# Hierarchical Heavy Hitter

TANG DING

### **Papers**

 Mitzenmacher, M., Steinke, T., & Thaler, J. Hierarchical heavy hitters with the space saving algorithm. (ALENEX 2012)

 Ben Basat, R., Einziger, G., Friedman, R., Luizelli, M. C., & Waisbard, E. Constant time updates in hierarchical heavy hitters. (SIGCOMM 2017)

## **Application**

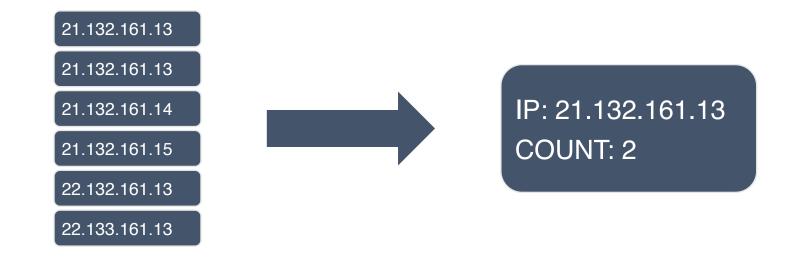
- Network traffic monitoring
- Anomaly detection
- DDoS detection.

## Heavy Hitter

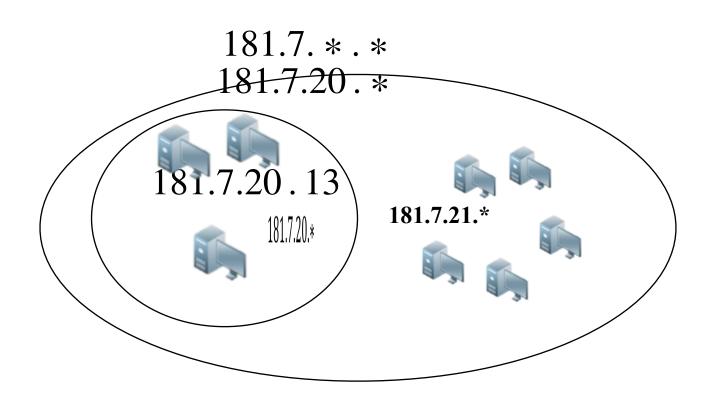
- Basically, equals to frequent items
- Identifies frequent:
  - Source.
  - Source-Destination pairs.

## Example

We want to find Source IP whose frequency >= 2



## Hierarchical Heavy Hitter



220.7.16.\*

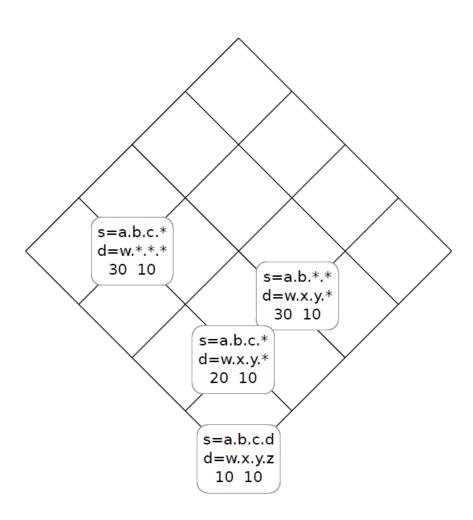


Destination

## 1D Hierarchy example

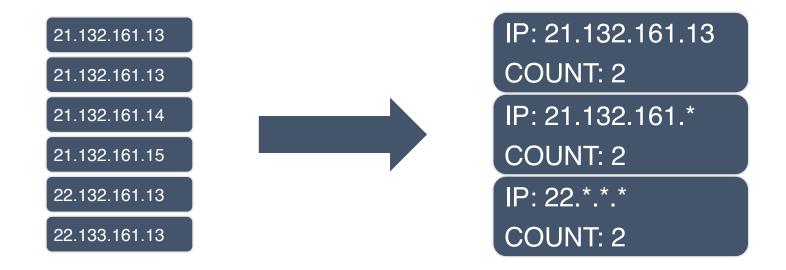
- Level4: \*.\*.\*.\*
- Level3: 21.\*.\*.\*
- Level2: 21.132.\*.\*
- Level1: 21.132.145.\*, 21.132.146.\*
- Level0: 21.132.145.13, 21.132.146.14

