

# Stochastic Analysis of Horizontal IP Scanning

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# Agenda

- Introduction
- Motivation
- Formalizing scanning
- Analysis of existing methods
- Stealth optimality
- Final thoughts

# Introduction

- IDS (Intrusion Detection Systems) are commonly deployed to protect network assets
- Algorithms in IDS aim to detect
  - Malicious payload
  - Anomalous traffic patterns
  - DoS attacks
  - Scanning for open services
- To maintain scalability and adapt over time, IDS periodically expires state and performs detection using packets received **only within a given time window**

## Introduction 2

- To reduce false-positive rates, IDS must observe a minimum number of packets in the window before triggering an underlying estimator
  - This makes IDS oblivious to attacks that span multiple windows and never reach this threshold
  - We call such exploits **stealthy**
- One malicious activity whose detection is particularly sensitive to amount of IDS state is **horizontal scanning** ← **our focus here**
  - This entails probing of all BGP space on a given port
  - Similar techniques can be applied to **vertical scanning** (probing of multiple ports on a given IP)

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# Motivation

- The only exposed technique for stealth scanning is to stretch it over several months (Staniford 2002)
- This leaves many open issues:
  - Is stealth scanning possible at faster rates?
  - For a given scan rate, with what probability will existing IDS installations notice the various types of scanners?
  - How to optimally permute the IP space during the scan?
  - How to distribute the load between multiple scanner IPs?
- We aim to address these questions through probabilistic modeling

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# Formalizing Scanning

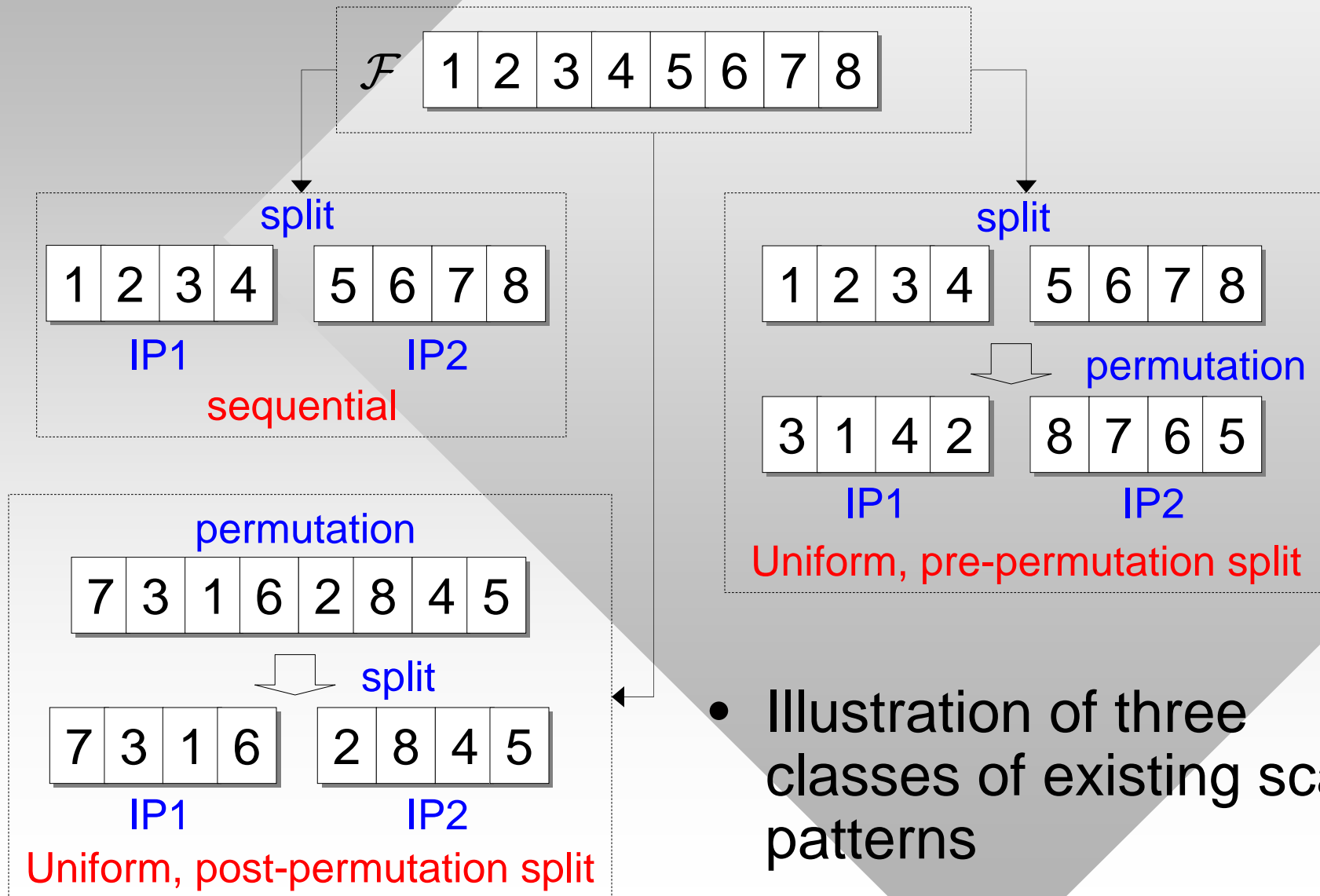
- Since no prior work analytically examined IDS detection rates, our first task is to develop a formalization that makes the problem tractable
- Assume  $\mathcal{F} = \{0, \dots, n - 1\}$  is the target IP space
  - For IPv4,  $n = 2^{32}$  addresses, later filtered by BGP
- Attacker has access to  $m$  source IPs (e.g., a botnet) from which it launches the scan
  - Not concerned with infection, only scanning
  - Thus, no new IPs are added to the botnet



