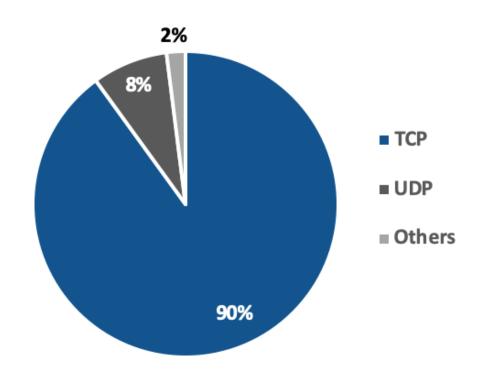
A Hybrid Method for TCP Connection Attack Prediction

Contents

- Introduction
- Stat-Filter
- SE-GNN
- Experiments
- Conclusion

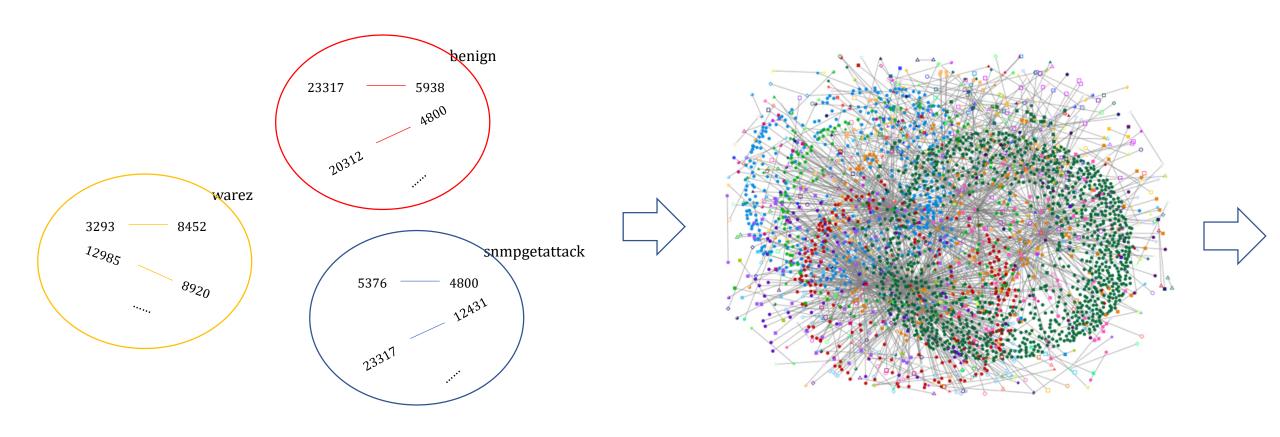
Introduction

• TCP Connection Attack Prediction





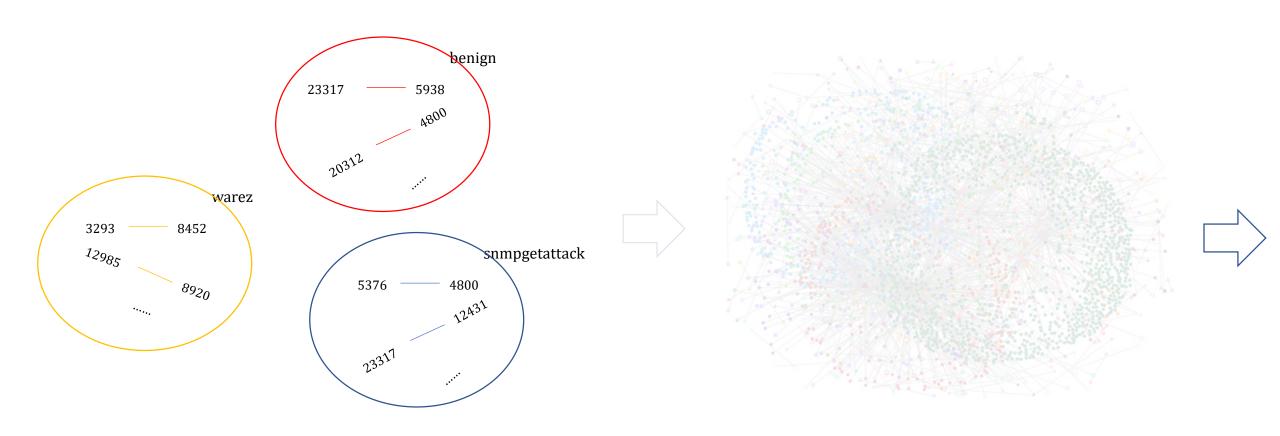
Overview of our model



Stat-Filter

SE-GNN

Overview of our model



Stat-Filter

SE-GNN

Stat-Filter

- A statistics-based algorithm using only rules
- Fast and straightforward
- Small dataset
- Simple information

Data preprocessing

- Input: <u>Source id</u>, <u>Destination id</u>, <u>Port</u>,
 <u>Timestamp</u>
- Output: <u>C</u>onnection type [stored in a list in dictionary order: ('-', 'apache2', ..., 'warezclient')]
- The tuple sets are collected and the corresponding type tables are built
- We also collect the timestamps for each SDP tuple

```
i in range(TRAIN SIZE):
with open(os.path.join('train', 'train %03d.txt' % i)) as f:
    data = f.readlines()
for line in data:
    if line[-1] == '\n':
        line = line[:-1]
    info list = line.split('\t')
    s, d, p, t = [int(info_list[j]) for j in range(4)]
    c = attack types.index(info list[4])
    train sdp total.add((s, d, p))
    train sd total.add((s, d))
    train dp total.add((d, p))
    train_sp_total.add((s, p))
    train s total.add(s)
    train d total.add(d)
    train p total.add(p)
    type_table_sdp[(s, d, p)].add(c)
    type_table_sd[(s, d)].add(c)
    type_table_dp[(d, p)].add(c)
    type_table_sp[(s, p)].add(c)
    type_table_s[s].add(c)
    type_table_d[d].add(c)
    type_table_p[p].add(c)
    time_list_sdp[(s, d, p)].add(t)
```

