 open sources/free JavaScript chart libraries with specific features.
2. Provided it a use case and asked it to produce a simple web application with the **ApexCharts.js**.
3. It's been a love affair ever since.



[www.apexcharts.com](http://www.apexcharts.com)

# Exploring Data Chart & Gr

- Simple Charts with A



# Exploring Data

## Share Creation Timeline

Below is a time series chart to help you understand when shares were created in this environment. This environment was configured with critical risk ACEs. The chart shows the cumulative number of critical and high risk ACEs created over time, along with everything else.

## Finding Exposure Summary

### Critical

**13**

findings

### High

**6**

findings

### Medium

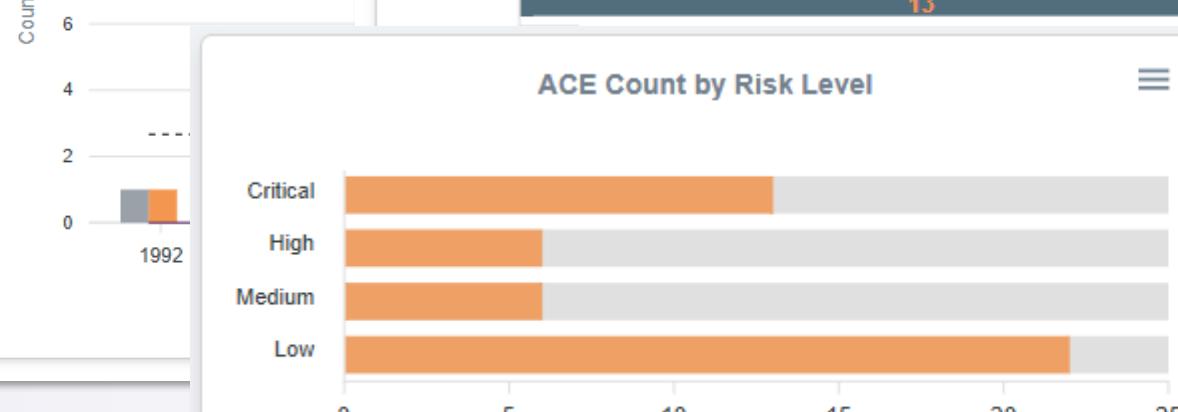
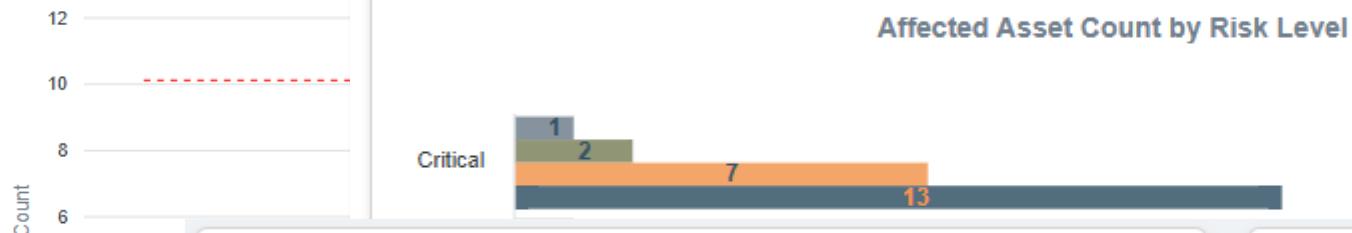
**6**

findings

### Low

**22**

findings



## Data Exposure Summary

### Interesting

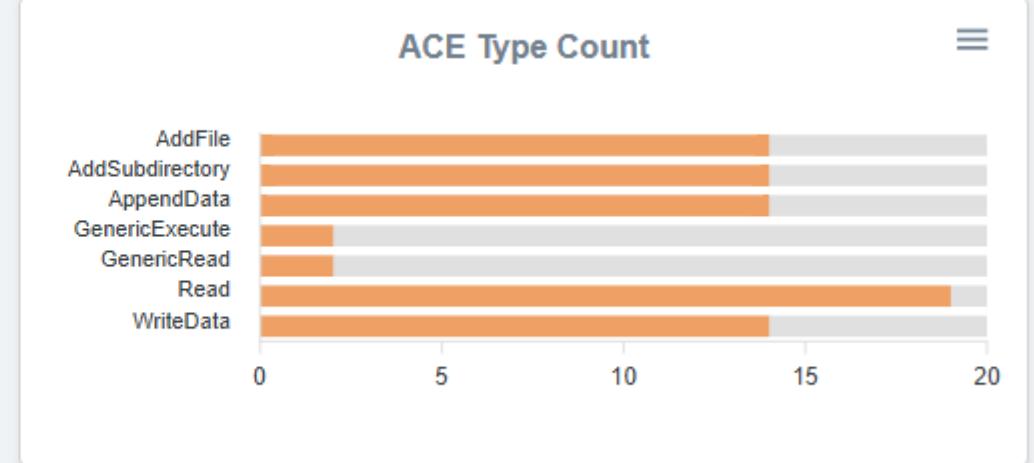
### Sensitive

### Secrets

### Extracted

**143**

secrets (50 files)