# Formalizing Scanning 2

- Define a *scan pattern* to consist of:
  - **Permutation**: order in which  $\mathcal{F}$  is probed
  - **Split**: partitioning of  $\mathcal{F}$  between source IPs
  - **Schedule**: instances when probes are transmitted
- In the literature
  - Two permutations mentioned, i.e., **sequential** ( $\mathcal{F}$  remains intact) and **uniform** ( $\mathcal{F}$  is randomly shuffled)
  - Split could be applied **before** or **after** permutation, but always involved contiguous chunks of space
  - Schedule amounted to constant inter-probe spacing

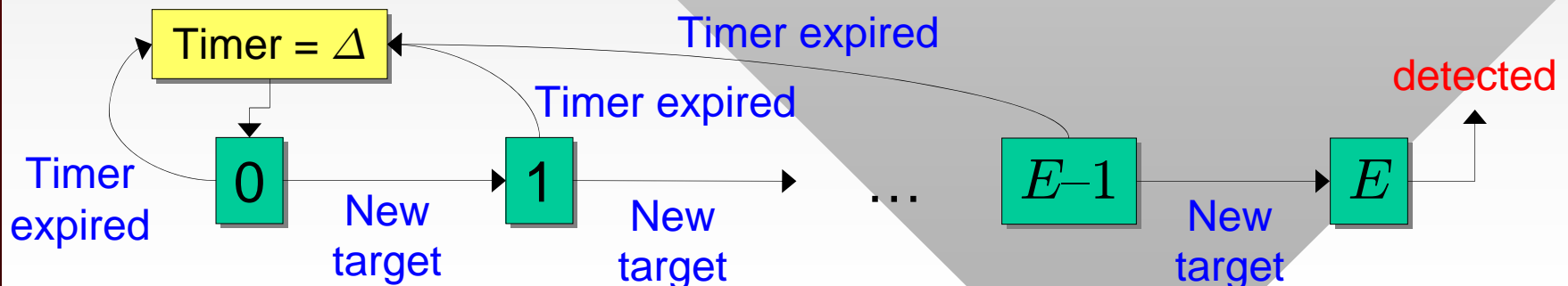
# Formalizing Scanning 3



- Illustration of three classes of existing scan patterns

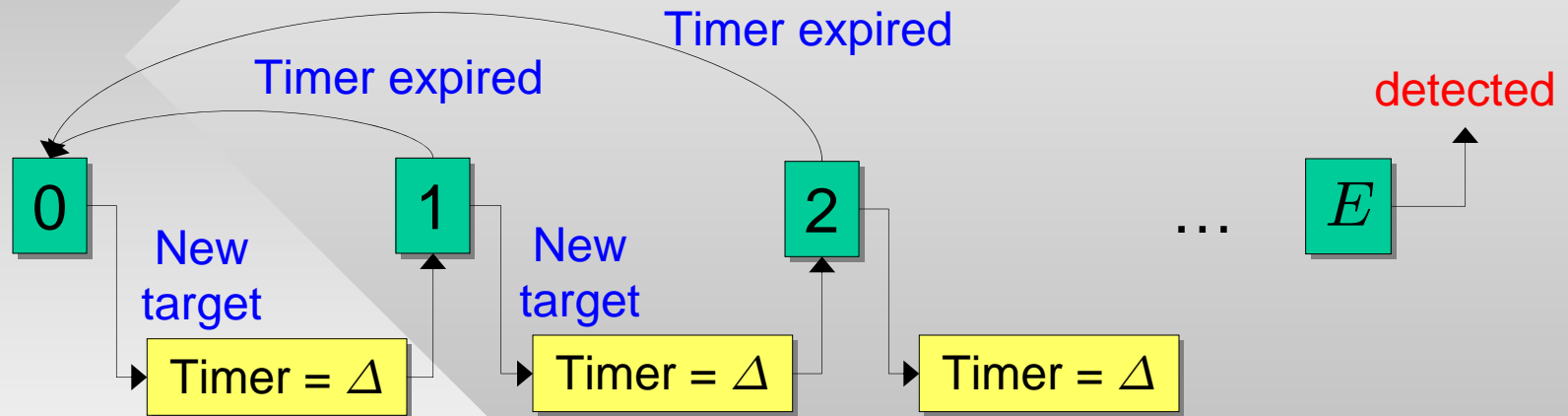
# Formalizing Scanning 4

- Consider two models of IDS behavior
  - Define  $\Delta$  to be window size in time units and  $E$  the number of scan packets that triggers an estimator
  - Estimator is assumed to **always** detect the scanner
- Model **IDS-A** (Snort and its commercial versions)
  - Described by a separate FSM for each source IP  $i$
  - FSM counts the number of unique targets probed by  $i$



# Formalizing Scanning 5

- Model **IDS-B** (Bro and certain firewalls)
  - Resets the timer each time new target is hit



- For the same pair of parameters  $(\Delta, E)$ , IDS-B detects all scanners that IDS-A does