Observations on type tables

- Almost all SDP tuples in the training set correspond to a unique type
- → We assume that each SDP tuple in the validation set or the test set corresponds to a unique type
- What about other types of tuples?
- unique type in training set $\stackrel{\checkmark}{\Rightarrow}$ same type in validation or test set

Reliabilities of unique-type tuples

• SDP: 530/530

• DP: 524/525

• SP: 709/709

• SD: 60702/60799

• S: 11185/11185

• D: 32801/32801

• P: 1338/1344

```
valid unique type sdp = [(s, d, p) for (s, d, p) in valid sdp total if len(type table <math>sdp[(s, d, p)]) == 1]
valid unique type dp = [(s, d, p) \text{ for } (s, d, p) \text{ in valid sdp total if len(type table dp[(d, p)]) == 1]}
valid_unique_type_sp = [(s, d, p) for (s, d, p) in valid_sdp_total if len(type_table_sp[(s, p)]) == 1]
valid unique type sd = [(s, d, p) \text{ for } (s, d, p) \text{ in valid sdp total if } len(type table <math>sd[(s, d)]) == 1]
valid_unique_type_s = [(s, d, p) for (s, d, p) in valid_sdp_total if len(type_table_s[s]) == 1]
valid_unique_type_d = [(s, d, p) for (s, d, p) in valid_sdp_total if len(type_table_d[d]) == 1]
valid_unique_type_p = [(s, d, p) \text{ for } (s, d, p) \text{ in valid_sdp_total if len(type table } p[p]) == 1]
print(len(valid unique type sdp),
      sum((type table_sdp[(s, d, p)].issubset(possible_answers[(s, d, p)])) for (s, d, p) in valid_unique_type_sdp))
print(len(valid unique type dp),
      sum((type_table_dp[(d, p)].issubset(possible_answers[(s, d, p)])) for (s, d, p) in valid_unique_type_dp))
print(len(valid unique type sp),
      sum((type_table_sp[(s, p)].issubset(possible_answers[(s, d, p)])) for (s, d, p) in valid_unique_type_sp))
print(len(valid_unique_type_sd),
      sum((type_table_sd[(s, d)].issubset(possible_answers[(s, d, p)])) for (s, d, p) in valid_unique_type_sd))
Counter(min(type_table_sd[(s, d)]) for s, d, p in valid_sdp_total if
        len(type_table_sd[(s, d)]) == 1 and min(type_table_sd[(s, d)]) not in possible answers[(s, d, p)])
print(len(valid unique type s),
      sum((type_table_s[s].issubset(possible_answers[(s, d, p)])) for (s, d, p) in valid_unique_type_s))
print(len(valid unique type d),
      sum((type table d[d].issubset(possible answers[(s, d, p)])) for (s, d, p) in valid unique type d))
print(len(valid unique type p),
      sum((type table p[p].issubset(possible answers[(s, d, p)])) for (s, d, p) in valid unique type p))
Counter((p, min(type_table_p[p])) for s, d, p in valid_sdp_total
        if len(type table p[p]) == 1 and min(type table p[p]) not in possible answers[(s, d, p)])
```

Possible answers

- The possible types of an SDP tuple that appears in a query file is the set of attack types listed in the corresponding answer file + non-attack
- If an SDP tuple appears in multiple query files, we take the intersection

```
possible answers = dict.fromkeys(valid sdp total, set())
for i in range(VALID_SIZE):
    with open(os.path.join('valid query', 'valid query %03d.txt' % i)) as f:
        query_data = f.readlines()
    with open(os.path.join('valid_answer', 'valid_answer %03d.txt' % i)) as f:
        answer data = f.readlines()
    if not answer data:
        ground truth = \{0\}
        ground truth = set([attack types.index(t)
                            for t in answer data[0].split('\t')]) | {0}
    for line in query data:
        if line[-1] == '\n':
            line = line[:-1]
        info list = line.split('\t')
        s, d, p = [int(info_list[j]) for j in range(3)]
        if not possible_answers[(s, d, p)]:
            possible_answers[(s, d, p)] = ground_truth.copy()
        else:
            possible_answers[(s, d, p)].intersection_update(ground truth)
```

Prediction on the unique-type tuples

- Q1: What about the 72 SDP tuples whose corresponding unique-type SD tuples fail to match?
- Q2: What about the tuples not containing unique-type tuples?
- Q3: What about the tuples never seen?

```
prediction valid = dict.fromkeys(valid sdp total)
    sdp in valid sdp total:
    s, d, p = sdp
    sd = (s, d)
    if s not in train s total: # s, never seen
       prediction valid[sdp] = {0}
    elif sdp in train sdp total: # sdp, seen before
       prediction_valid[sdp] = type_table_sdp[sdp].copy()
            if len(type table sdp[sdp]) == 1 else {0}
    elif unique type s(s): # s, unique type
        prediction valid[sdp] = type table s[s].copy()
    elif unique type d(d): # d, unique type
        prediction valid[sdp] = type table d[d].copy()
    elif unique type p(p): # p, unique type
        prediction_valid[sdp] = type_table_p[p].copy()
    elif unique_type_dp(s, d, p): # dp, unique type
        prediction_valid[sdp] = type_table_dp[dp].copy()
    elif unique type sp(s, p): # sp, unique type
        prediction valid[sdp] = type_table_sp[sp].copy()
    elif unique type sd(s, d): # sd, unique type
        prediction_valid[sdp] = type_table_sd[sd].copy()
```