## RESULTS

[Summary Report](#)[Scan Information](#)

## EXPLORE

[Networks](#)[Computers](#)[Share Names](#)[Folder Groups](#)[Insecure ACEs](#)[Identities](#)[ShareGraph](#)

## TARGET

[Interesting Files](#)[Extracted Secrets](#)

## ACT

[Exploit](#)[Detect](#)[Remediate](#)

## Summary Report

Testing was conducted between 11/07/2024 08:08:31 and 11/07/2024 08:10:31 to identify network shares configured with excessive privileges hosted on computers joined to the demo.local domain. In total, 13 critical, 6 high, 6 medium, and 22 low risk **ACE (Access Control Entry)** configurations were discovered across 16 shares, hosted by 2 computers in the demo.local domain. Overall, 83 interesting files were found accessible to all domain users that could potentially lead to unauthorized data access or remote code execution. The affected shares were found hosting 53 files that may contain passwords and 0 files that may contain sensitive data. 143 credentials were recovered from 50 of the discovered 53 secrets files.

The section provides a summary of the affected assets, findings, data exposure, share creation timelines, peer comparison and general recommendations.

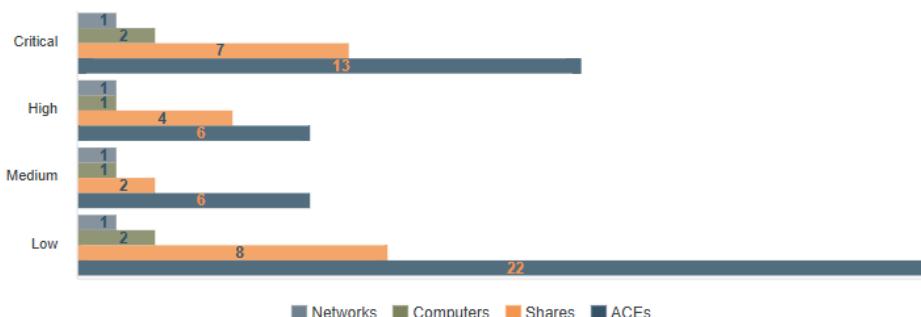
## Finding Exposure Summary

Critical	13	findings
High	6	findings
Medium	6	findings
Low	22	findings

## Data Exposure Summary

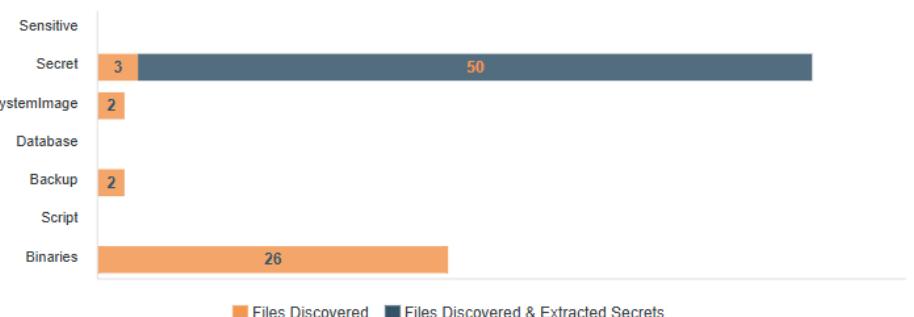
Interesting	83	files found
Sensitive	0	files found
Secrets	53	files found
Extracted	143	secrets (50 files)

Affected Asset Count by Risk Level



More details available in the [Networks](#), [Computers](#), [Shares](#), and [ACEs](#) sections.

Interesting File Exposure



More details are available in the [Extracted Secrets](#), and [Interesting Files](#) sections.

## Asset Exposure Summary

47 ACL entries, on 16 shares, hosted by 2 computers were found configured with excessive privileges on the demo.local domain. In this environment, we observed a total of 19 application instances, with 4 unique

## Affected Asset Peer Comparison

Below is a comparison between the percent of affected assets in this environment and the average percent of affected assets observed in other environments. The percentage is calculated based on the total number of live

# Exploring Data Chart & Graphs

- Simple Charts with ApexCharts.js
- Exploring Data with Graphs: Cytoscape.js

## CytoScape.js

*"Can you help me visualize these share relationships?"*

### Similar Story

1. Asked ChatGPT for the top 5 open sources/free JavaScript graphing libraries with specific features
2. Provided it a use case and asked it to produce a simple web application with the graph using **Cytoscape.js**.
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[js.cytoscape.org](http://js.cytoscape.org)



# Exploring Data Chart & Graphs

- Simple Charts with ApexCharts.js
- Exploring Data with Graphs: Cytoscape.js

## CytoScape.js

*"Can you help me visualize these share relationships?"*

### Native Features

- Generate Graph
- Modify Graph Nodes & Layout
- Search & Filter Graph
- Algorithm support for things like shortest Path
- Easy to customize styles
- Easy to wrap code around



# Exploring Data Chart & Graphs

- Simple Charts with ApexCharts.js
- Exploring Data with Graphs: Cytoscape.js



## CytoScape.js Prompt Example

Please create an html graph using Cytoscape.js that includes:

### Layout Options

1. Add buttons to change the layout to breadthfirst and the top five other layouts like grid.
2. Add buttons to show Pageranked most influential nodes in bright orange.
3. When Pageranked button is clicked resize nodes based on pagerank.
3. Add buttons to show Betweenness Centrality nodes in bright tan and create a px border in black.