  - But this comes at the expense of keeping separate timers for each source IP and longer lists of seen targets in steady-state

## Formalizing Scanning 6

- For each source  $i$ , IDS can be modeled as a discrete-state stochastic process (counter)  $C_i(t)$ 
  - Define  $\tau_i(t)$  to be the **first hitting time** of  $C_i(t)$  on the absorbing state  $E$  after the first packet arrives from  $i$

$$\tau_i(t) = \inf\{t > 0 : C_i(t) = E | C_i(0) = 1\}$$

- Assume  $T$  is the fixed duration of the scan
  - Then, the number of detected scanner IPs is given by random variable  $D$ :

$$D = \sum_{i=1}^m \mathbf{1}_{\{\tau_i(t) < T\}}$$

- and the IDS succeeds at detecting the scan with probability  $\rho(T) = P(D \geq 1)$

# Formalizing Scanning 7

- Define **stealth-cover time** (SCT) to be the duration of the scan that keeps detection probability  $\rho(T)$  below some threshold  $\epsilon$

$$\delta = \inf\{T > 0 : \rho(T) \leq \epsilon\}$$

- Main objectives:
  - Derive  $\delta$  for existing methods (sequential, uniform) and analyze how  $m$  and pre/post-permutation splits affect it
  - Investigate the existence of **optimal** scan patterns that minimize  $\delta$  under both IDS-A and IDS-B
  - Compare the various scan techniques to each other
- Only a portion of this is covered today