Vague types

• Recall:

- 7: ipsweep, 17: satan, 18: smurf, 19: smurfttl, 23: warez, 24: warezclient
- Our strategy is dealing with these "vague types" first and then ignoring them in the following processes
- After dealing with them, we can assume that we can always pick the answer in the types of each sd tuple as long as it can be found in the training set

7: ipsweep

```
Counter(tuple(v) for v in type_table_dp.values() if 7 in v)
any(type_table_dp[(d, p)] == {7} for s, d, p in valid_sdp_total)
Counter(tuple(v) for v in type_table_p.values() if 7 in v)
Counter(tuple(v) for v in type table s.values() if 7 in v)
Counter(tuple(v) for v in type_table_d.values() if 7 in v)
sum(type_table_d[d] == {7} for s, d, p in valid_sdp_total)
sum(type_table_d[d] == {7} and 7 in possible_answers[(s, d, p)]
    for s, d, p in valid_sdp_total)
sum(type_table_d[d] == {7, 16} for s, d, p in valid_sdp_total)
sum(type_table_d[d] == {7, 16} and 7 in possible_answers[(s, d, p)]
    for s, d, p in valid_sdp_total)
any(type table d[d] == \{7, 16\} and 16 in type table s[s]
    for s, d, p in valid_sdp_total)
```

- It is highly concentrated on destination ids with unique type (7) or types 7 and 16
- All SDP tuples in the validation set containing SD tuples with types 7 and 16 actually have 7 as one of its possible answers
- Checking type tables of source ids is a possible way to provide statistical evidence

```
elif unique_type_d(d): # d, unique type
    prediction_valid[sdp] = type_table_d[d].copy()
elif type_table_d[d] == {7, 16}:
    prediction_valid[sdp] = {0} if 16 in type_table_s[s] else {7}
```

17: satan

- It is concentrated neither on source ids nor destination ids
- Besides, no obvious temporal patterns can be observed
- We just use some strong conditions

```
elif 17 in type_table_sd[sd]:
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction_valid[sdp] = {17} if
    connection_times > 1000 and
    span <= 0.1 * connection_times else {0}</pre>
```

18: smurf & 19: smurfttl

- Basically we need to check cases when source ids have type 18 and 19
- As 19 is much rarer than 18 and the temporal patterns are similar, we add a strict condition to make the algorithm tend to predict 18 in such cases

```
Counter(tuple(v) for v in type_table_s.values() if 18 in v)
 Counter(tuple(v) for v in type table d.values() if 18 in v)
 Counter(tuple(v) for v in type table d.values() if 19 in v)
Counter(tuple(v) for v in type_table_s.values() if 19 in v)
len([(s, d, p) for s, d, p in valid_sdp_total if type_table_s[s] == {18}])
len([(s, d, p) for s, d, p in valid_sdp_total if type_table_s[s] == {19}])
len([(s, d, p) for s, d, p in valid_sdp_total if type_table_s[s] == {18, 19}])
any((s, d, p) for s, d, p in valid_sdp_total if 19 in possible_answers[(s, d, p)])
```

```
elif unique_type_s(s): # s, unique type
    prediction_valid[sdp] = type_table_s[s].copy()
elif type_table_s[s] == {18, 19}:
    prediction_valid[sdp] = {19} if 19 in type_table_dp[dp] else {18}
```

23: warez & 24: warezclient

• They are the most difficult two types in our analysis, which usually appear together and have very similar temporal patterns.

Besides, it's observed that each validation file contains at most one

of them.

```
elif type_table_sd[sd] in ({23}, {24}, {23, 24}): # 23, 24
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    if connection_times == 1 or connection_times >= 5:
        prediction_valid[sdp] = {0}
    else:
        prediction_valid[sdp] = {24} if span >= 1 else {23}
```

23: warez & 24: warezclient

 We also tried to analysis the inter-connection temporal patterns, but couldn't get any results

```
sd_23_24 = [sd for sd in train_sd_total if type_table_sd[sd] == {23, 24}]
time_type_23_24 = defaultdict(SortedList)
for t, s, d, p, c in tsdpc_list:
    if (s, d) in sd_23_24:
        time_type_23_24[(s, d)].add((t, c))
for s, d, p in valid_sdp_total:
    if (s, d) in sd_23_24 and \
        (23 in possible_answers[(s, d, p)] or 24 in possible_answers[(s, d, p)]):
        c = min(possible_answers[(s, d, p)].intersection({23, 24}))
        for t in valid_time_list_sdp[(s, d, p)]:
            time_type_23_24[(s, d)].add((t, c + 0.1))
for s, d, p in test_sdp_total:
    if (s, d) in sd_23_24:
    # c = min(possible_answers[(s,d,p)].intersection({23, 24}))
        for t in test_time_list_sdp[(s, d, p)]:
            time_type_23_24[(s, d)].add((t, 23.5))
```