### Nodes with the details below:

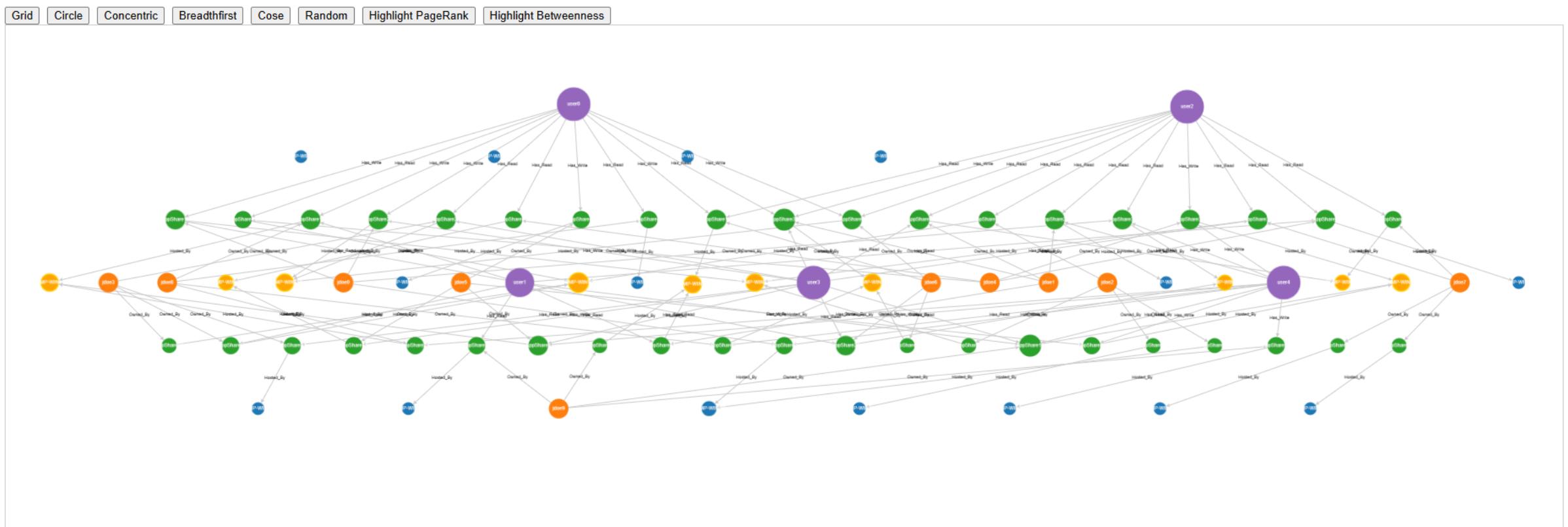
1. Four node types: ComputerName, ShareName, Owner, and User nodes.
2. Generate a list of 10 Owner nodes that look like user names.
3. Generate a list of 25 ComputerName nodes that look like they would be part of a common enterprise network.
4. Generate a list of 40 ShareName nodes that look like SMB shares used by applications
5. Generate a list of 5 UserName nodes that see like simple user names.
6. Ensure all nodes are large enough to be read.
7. Ensure all nodes are the same shape.
8. Ensure all nodes types have a different color.

### Edges with the details below:

1. Four edge types: Owned\_By, Hosted\_By, Has\_Write, and Has\_Read.
2. Generate Owned\_By edges between Owner nodes and ShareName nodes. Ensure each ShareName has one owner.
3. Generate Hosted\_By edges between ShareName nodes and ComputerName nodes. Assign those Hosted\_By edges randomly, but ensure at least 80% of ComputerName nodes have at least one ShareName node connected.
4. Generate 20 Has\_Write edge between randomly selected UserName nodes and ShareNames. Do not allow the same UserName to link to a ShareName node more than once using a Has\_Write edge.
5. Generate 30 Has\_Read edge between randomly selected UserName nodes and ShareNames. Do not allow the same UserName to link to a ShareName node more than once using a Has\_Read edge.
6. Ensure all nodes are large enough to be read.
7. Ensure all nodes are the same shape.

Please dont forget to add the nodes and edges.

# Exploring Data Chart & Graphs



5. Generate 30 Has\_Read edge between randomly selected UserName nodes and ShareNames. Do not allow the same UserName to link to a ShareName node more than once using a Has\_Read edge.
6. Ensure all nodes are large enough to be read.
7. Ensure all nodes are the same shape.

Please dont forget to add the nodes and edges.