## 2D Hierarchy example



## Example

We want to find HHH whose frequency >= 2



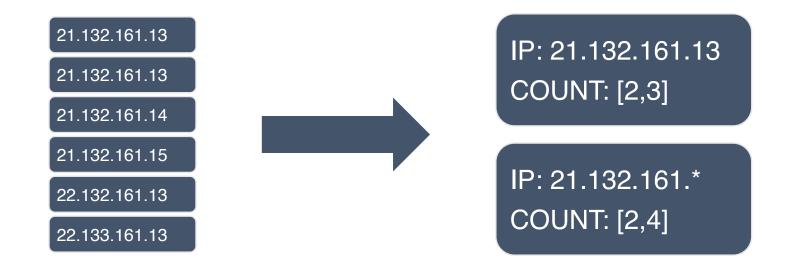
Why count of 21.132.161.\* is 2?

#### **Problems**

- Often, the large volume of network traffic makes it infeasible to store the relevant data in memory.
- Space-saving streaming algorithm.

## Example

We want to find HHH whose frequency >= 2



Output confidence interval instead of actual frequency

## **Target**

- Accuracy: Limited Errors
- Coverage: No False Negatives

## **Target**

- Given parameter, we want to output a set of items, and lower and upper bounds of frequency and, such that they satisfy two properties, as follows:
  - 1. Accuracy., and for all.
  - 2. Coverage. For set, . Define the conditioned count of with respect to to be. We require for all prefixes

## Algorithm

- Two main procedures:
  - Update
  - Output

## Update

Time Step	Update	Counter 1	Counter 2	Counter 3
0		unused	unused	unused
1	(a,+2)	(a,2)	unused	unused
2	(b,+6)	(a,2)	(b,6)	unused
3	(c,+4)	(a,2)	(b,6)	(c,4)
4	(a,+3)	(a,5)	(b,6)	(c,4)
5	(d,+4)	(a,5)	(b,6)	(d,8)
6	(e,+4)	(e,9)	( <b>b</b> , <b>6</b> )	(d,8)

For any item C being tracked, the actual frequency f(C) is in (Nc – Nmin, Nc),

where Nc is the count of that item, and Nmin is the minimum count in all counters

## Update

Time T	Counter 1	Counter 2	Counter 3
Level 1	(a.*.*.*, 20)	(h.*.*.*, 12)	(q.*.*.*, 16)
Level 2	(a.b.*.*, 18)	(h.i.*.*, 12)	(q.r.*.*, 16)
Level 3	(a.b.c.*, 18)	(q.r.s.*, 9)	(h.i.j.*, 14)
Level 4	(a.b.c.d, 10)	(h.i.m.n, 15)	(a.b.c.e, 8)

(w.x.y.z, +3)

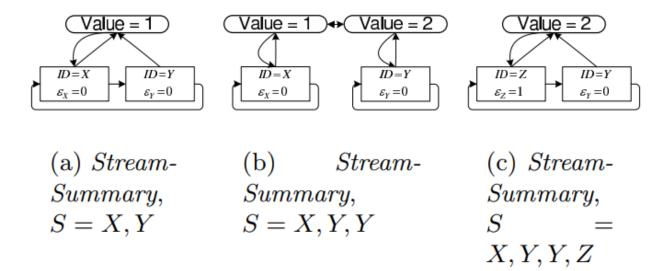
Time T+1	Counter 1	Counter 2	Counter 3
Level 1	(a.*.*.*, 20)	(w.*.*.*, 15)	(q.*.*.*, 16)
Level 2	(a.b.*.*, 18)	(w.x.*.*, 15)	(q.r.*.*, 16)
Level 3	(a.b.c.*, 18)	(w.x.y.*, 12)	(h.i.j.*, 14)
Level 4	(a.b.c.d, 10)	(h.i.m.n, 15)	(w.x.y.z, 11)

## Update

- For each level, set counters
- Three procedures
  - Find counter for certain IP (pairs)
  - Find counter with minimum count
  - Update counter found (item / count)
- Update Complexity
  - Arbitrary Update: Using heaps –
  - Unitary Update:
  - is the number of levels