				1000 24	433 23(4)	177024	100 1151
(4323, 8452)	1142 24	647 23		1355 24	584 TEST		192 TEST
140 TEST	1256 23	647 23	(12985, 15235)	1367 24	584 TEST		326 24
290 23	1324 TEST	669 23	19 24	1371 24	616 23(V)	(1013, 8452)	326 24
293 TEST	1373 23	669 23	19 24	1482 TEST	616 23(V)	7 23	338 24
306 24(V)	1480 23	736 TEST	263 24	1482 TEST	681 23	81 23	342 24
482 23	1554 23	736 TEST	444 TEST	1528 24	681 <mark>23</mark>	138 23	371 24(V)
633 TEST	1571 23	754 TEST	444 TEST	1529 24	747 TEST	155 23(V)	372 24(V)
784 23	1735 23	812 TEST	499 24	1675 23	747 TEST	304 23	382 23
811 24	1770 24	812 TEST	593 TEST	1675 23	774 TEST	392 24	382 23
920 TEST		841 23	593 TEST	1692 24(V)	775 TEST	400 24	470 TEST
1108 TEST	(12985, 12653)	841 23	605 TEST	1715 TEST	813 TEST	484 24	471 TEST
1195 23	47 24	922 23	609 TEST	1715 TEST	1032 24	720 24	474 24
1258 23	201 24	922 23	706 24		1032 24	803 23	475 24
1272 23(V)	201 24	981 23	707 23		1091 24(V)	866 23	589 23
1284 23	304 23	981 23	707 23	(12985, 22570)	1234 24	940 TEST	589 23
1489 23	304 23	1005 23	799 24	25 23	1253 24	990 TEST	796 23
1518 24(V)	376 23(V)	1005 23	799 24	25 23	1254 24	1038 24	796 23
1544 23	376 23(V)	1059 23	824 24	54 TEST	1412 24(V)	1090 TEST	852 23(V)
1601 24	390 24	1059 23	824 24	54 TEST	1412 24(V)	1413 24	852 23(V)
1707 23	390 24	1097 24	837 24	66 TEST		1426 24	
1758 23	402 24	1146 23	841 24	70 TEST	1424 24(V)	1437 23	990 23
	406 24	1146 23	1022 TEST	98 23	1428 24(V)	1470 23	990 23
(3293, 8452)	529 23	1335 24	1022 TEST	98 23	1576 24	1528 23	1029 23
191 23	529 23	1335 24	1033 24	214 24	1576 24	1619 TEST	1029 23
248 23	580 23	1347 24		214 24	1589 24		1094 23
		1348 TEST	1085 23(V)	276 24	1593 24		1094 23
344 23	580 23	1348 TEST	1085 23(V)	277 24	1594 24	(12985, 12678)	1284 24
428 TEST	592 TEST	1351 24	1290 23	298 24	1594 24	81 23	1368 24
939 23	592 TEST	1494 24	1290 23	299 24	1714 TEST	81 <mark>23</mark>	1446 24
1090 23	595 24	1498 23	1317 23	459 23(V)	1714 TEST	176 TEST	1568 23(V)
1128 24	595 24	1498 23	1317 <mark>23</mark>	435 23(V)	1776 24	176 TEST	1568 23(V)

TODO: Non-unique-type tuples

• Now, we need to deal with SD tuples with multiple types.

```
if s not in train s total: # s, never seen
   prediction_valid[sdp] = {0}
elif sdp in train sdp total: # sdp, seen before
   prediction valid[sdp] = type table sdp[sdp].copy() \
   if len(type table sdp[sdp]) == 1 else {0}
elif unique type s(s): # s, unique type
   prediction valid[sdp] = type table s[s].copy()
elif unique_type_d(d): # d, unique type
   prediction valid[sdp] = type table d[d].copy()
elif type table d[d] == \{7, 16\}:
   prediction valid[sdp] = {0} if 16 in type table s[s] else {7}
elif unique type p(p): # p, unique type
   prediction valid[sdp] = type table p[p].copy()
elif 17 in type table sd[sd]:
   timestamps = valid time list sdp[sdp]
   connection times = len(timestamps)
   span = timestamps[-1] - timestamps[0]
   prediction_valid[sdp] = {17} if connection_times > 1000 \
   and span \leftarrow 0.1 * connection times else \{0\}
elif unique type dp(s, d, p): # dp, unique type
   prediction_valid[sdp] = type_table_dp[dp].copy()
elif unique type sp(s, p): # sp, unique type
   prediction_valid[sdp] = type_table_sp[sp].copy()
```

```
elif type_table_sd[sd] in ({23}, {24}, {23, 24}): # 23, 24
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    if connection_times == 1 or connection_times >= 5:
        prediction_valid[sdp] = {0}
    else:
        prediction_valid[sdp] = {24} if span >= 1 else {23}
elif unique_type_sd(s, d): # sd, unique type
    prediction_valid[sdp] = type_table_sd[sd].copy()
```

Which types need our attention?

- Based on the prediction results on the validation set, we found that types 3, 6, 12, 13, 16, 20, 21 can be easily missed
- Besides, there are also some types can never be covered by unique-type SD tuples, because none has a unique type as one of them
- Note that some types don't appear in the validation set

```
sorted(set([min(v) for v in type_table_sd.values() if len(v) == 1]))
# [0, 3, 6, 7, 9, 10, 11, 12, 13, 16, 17, 18, 19, 20, 22, 23]
# i.e., the remaining types are [1, 2, 4, 5, 8, 14, 15, 21]
```

12: pod & 13: portsweep

```
Counter(tuple(type_table_sd[(s, d)]) for s, d, p in valid_sdp_total if 12 in type_table_sd[(s, d)] and 12 in possible_answers[(s, d, p)])
Counter(tuple(type_table_sd[(s, d)]) for s, d, p in valid_sdp_total if 13 in type_table_sd[(s, d)] and 13 in possible_answers[(s, d, p)])
Counter((len(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0]) for sdp in train_sdp_total if type_table_sdp[sdp] == {13})
Counter((len(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0]) for sdp in train_sdp_total if type_table_sdp[sdp] == {12})
Counter((len(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0]) for sdp in train_sdp_total if type_table_sdp[sdp] == {7})
Counter((len(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0]) for sdp in train_sdp_total if type_table_sdp[sdp] == {18})
Counter((len(time list sdp[sdp]), time list sdp[sdp][-1] - time list sdp[sdp][0]) for sdp in train sdp total if type table sdp[sdp] == {2})
Counter((len(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0]) for sdp in train_sdp_total if type_table_sdp[sdp] == {3})
```