## RESULTS

- Summary Report
- Scan Information

## EXPLORE

- Networks
- Computers
- Share Names
- Folder Groups
- Insecure ACEs
- Identities

## ShareGraph

## TARGET

- Interesting Files
- Extracted Secrets

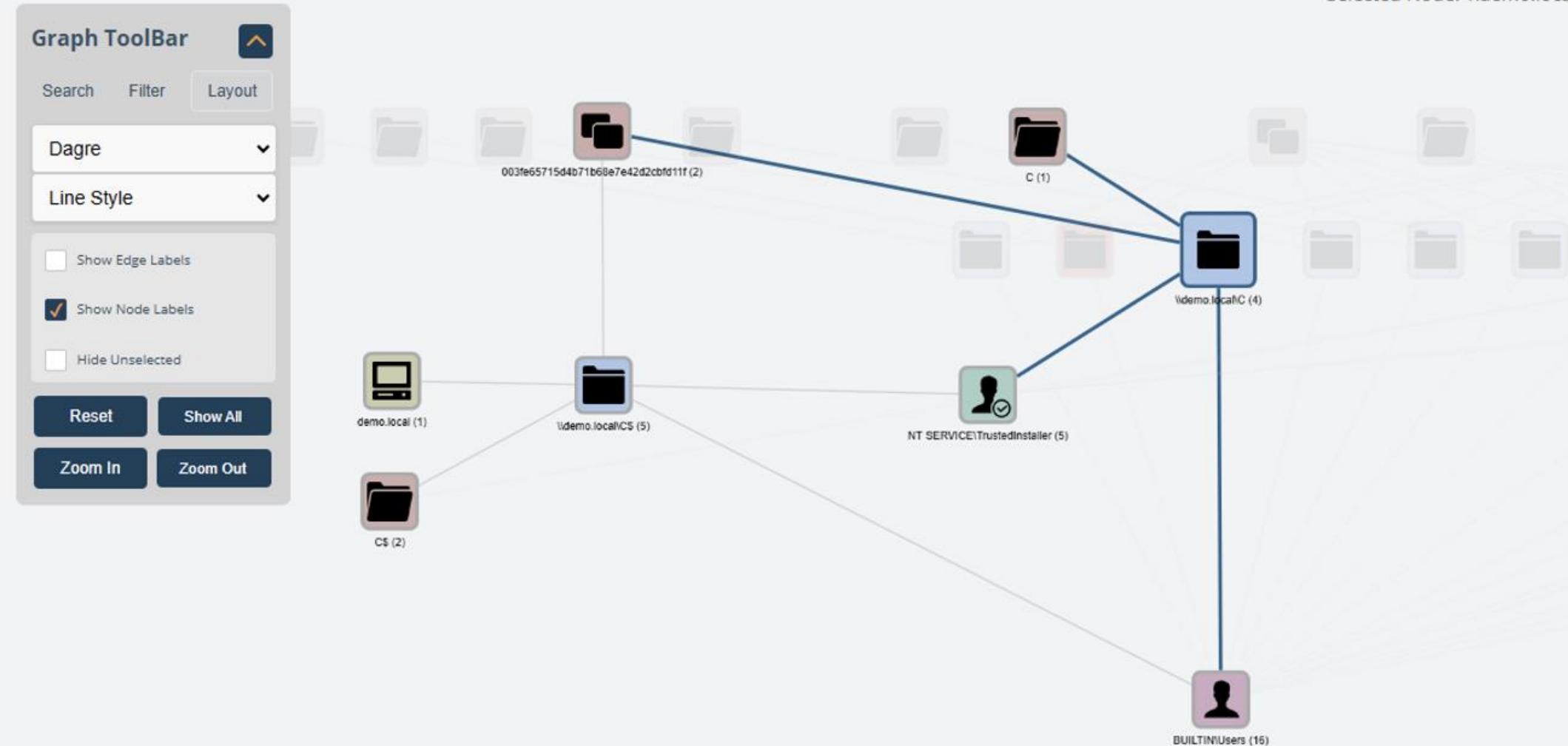
## ACT

- Exploit
- Detect
- Remediate

## ShareGraph

This section includes an experimental interactive graph for exploring share relationships.

8 Nodes 9 Edges  
Selected Node: \\demo.local\C



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**ShareGraph****TARGET**

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## ShareGraph

This section includes an experimental interactive graph for exploring share relationships.