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# Analysis of Existing Methods

- Sequential scanning is very simple to analyze
  - SCT is computed for  $\epsilon = 0$  (no detection):

$$\delta = \Delta \frac{n}{m\zeta}, \quad \text{where} \quad \zeta = \begin{cases} E - 1 & \text{IDS-A} \\ 1 & \text{IDS-B} \end{cases}$$

- Observations:
  - IDS-B requires a factor of  $(E - 1)$  longer scan durations than IDS-A
  - Scan time reduces linearly with botnet size  $m$
- Scan rate at all networks is constant  $n/(mT)$ 
  - For  $T = 24$  hrs and  $m = 1$ , this is **49.7 thousand pps**
  - Clearly noticeable and intrusive



## Analysis of Existing Methods 2

- Uniform scanning is more interesting
  - The paper develops a single unifying model to handle pre/post permutation splits and different botnet sizes  $m$
- With certain approximations, IDS-A is tractable
  - Probability of noticing a scan at subnet  $s$ :

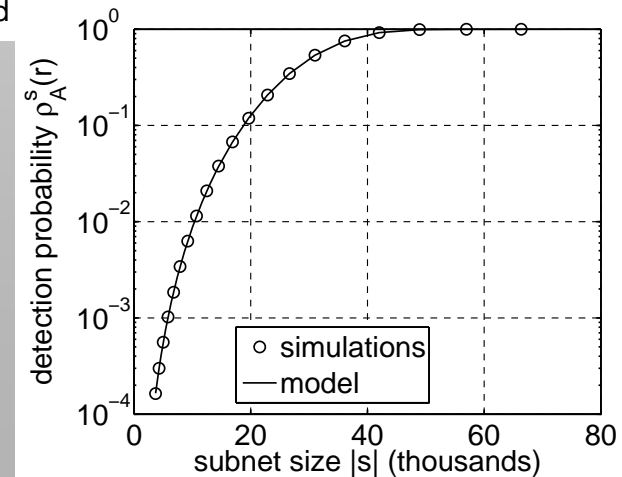
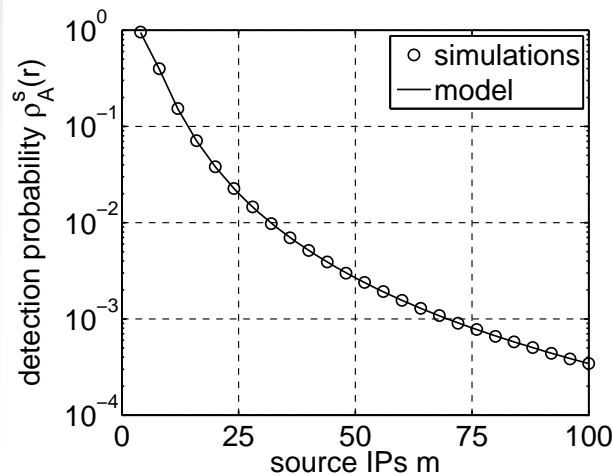
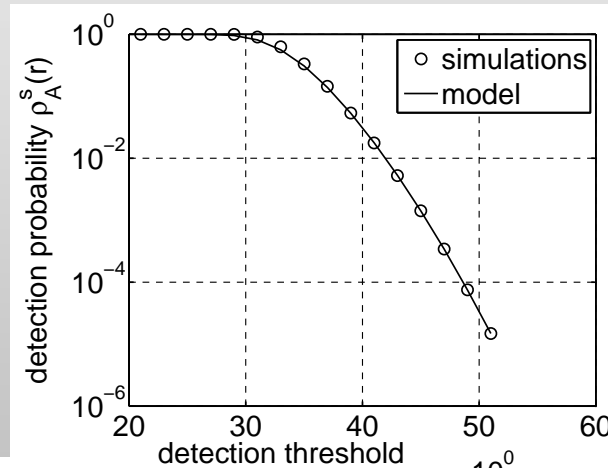
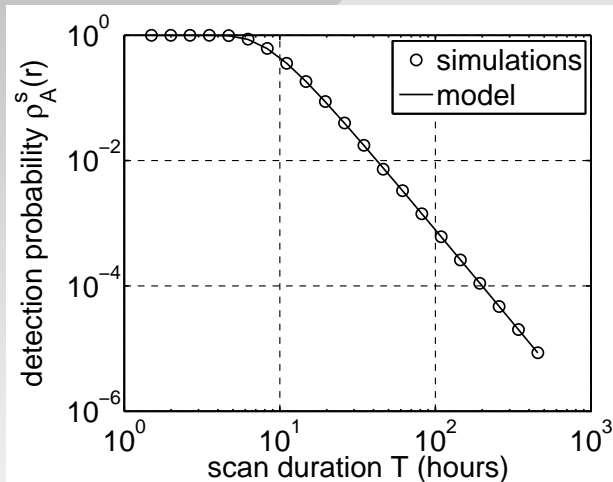
$$\rho(T) \approx 1 - \left( \sum_{j=0}^{E-1} \binom{|s|}{j} q^j (1-q)^{|s|-j} \right)^{1/q}$$