12: pod & 13: portsweep

```
elif type table sd[sd] == {12, 13}:
    timestamps = valid time list sdp[sdp]
    connection times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    if (2 \leftarrow connection times \leftarrow 3 and 35 \leftarrow span \leftarrow 50) or \
             (27 \leftarrow connection times \leftarrow 30 and 56 \leftarrow span >= 59) or \
             (connection times == 2 and span == 20):
        prediction valid[sdp] = {13}
    elif connection times >= 10 and span <= 1:</pre>
        prediction valid[sdp] = {12}
        prediction valid[sdp] = {0}
elif 13 in type table sd[sd]:
    timestamps = valid time list sdp[sdp]
    connection times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction valid[sdp] = \{13\} if (2 <= connection times <= 3 and 35 <= span <= 50) or \
                                        (27 \leftarrow connection times \leftarrow 30 and 56 \leftarrow span >= 59) else {0}
elif 12 in type table sd[sd]:
    timestamps = valid time list sdp[sdp]
    connection times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction valid[sdp] = \{12\} if span <= 1 and connection times >= 10 else \{0\}
```

3: dict

- Connections that connect only once are indistinguishable by temporal patterns
- We need "special" temporal patterns of each type

```
Counter(tuple(v) for v in type_table_sd.values() if 3 in v)
Counter(tuple(type_table_sd[(s, d)]) for s, d, p in valid_sdp_total if
       3 in type_table_sd[(s, d)] and 3 in possible_answers[(s, d, p)])
Counter((len(time list sdp[sdp]), time list sdp[sdp][-1] - time list sdp[sdp][0])
       for sdp in train sdp total if type table sdp[sdp] == {3})
Counter((len(valid time list sdp[(s, d, p)]),
        valid time list sdp[(s, d, p)][-1] - valid_time_list_sdp[(s, d, p)][0])
       for s, d, p in valid_sdp_total if 3 in type_table_sd[(s, d)]
       and 3 in possible_answers[(s, d, p)])
```

```
elif type_table_sd[sd] == {3, 18}:
    prediction_valid[sdp] = {0} if 18 in type_table_dp[dp] else {3}
elif 3 in type_table_sd[sd]:
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction_valid[sdp] = {3} if 51 <= span <= 59 and 11 <= connection_times <= 14 else {0}</pre>
```

20: snmpgetattack & 21: snmpguess

```
Counter(tuple(v) for v in type table sd.values() if 20 in v)
Counter(tuple(v) for v in type_table_sd.values() if 21 in v)
Counter((len(time list sdp[sdp]), time list sdp[sdp][-1] - time list sdp[sdp][0])
        for sdp in train sdp total if type table sdp[sdp] == {20})
Counter((len(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0])
        for sdp in train_sdp_total if type_table_sdp[sdp]
```

```
elif type_table_sd[sd] == {20, 21}:
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    if connection_times == 1:
        prediction_valid[sdp] = {0}
    elif connection_times <= 13 and span >= 20:
        prediction_valid[sdp] = {20}
    elif connection_times >= 37 and 49 <= span <= 59:
        prediction_valid[sdp] = {21}
    else:
        prediction_valid[sdp] = {0}</pre>
```

6: ignore

```
Counter(tuple(v) for v in type_table_sd.values() if 6 in v)
# Counter({(6,): 13, (0, 16, 6, 7): 4, (0, 6, 7, 16, 23, 24): 2, (16, 6, 7): 2, (0, 2, 6, 7): 1, (0, 23, 6, 7): 1, (0, 6): 1, (0, 1, 18, 6): 1})
Counter(tuple(type_table_sd[(s, d)]) for s, d, p in valid_sdp_total if 6 in type_table_sd[(s, d)] and 6 in possible_answers[(s, d, p)])
# Counter({(en(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0]) for sdp in train_sdp_total if type_table_sdp[sdp] == {6})
# Counter({(1, 0): 130, (2, 38): 43, (2, 40): 30, (2, 39): 22, (2, 41): 4, (2, 42): 3, (2, 52): 1, (2, 47): 1, (2, 45): 1})
```

```
elif 6 in type_table_sd[sd]:
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction_valid[sdp] = {6} if connection_times == 2 and 38 <= span <= 42 else {0}</pre>
```

16: saint

1: apache2

• Note that type 1 doesn't appear in the validation set.

```
Counter(tuple(v) for v in type_table_s.values() if 1 in v)
# Counter({(0, 1, 6, 13, 18, 23, 24): 1, (1, 11, 17, 18, 23, 24): 1, (1, 18, 23): 1})
Counter(tuple(v) for v in type_table_d.values() if 1 in v)
# Counter({(0, 1, 2, 3, 6, 7, 9, 12, 13, 15, 17, 18, 22): 1})
Counter((len(time_list_sdp[sdp]), time_list_sdp[sdp][-1] - time_list_sdp[sdp][0])
for sdp in train_sdp_total if type_table_sdp[sdp] == {1})
# Counter({(48, 10): 2, (10, 6): 2, (216, 11): 1, (305, 59): 1, (37, 9): 1,
# (276, 28): 1, (185, 1797): 1, (52, 24): 1, (18, 26): 1, (463, 37): 1,
# (853, 25): 1, (31, 1): 1})
```

```
elif 1 in type_table_sd[sd]: # 1
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction_valid[sdp] = {1} if connection_times >= 100 and span <= connection_times / 5 else {0}</pre>
```

2: back

• Type 2 has fairly special temporal patterns.

```
elif 2 in type_table_s[s]: # 2
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction_valid[sdp] = {2} if 60 <= connection_times <= 100 and 58 <= span <= 59 else {0}</pre>
```