8 Nodes 9 Edges  
Selected Node: \\demo.local\C

**Graph ToolBar**

Search Filter Layout

Dagre

Line Style

Show Edge Labels

Show Node Labels

Hide Unselected

Reset

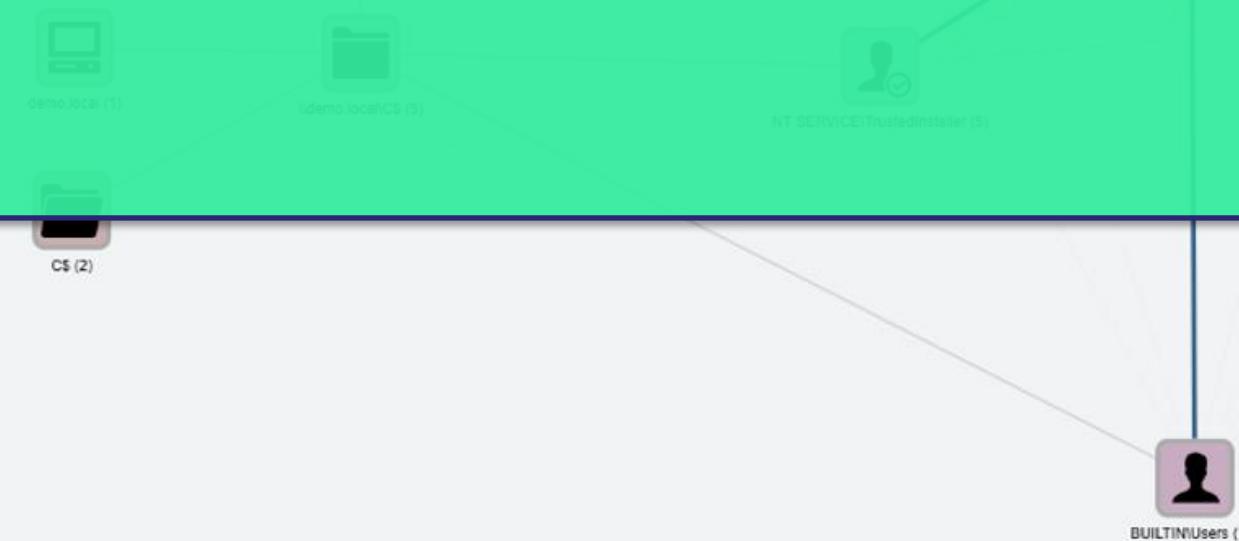
Show All

Zoom In

Zoom Out

**Not designed to be an attack path graphing tool.**

**Intended for share exploration and story telling.**



# Finding Nodes That Matter

# Finding Nodes that Matter

*"Are there things I'm not thinking of and what other tools are available?"*

## Explored Neo4j Graph Data Science (GDS) library

<https://neo4j.com/docs/graph-data-science/current/algorithms/>

## 30 algorithms reviewed

I was looking for problems for these solutions ;)

## Algorithms I liked in Neo4j

- Page Rank - What nodes have the most influence?
- Betweenness Centrality - What nodes act as a bridge?
- Louvain - What are natural node clusters?

All of the algorithms I liked were also available in Cytoscape.js 😊



# Finding Nodes That Matter

- **Finding Nodes that Matter: PageRank**

Supported by [Cytoscape.js](#) and [Neo4j](#)

## Page Rank

*"What are the most Influential Nodes?"*

### Why Should I Care?

- **Offense** can identify which nodes will provide access to resources, routes, etc.
- **Defense** can do the same and add preventative, detective and corrective controls to make them more resilient to attack

### Simple Example

When experimenting with simple Active Directory environment graphs, Page Rank could be used to identify the most influential nodes... guess which node do you think was most influential?



# Finding Nodes That Matter

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Supported by [Cytoscape.js](#) and [Neo4j](#)

## Page Rank

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### Domain

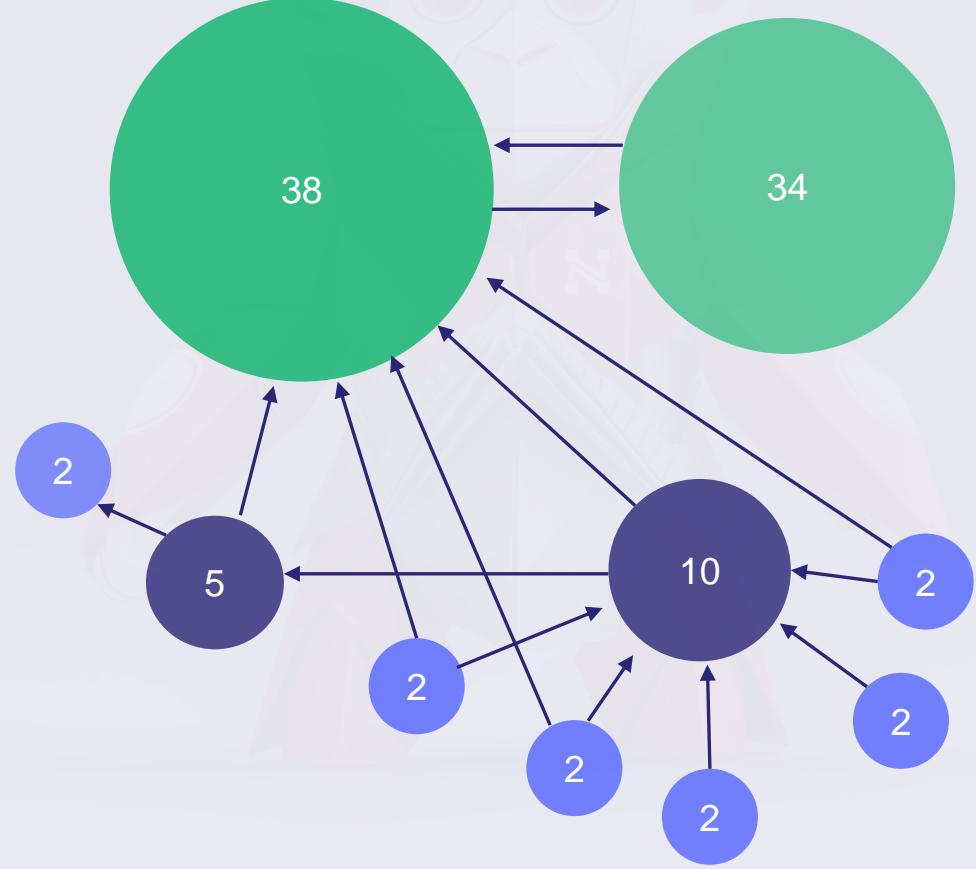


# Finding Nodes That Matter

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Supported by [Cytoscape.js](#) and [Neo4j](#)