## **Unitary Update**

- Data Structure
  - HashTable
  - Item{HashEntry, GroupEntry, NextItemInGroup, PreviousItemInGroup}
  - Group {Items, Count, PreviousGroup, NextGroup}
    - · Sorted link list with entries of same count combined



## Output (for 1D)

```
OUTPUTHHH1D(threshold \phi)
1 /* par(e) is parent of e^*/
2 Let s_e = 0 for all e
3 / *s_e conservatively estimates the difference
    between unconditioned and conditioned counts of e^*
   for each e in postorder
         (f_{\min}(e), f_{\max}(e)) = \text{GetEstimateSS}(SS(n), e)
5
         if f_{\max}(e) - s_e \ge \phi N
6
               print(e, f_{min}(e), f_{max}(e))
               s_{\text{par}(e)} + = f_{\min}(e)
9
         else s_{par(e)} + = s_e
```

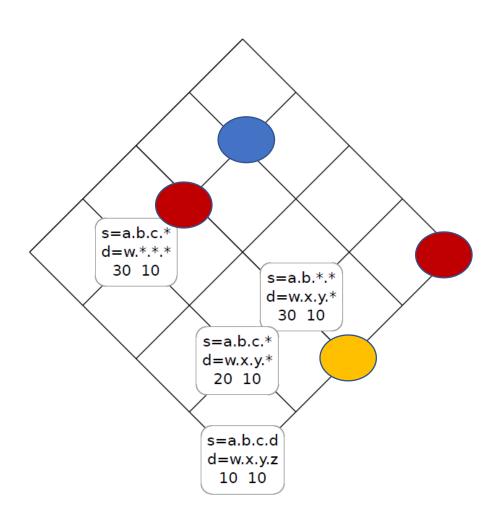
### Proof

- By using space (counters in each level), we can achieve our target in both accuracy and coverage
  - Accuracy:
    - if ,
    - =
  - Coverage

## Output (for 2D)

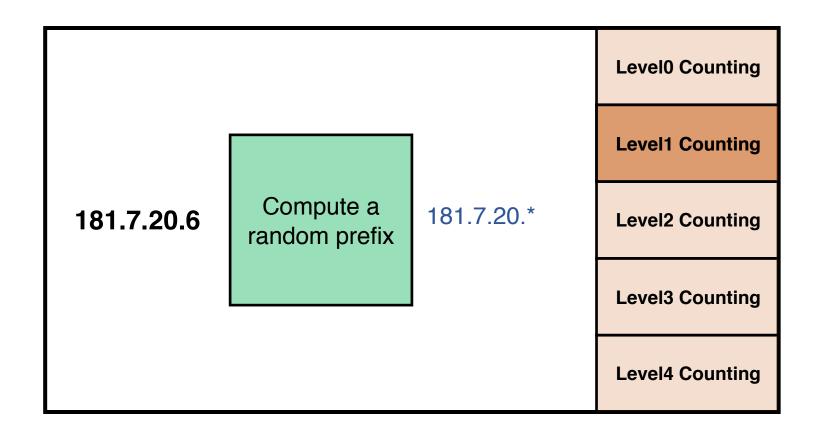
```
OUTPUTHHH2D(threshold \phi)
     P = \emptyset
     for level l=L downto 0
 3
            for each item p at level l
                  Let n be the lattice node that p belongs to
 5
                  (f_{\min}(p), f_{\max}(p)) = \text{GetEstimateSS}(SS(n), p)
 6
                  F_p' = f_{\text{max}}(p)
                  \hat{H}_p = \{ h \in P \text{ such that } \nexists h' \in P : h \prec h' \prec p \}
 8
                  for each h \in H_p
                        F_p' = F_p' - f_{\min}(h)
 9
                  for each pair of distinct elements h, h' in H_p
10
                        q = \text{glb}(h, h')
11
                        if \nexists h_3 \neq h, h' in H_p s.t. q \leq h_3
12
                               F_p' = F_p' + f_{\max}(q)
13
                  if F_p' \geq \phi N
14
                        P = P \cup \{p\}
15
                        print(p, f_{min}(p), f_{max}(p))
16
```

## Output



### Randomized

Select a prefix at random and count it



## Additional Speedup

188.3.12.3 181.7.20.3 92.67.7.81 Compute a With 181.7.20.\* 181.7.20.6 random some probability 188.67.7.1 prefix 181.7.20.2 Ignore packet