4: guest & 15: rootkit

• As most connections with these types connect only once, and there are not enough samples for us to get more information, we don't do positive predictions for them

5: httptunnel-e

 Most connections with type 5 have the same temporal pattern, we use that as our only prediction reference

```
elif 5 in type_table_sd[sd]: # 5
   timestamps = valid_time_list_sdp[sdp]
   connection_times = len(timestamps)
   span = timestamps[-1] - timestamps[0]
   prediction_valid[sdp] = {5} if connection_times == 2 and span == 30 else {0}
```

8: mailbomb & 14: processtable

```
elif 8 in type_table_sd[sd]: # 8
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction_valid[sdp] = {8} if 47 <= connection_times <= 49 \
        and 57 <= connection_times <= 59 else {0}

elif 14 in type_table_sd[sd]: # 14
    timestamps = valid_time_list_sdp[sdp]
    connection_times = len(timestamps)
    span = timestamps[-1] - timestamps[0]
    prediction_valid[sdp] = {14} if connection_times == 15 and \
        56 <= connection_times <= 57 else {0}</pre>
```

Final algorithm

- It is a highly "if-else" algorithm based on rules
- It works well on the validation set. but is hard to be generalized

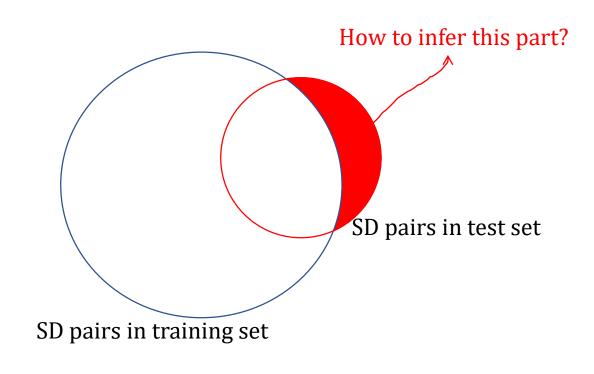
Algorithm 1: Stat-Filter

```
input: An SDP tuple x = (s, d, p) in V_{SDP} or E_{SDP}
output: Prediction of its type c(x)
CT \leftarrow |\mathbb{T}(x)|
SP \leftarrow \max \mathbb{T}(x) - \min \mathbb{T}(x)
if s \notin T_S then
return 0
else if x \in T_{SDP} then
   if |A_{SDP}(x)| = 1 then
      return the only element in A_{SDP}(x)
   return 0
else if |A_S(s)| = 1 then
   return the only element in A_S(s)
else if A_S(s) = \{18, 19\} then
   if 19 \in A_{DP}(d, p) then
   return 19
   else
   return 18
else if |A_D(d)| = 1 then
   return the only element in A_D(d)
else if A_D(d) = \{7, 16\} then
   if 16 \in A_S(s) then
    return 0
   else
    return 7
   end
else if |A_P(p)| = 1 then
   return the only element in A_P(p)
else if 17 \in A_{SD}(s, d) then
   if CT > 1000 and SP \le CT/10 then
    return 17
   else
    return 0
   end
else if |A_{DP}(d, p)| = 1 then
   return the only element in |A_{DP}(d, p)|
else if |A_{SP}(s, p)| = 1 then
  return the only element in |A_{SP}(s, p)|
else if A_{SD} \subseteq \{23, 24\} then
   if CT = 1 or CT \ge 5 then
   return \theta
   else if SP \ge 1 then
    return 24
   else
    return 23
else if |A_{SD}(s,d)| = 1 and A_{SD} contains no vague
 types then
   return the only element in A_{SD}(s, d)
else if A_{SD}(s,d) = \{12,13\} then
   if 2 < CT < 3 and 35 < SP < 50 OR
    27 \le CT \le 30 and 56 \le SP \le 59 OR
    (CT, SP) = (2, 20) then
    return 13
   else if CT \ge 10 and SP \le 1 then
   return 12
   else
   return 0
else if 13 \in A_{SD}(s, d) then
   if 2 \le CT \le 3 and 35 \le SP \le 50 OR
    27 \le CT \le 30 and 56 \le SP \le 59 then
    return 13
   return \theta
```

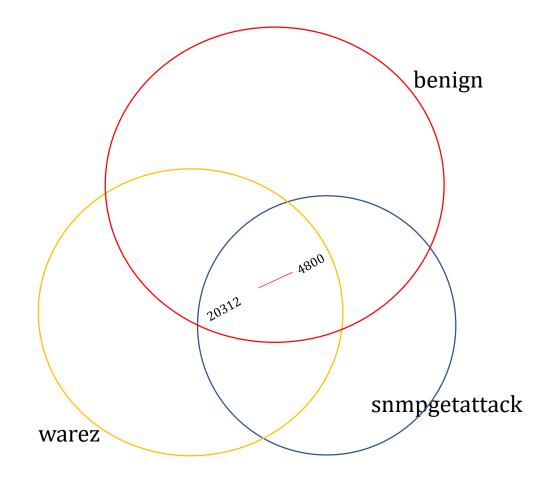
end

```
else if A_{SD}(s, d) = \{3, 18\} then
   if 18 \in A_{DP}(d, p) then
   return \theta
   return 3
else if 3 \in A_{SD}(s, d) then
  if 51 \le SP \le 59 and 11 \le CT \le 14 then
   return 3
   else
   return 0
else if 12 \in A_{SD}(s, d) then
  if CT > 10 and SP < 1 then
   return 12
   return 0
else if A_{SD}(s, d) = \{20, 21\} then
  if CT = 1 then
   return 0
   else if CT \le 13 and SP \ge 20 then
   return \overline{20}
   else if CT \ge 37 and 49 \le SP \le 59 then
   return \overline{2}1
   else
   ∟ return 0
else if 6 \in A_{SD}(s, d) then
   if CT = 2 and 38 \le SP \le 42 then
   return 6
   L return 0
else if 16 \in A_{SD}(s, d) then
   if (CT, SP) = (2, 10) OR CT = 3 and
    (\mathbb{T}(x)[2] - \mathbb{T}(x)[1], \mathbb{T}(x)[1] - \mathbb{T}(x)[0]) \in
    \{(10, 10), (9, 10), (9, 9)\} then
    return 16
   else
   L return 0
else if 1 \in A_{SD}(s, d) then
   if CT \ge 100 and SP \le CT/5 then
   return 1
   return 0
else if 2 \in A_S(s) then
  if 60 \le 100 and 58 \le SP \le 59 then
   return 2
   else
   return 0
else if 5 \in A_{SD}(s, d) then
  if (CT, SP) = (2, 30) then
   return 5
   return 0
else if 8 \in A_{SD}(s, d) then
   if 47 \le CT \le 49 and 57 \le CT \le 59 then
   return 8
   else
   return 0
else if 14 \in A_{SD}(s, d) then
  if CT = 15 and 59 \le CT \le 57 then
   return 14
   else
   return 0
else
∟ return 0
```