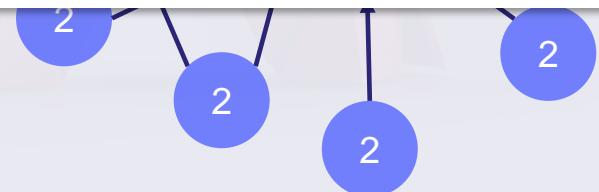
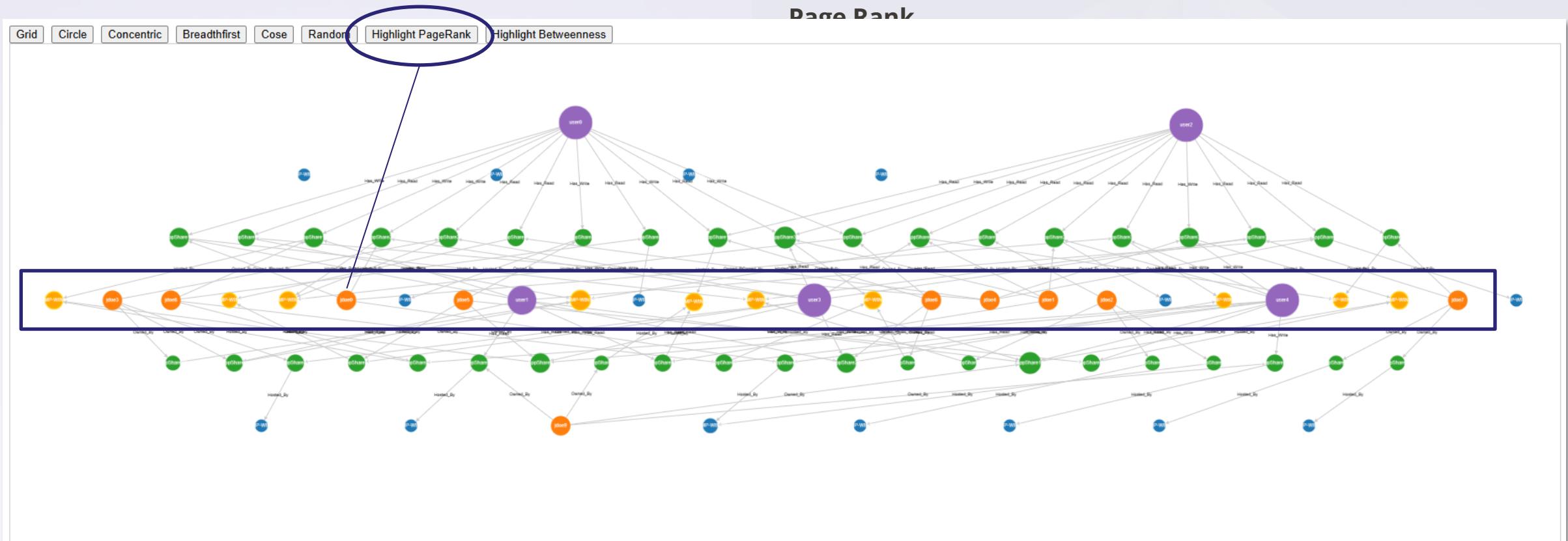
## Page Rank

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# Finding Nodes That Matter

## Page Rank



# Finding Nodes That Matter

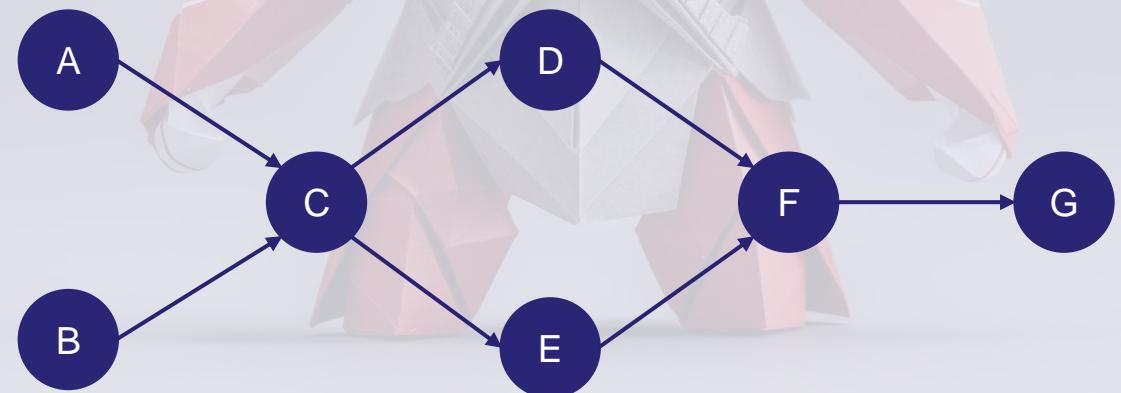
- Finding Nodes that Matter: PageRank
- **Finding Nodes that Matter: Betweenness**  
Supported by [Cytoscape.js](#) and [Neo4j](#)

## Betweenness Centrality

"Which nodes lie on the **shortest paths** between other nodes?" aka they act like bridges between communities of nodes.