**Level0 Counting** 

**Level1 Counting** 

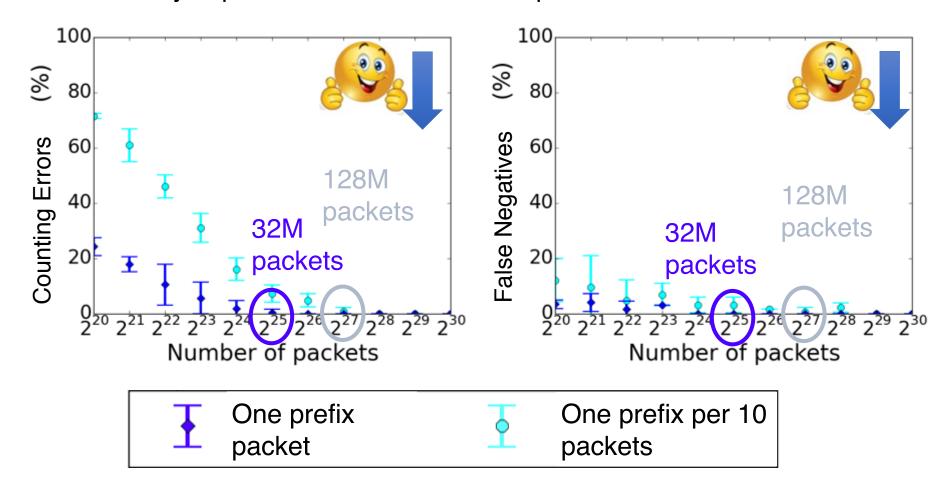
**Level2 Counting** 

**Level3 Counting** 

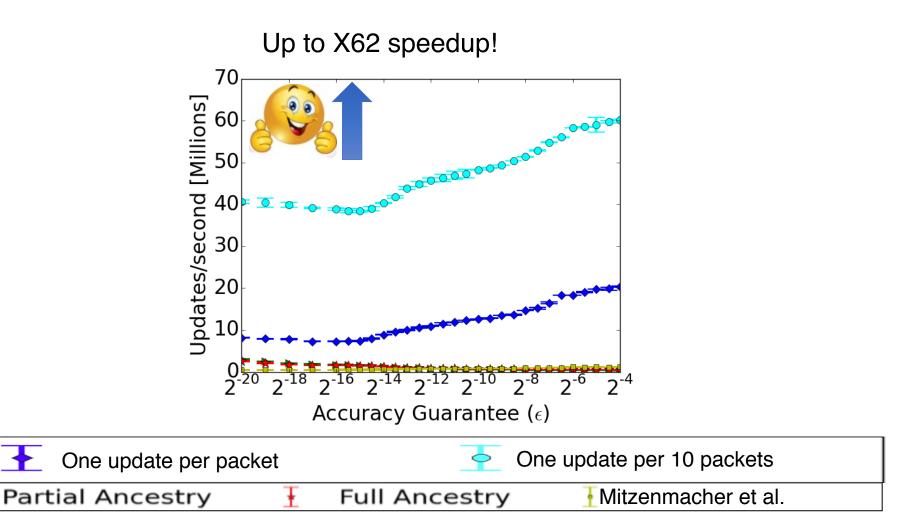
**Level4 Counting** 

#### Accuracy

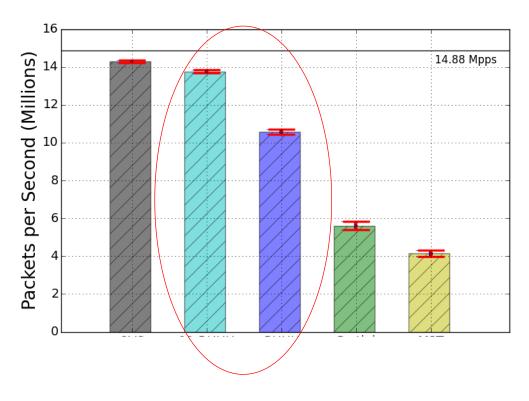
"Accuracy improves with the number of packets"



### Performance



### Performance



#### **Highlights:**

- Only -4% overheads for HHH in the Open vSwitch data plane!
- +250% throughput improvement compared to previous work.