Current challenges

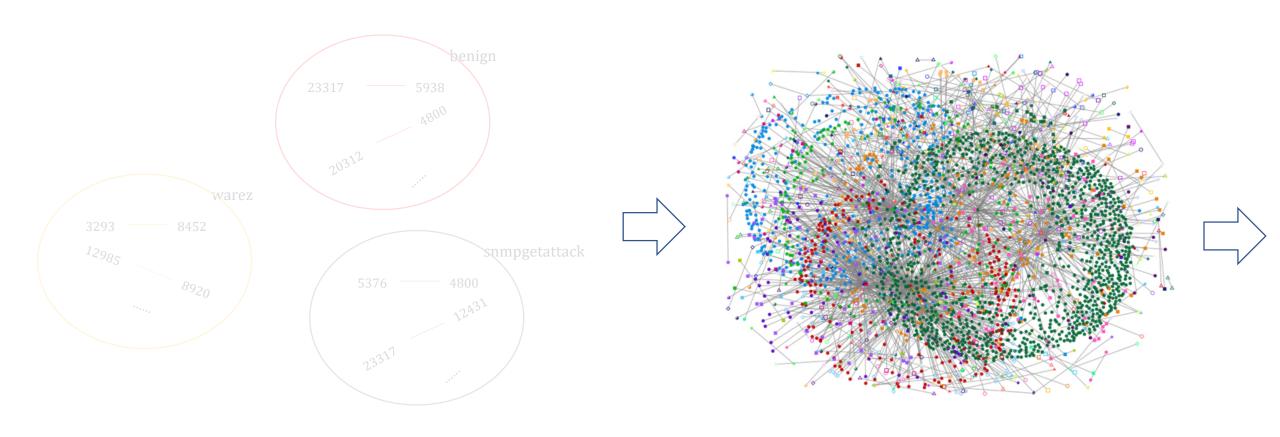


• Some SD pairs never appear in training set.



 Some SD pairs have many attack types

Overview of our model



Stat-Filter

SE-GNN

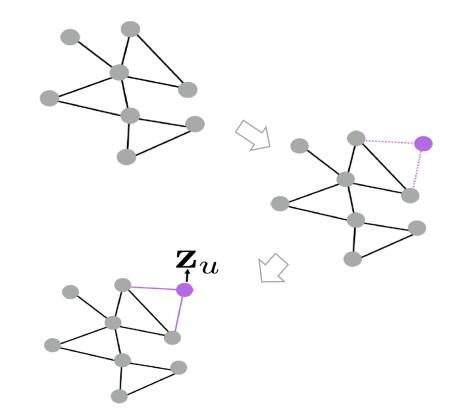
Solution: Graph



Drawn using Dephi

GraphSage

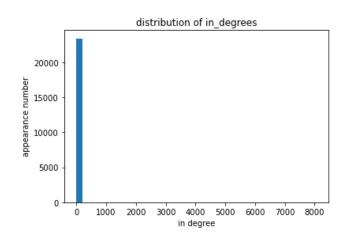
$$\mathbf{h}_{v}^{k} = \sigma \left(\left[W_{k} \cdot \sum_{u \in N(v)} \frac{\mathbf{h}_{u}^{k-1}}{|N(v)|}, \mathbf{B}_{k} \mathbf{h}_{v}^{k-1} \right] \right)$$
(1)

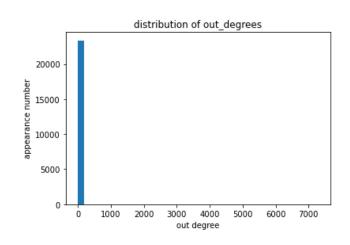


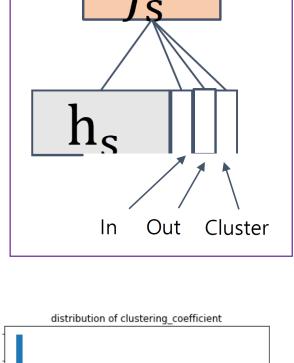
Spatial Information

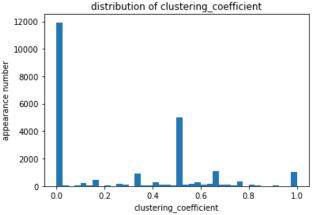
• We mainly consider in-degree, out-degree and clustering coefficient of each node.