- where

$$q = \frac{\Delta}{\omega T} \quad \text{and} \quad \omega = \begin{cases} 1 & \text{pre-permutation} \\ m & \text{post-permutation} \end{cases}$$

# Analysis of Existing Methods 4

- Model accurate across all input parameters



# Analysis of Existing Methods 5

- IDS-B is more challenging
  - Larger threshold  $E$  creates non-trivial memory of previous observations of scanner probes
- Only asymptotic results are possible
  - Using the Chen-Stein theorem for sums of dependent Bernoulli variables, we have:

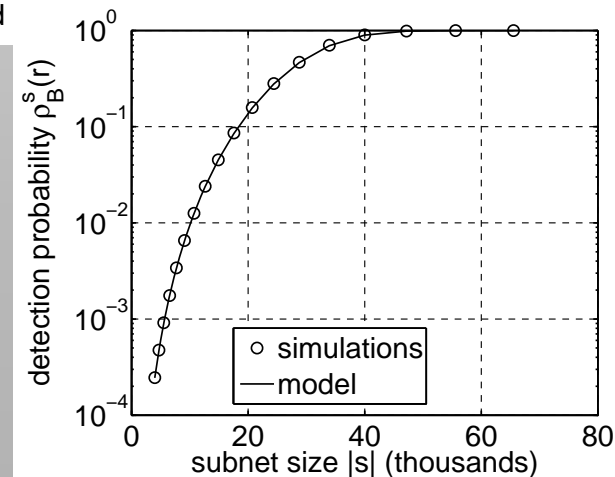
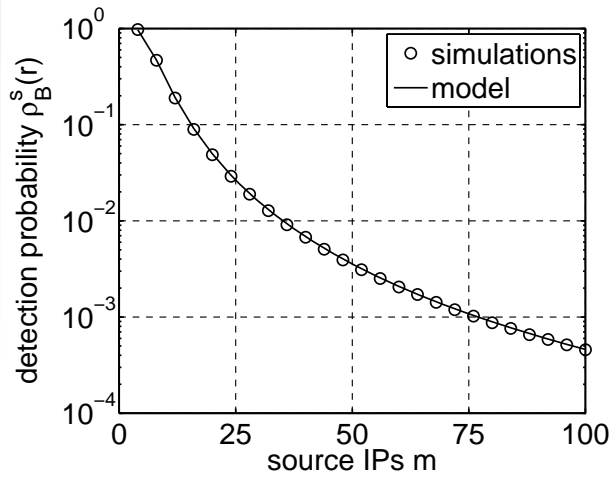
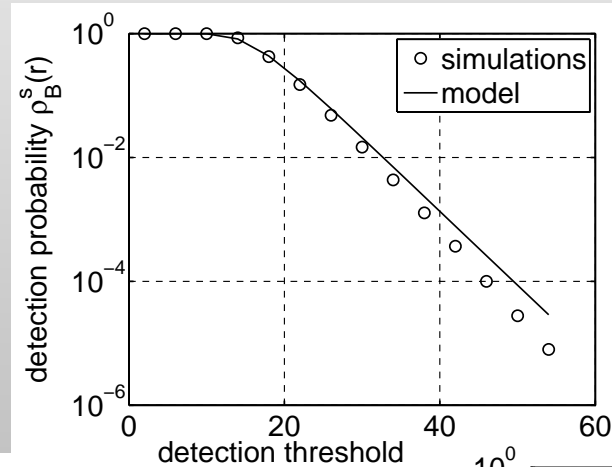
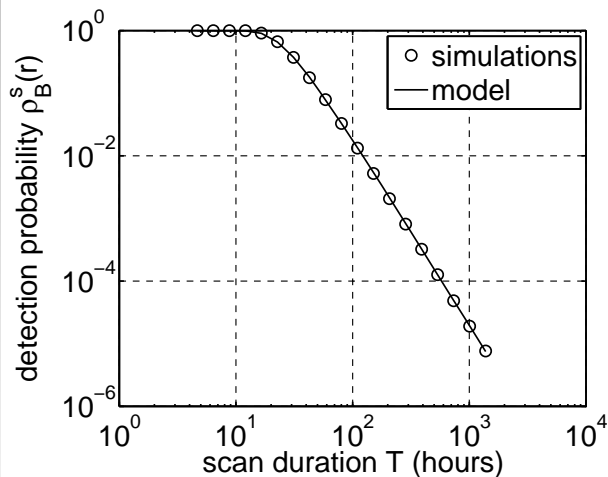
$$\rho(T) \approx 1 - e^{-(|s|-E+1)(1-\chi)\chi^{E-1}}$$