### Why Should I Care?

- We may be able to determine which nodes are providing attackers with the greatest mobility.
- Prioritizing their remediation may help reduce risk or the speed at which attackers can move.



# Finding Nodes That Matter

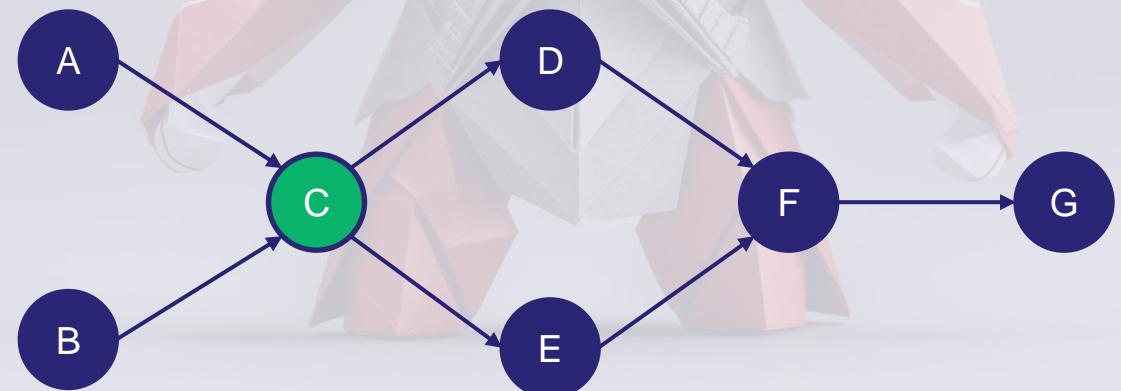
- Finding Nodes that Matter: PageRank
- **Finding Nodes that Matter: Betweenness**  
Supported by [Cytoscape.js](#) and [Neo4j](#)

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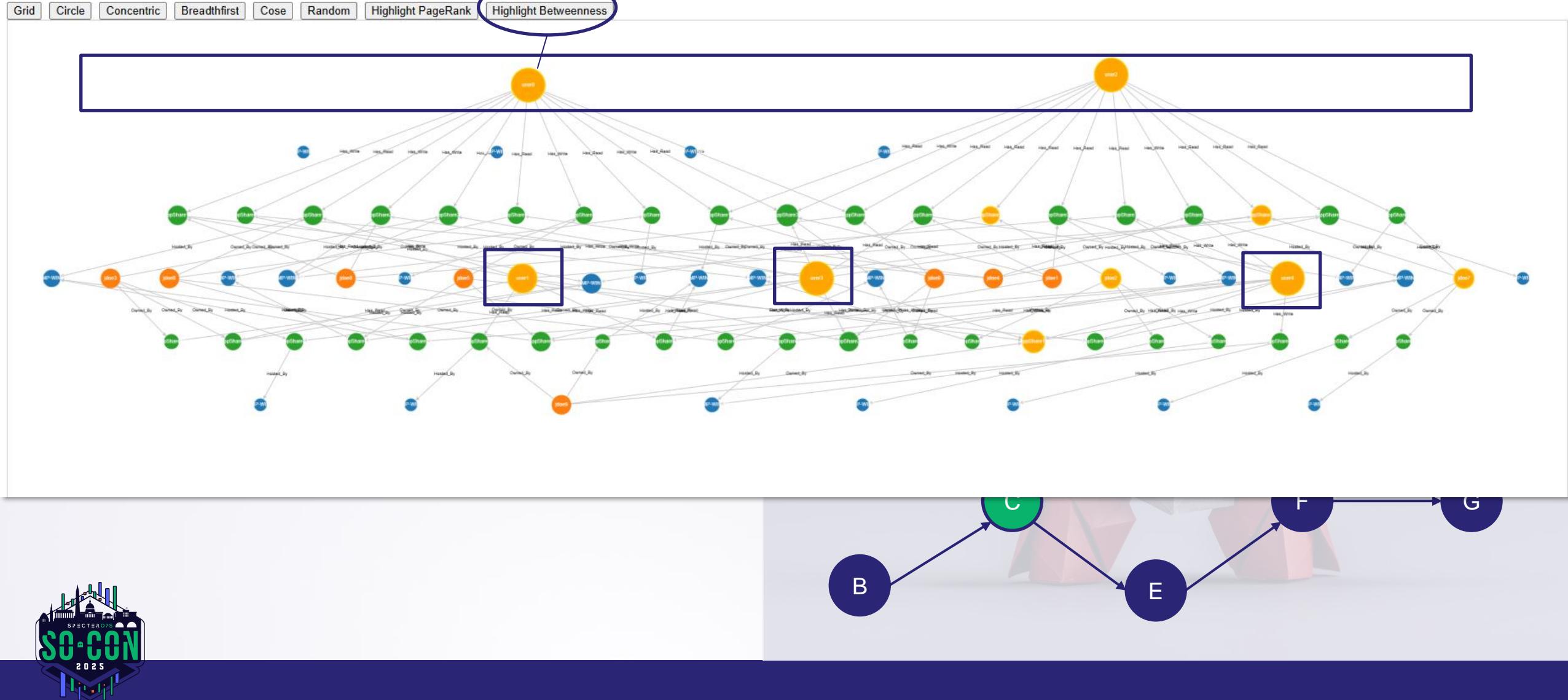
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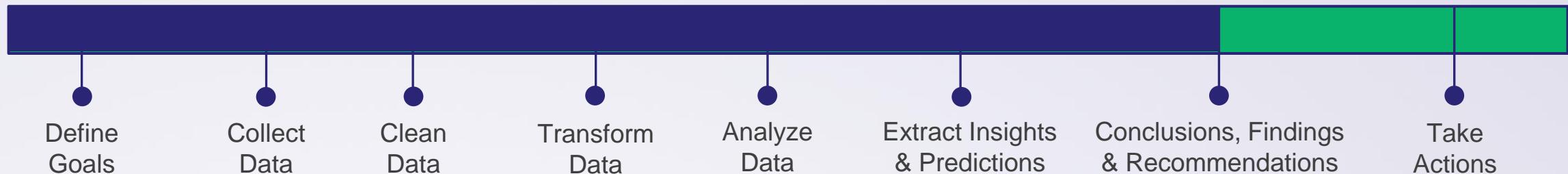