$$\mathbf{f}_{v} = MLP_{1}\left(\left[\mathbf{h}_{v}, In_{v}, Out_{v}, Cluster_{v}\right]\right), \tag{2}$$









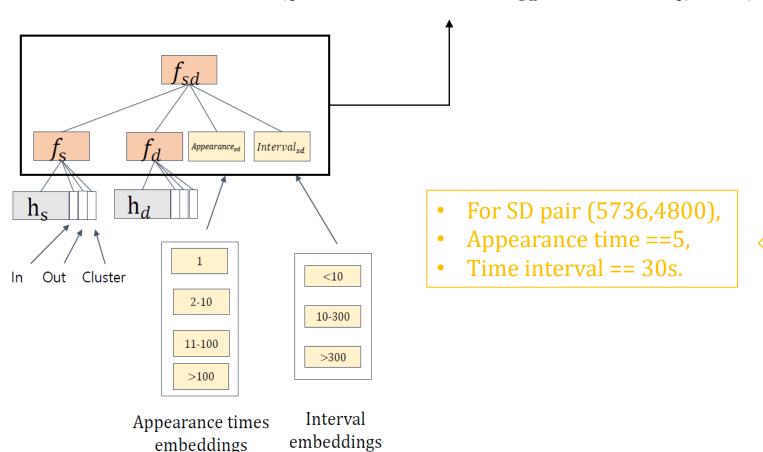
in-degree out-degree

clustering coefficient

Temporal Information

• We mainly consider the appearance times and time interval of each SD pair in a limited time window.

$$\mathbf{f}_{sd} = MLP_2\left(\left[\mathbf{f}_s, \mathbf{f}_d, \mathbf{Appearance}_{sd}, \mathbf{Interval}_{sd}\right]\right).$$
 (3)



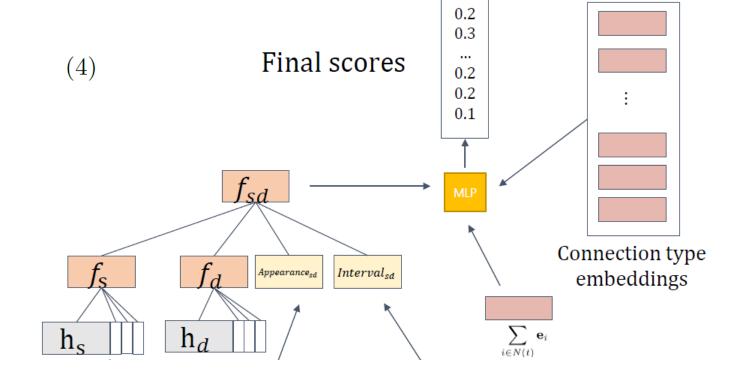
				I
5376	4800	42629	243	snmpgetattack
23317	12431	42629	243	snmpgetattack
20312	4800	42629	243	-
23317	5938	42629	243	-
15394	4800	6870	245	-
23317	882	6870	245	-
3293	8452	6870	248	warez
12985	8920	6870	249	warez
12985	8920	6870	249	warez
20312	4800	6870	253	-
23317	5038	6870	253	-
5376	4800	6870	253	snmpgetattack
23317	12431	6870	253	snmpgetattack
5376	4800	6870	258	snmpgetattack
23317	12431	6870	258	snmpgetattack
20312	4800	6870	258	-
23317	5938	6870	258	-
20312	4800	6870	263	-
23317	5938	6870	263	-
5376	4800	6870	263	snmpgetattack
23317	12431	6870	263	snmpgetattack
20312	4800	6870	268	-
23317	5938	6870	268	-
5376	4800	6870	268	snmpgetattack
23317	12451	6870	268	snmpgetattack
20312	4800	6870	273	-
23317	5038	6870	273	-
5376	4800	6870	273	snmpgetattack
23317	12431	6870	273	snmpgetattack
20312	4800	6870	278	-

Relations among connection types

• We mainly consider the appearance times and time interval of each SD pair in a limited time window.

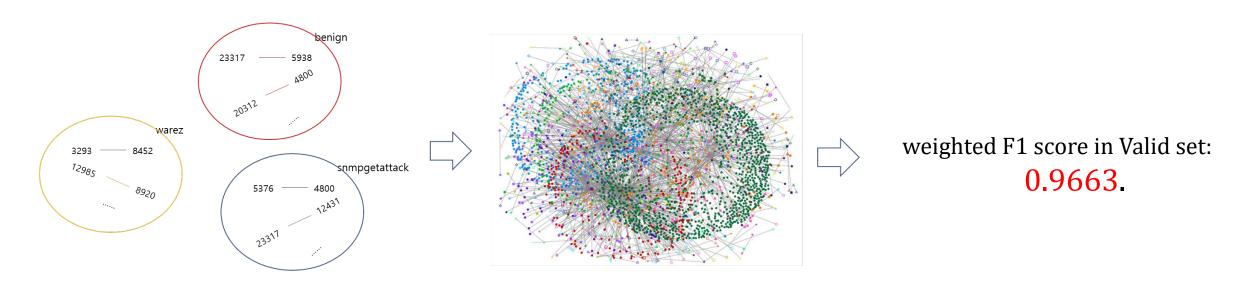
$$\hat{p}_{sd,t} = MLP_3 \left(\left[\mathbf{f}_{sd}, \sum_{i \in N(t)} \mathbf{e}_i, \mathbf{e}_t \right] \right),$$

• Negative sampling is used.



Result and Conclusion

- We gave an extensive statistical analysis, even detailed into the appearance pattern of each attack type.
- We combined both the local information and global graph information.
- We achieved an amazing weighted F1 score in Valid set: 0.9663.



Stat-Filter

SE-GNN