

A Machine Learning Approach to Forensic Investigations (KNN)

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Problem/Model?

- Large datasets are time consuming to manually comb through, ML models can be applied for classification (i.e. Malicious, Not malicious)
- We will be looking at Windows event logs (4624 – Successful log on) as a PoC
- Model selection – K- Nearest neighbours
 - Supervised (labelled) ML model, Uses feature similarity for classification, Ignores data distribution patterns



Data

- Models are only as good as the data they are trained on.
- Collect and label data
 - Honeypot; naturally generates varied malicious activity
 - Pre-existing data sets; Labelled/Not Labelled
 - Simulation of attacks in a controlled Lab; highly specific
- My data: Simulated Active Directory lab, where lateral movement (pass-the-hash) and User and Device enumeration (Bloodhound)
- Raw data was extracted using EvtxECmd, Labelled with the help of APT

Feature Generation

- Time, Account name, Workstation name, Source IP
- Raw data -> Meaningful data
- Features of this data:
 - Islan: Is it a Lan IP?
 - Isnewip: Are there reoccurring logins in past 7 days from this IP?
 - Isvpn: Is it a VPN?
 - Precent: % of logins in the past day?
 - Src_ip_c: Successful login in 15min window?
 - Tag: Malicious or not?

Code

```
✓ ▶ feature_cols = ['isnewuser', 'isnewip', 'isvpn', 'islan', 'percent', 'src_ip_c']
X = df[feature_cols]
y = df['tag']
# train and split data for testing
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4)
k_range = range(1, 26)
scores = []
# Finding the best K
for k in k_range:
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(X_train, y_train)
    y_pred = knn.predict(X_test)
    scores.append(metrics.f1_score(y_test, y_pred))

    targetname = ['Normal', 'Malicious']
    result = metrics.classification_report(y_test, y_pred, target_names=targetname)
    matrix = metrics.confusion_matrix(y_test, y_pred, labels = [0, 1])
    print('Currently running K:' + str(k))
    print(result)
    print(matrix)
# Plotting as graph
plt.plot(k_range, scores)
plt.xlabel('Value of K for KNN')
plt.ylabel('Testing Accuracy')
plt.show()
```

- Models tested:
Xgboost, KNN, ANN
- Snippet of the KNN model
- 60/40, train test split
- 25 epochs
- Code/data available on github

Training and Results

| | | | | | | | | |
|-----------|--------|---------|-----------|--------|---------|----------|--------|---------|
| islan | 1 | -0.07 | -0.088 | -0.012 | -0.39 | -0.34 | -0.22 | -0.32 |
| isnewip | -0.07 | 1 | 0.98 | 0.036 | -0.37 | -0.62 | 0.85 | -0.63 |
| isnewuser | -0.088 | 0.98 | 1 | -0.02 | -0.37 | -0.62 | 0.87 | -0.63 |
| isvpnp | -0.012 | 0.036 | -0.02 | 1 | -0.076 | -0.029 | -0.017 | -0.024 |
| percent | -0.39 | -0.37 | -0.37 | -0.076 | 1 | 0.5 | -0.38 | 0.48 |
| src_ip_c | -0.34 | -0.62 | -0.62 | -0.029 | 0.5 | 1 | -0.54 | 1 |
| tag | -0.22 | 0.85 | 0.87 | -0.017 | -0.38 | -0.54 | 1 | -0.55 |
| total_c | -0.32 | -0.63 | -0.63 | -0.024 | 0.48 | 1 | -0.55 | 1 |
| | islan | isnewip | isnewuser | isvpnp | percent | src_ip_c | tag | total_c |

```

Normal      1.00    0.99    1.00    3997
Malicious   0.98    1.00    0.99    1267

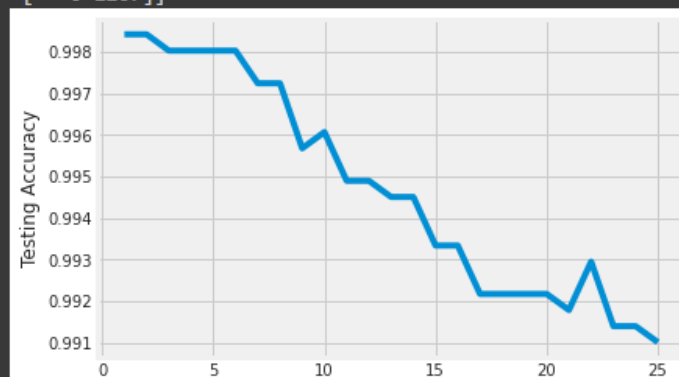
accuracy    1.00    0.99    1.00    5264
macro avg   0.99    1.00    0.99    5264
weighted avg 1.00    1.00    1.00    5264

```

```

[[3974  23]
 [  0 1267]]

```



Future works

- Create more complex model (deep learning) with larger databases and more inputs.
 - > Leverage currently archived investigation data using investigation notes, kape, apt...etc to create the ***holy grail*** in AI models for IR
- Application of ML in SOC? – Open call for ideas
- Ongoing ML projects – Malware classification using tagging
- Thank you for your attention!