- where  $\chi = 1 - (1 - q)^{|s|}$

- as long as  $\frac{(|s| - E)(1 - \chi)}{m} \gg 1$

# Analysis of Existing Methods 6

- Even for small subnets ( $|s| = 2^8$ ), model is quite accurate, except when threshold  $E$  is large



# Analysis of Existing Methods 7

- We invert both IDS-A/B models to obtain stealth cover time (SCT)  $\delta$

- After simplifications and approximations for  $\epsilon \rightarrow 0$ :

$$\delta \approx \frac{|s|^{1+c} \Delta}{\omega \gamma \epsilon^c}$$

- where

$$c = \frac{1}{E-1} \quad \text{and} \quad \gamma = \begin{cases} (E!)^c & \text{IDS-A} \\ 1 & \text{IDS-B} \end{cases}$$

- Observations

- Compared to IDS-A, scans against IDS-B must be slower by a factor of  $(E!)^c$  (rather than  $E-1$  as for sequential) for the same probability of detection

# Analysis of Existing Methods 8

- Pre-permutation split ( $\omega = 1$ ) does not improve scan time with botnet size  $m$ ; post-permutation benefits linearly
- SCT scales **super-linearly**  $\sim |s|^{1+c}$  with subnet size
  - In fact, for  $E = 2$  ( $c = 1$ ), this rate is **quadratic**
  - This means that sometimes sequential is less detectable than uniform for the same scan rate!
  - Specifically, sequential is more stealthy in subnets of size

$$|s| > \left( \frac{n\gamma\epsilon^c}{\zeta} \right)^{\frac{E-1}{E}}$$

for  $E = 2$  and  $\epsilon = 10^{-3}$ , this means all /20 and larger networks

- Uniform has optimal **average** scanning rate
  - But on small timescales, it can be bursty

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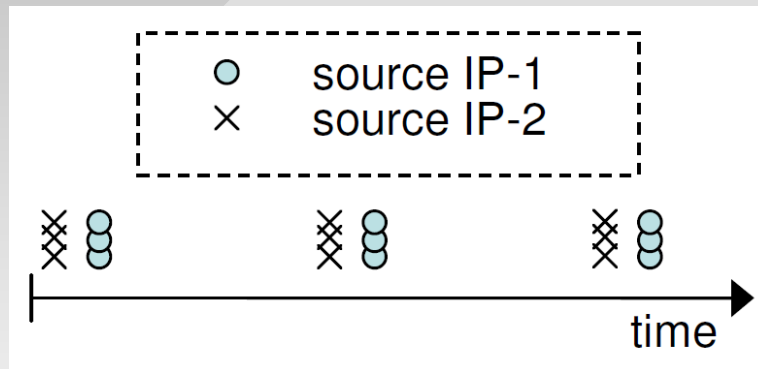
# Stealth Optimality