# Finding Nodes That Matter

## Betweenness Centrality



# PowerHuntShares Process

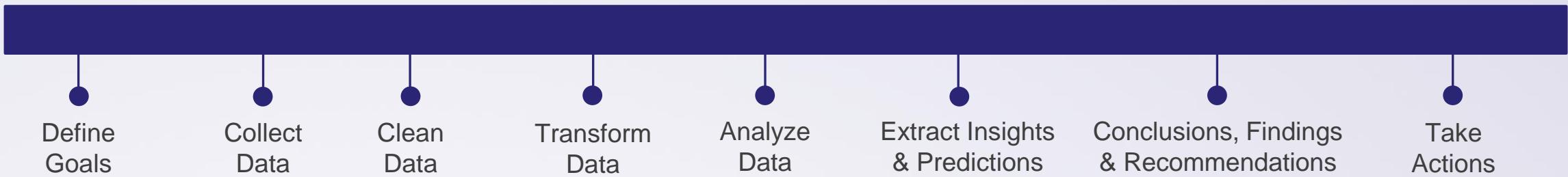


## Conclusions, Findings, & Recommendations

- How many shares are vulnerable?
- What shares are most vulnerable?
- When were the shares created?
- What application will be affected if we fix this?
- How can I remediate shares efficiently?
- How they should and do compare to peers?



# PowerHuntShares Process



## Take Actions

- Exploit
- Remediate
- Detect



# PowerHuntShares

## Demo



POWERHUNTHASHES demo.local NetSPI™

**RESULTS**

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- ShareGraph

**TARGET**

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- Remediate

## Summary Report

Testing was conducted between 11/07/2024 08:08:31 and 11/07/2024 08:10:31 to identify network shares configured with excessive privileges hosted on computers joined to the demo.local domain. In total, 13 critical, 6 high, 6 medium, and 22 low risk **ACE (Access Control Entry)** configurations were discovered across 16 shares, hosted by 2 computers in the demo.local Active Directory domain. Overall, 83 interesting files were found accessible to all domain users that could potentially lead to unauthorized data access or remote code execution. The affected shares were found hosting 53 files that may contain passwords and 0 files that may contain sensitive data. 143 credentials were recovered from 50 of the discovered 53 secrets files.

The section provides a summary of the affected assets, findings, data exposure, share creation timelines, peer comparison and general recommendations.

**Finding Exposure Summary**

Critical <b>13</b> findings	High <b>6</b> findings	Medium <b>6</b> findings	Low <b>22</b> findings
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**Affected Asset Count by Risk Level**

Risk Level	Count
Critical	13
High	6
Medium	6
Low	22

**Data Exposure Summary**

Interesting <b>83</b> files found	Sensitive <b>0</b> files found	Secrets <b>53</b> files found	Extracted <b>143</b> secrets (50 files)
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**Interesting File Exposure**

Type	Count
Secret	50
SystemImage	2
Database	0
Backup	2
Script	0
Binaries	26

# Take Aways

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- Play with your data!
- Use data analysis tools to help improve your quality of life as a defender or tester.
- Not all solutions require LLMs, but they can help save time!
- PowerHuntShares can be another tool in the box





# Thank you

Good luck and hack responsibly.



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