- Analysis above begs a few questions
  - Can lower SCT be achieved?
  - What is the stealthiest possible scan pattern?
  - Can both IDS-A and IDS-B be scanned with equal detection rates?
- Our solution is a new scan method we call **STealth-OPtimal (STOP)** that consists of 3 elements
  - A new **permutation** that delivers packets to all subnets maximally spaced apart (see paper)
  - A novel **split** that guarantees optimal spacing across multiple botnet IPs (see paper)
  - A new **schedule** that makes evading IDS-B as easy as IDS-A (briefly covered next)

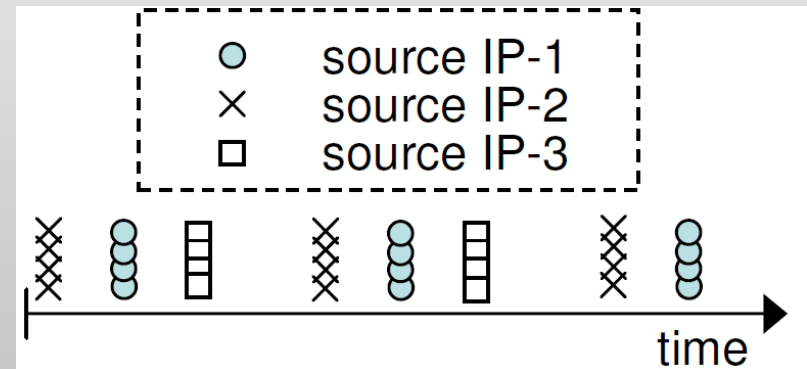


## Stealth Optimality 2

- STOP pattern seen at each subnet
  - Raises counter to  $E-1$ , then delays the next burst by  $\Delta$



$$m = 2, E = 4$$



$$m = 3, E = 5$$

- Instead of one packet per  $\Delta$  window, STOP can scan IDS-B (and similarly IDS-A) with  $E-1$  packets per window without detection

## Stealth Optimality 3

- Requires knowledge of some lower bound  $\beta$  on  $E$ 
  - For example, no mainstream IDS utilizes  $E$  less than 4
  - Some have  $E$  between 20-200 (Bro, NIKSUN, Juniper)
  - The larger this lower bound  $\beta$ , the better STOP's performance compared to prior methods
- STOP provably achieves the lowest possible SCT against both IDS-A and IDS-B:

$$\delta = \frac{|s|\Delta}{m(\beta - 1)}$$

- Linear in all parameters  $m, |s|, \beta-1, \Delta$
- How does this compare to existing methods?

## Stealth Optimality 4

- Compared to **sequential** (/16 subnets,  $\beta=4$ )
  - STOP can scan 64K times faster against IDS-A and 196K times faster against IDS-B
  - This translates into a reduction of total scan duration  $T$  from 1 year to 8 and 2.6 minutes, respectively
- Compared to **uniform** (/16 subnets,  $\beta=4$ ,  $\epsilon = 10^{-3}$ )
  - STOP is 419 times faster against IDS-A and 1209 times faster against IDS-B
  - Reduction in  $T$  from 1 year to 21 and 7 hours, respectively
- Many more results and comparisons in the paper

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# Final Thoughts

- Linear increase in stealth with  $m$  is quite peculiar
  - Suggests that hijacking unused IPs on the subnet can significantly benefit viruses
  - Aliasing  $k$  IPs to the same NIC allows the host to become  $k$  times stealthier in terms of SCT
  - Extra steps needed are detection of NAT and DHCP conflicts with existing hosts, but both are doable
- Methods to improve IDS?
  - While tweaking  $E$  and  $\Delta$  is possible, this may lead to increased false-positive rates
  - Future work will address design of new algorithms for better IDS window maintenance