RFID Cardinality Estimation with Blocker Tags

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Outline

1 Background & Motivation

2 Problem Formulation

3 <u>R</u>FID <u>E</u>stimation scheme with <u>B</u>locker tags (REB)

4 Theoretical Analysis

5 Performance Evaluation

6 Conclusion

- Radio Frequency Identification.
- An identification system that consists of chip-based tags, readers, and a back-end.
- Each tag has a unique 96-bit ID to identify the tagged object.



RFID Background

Two types of RFID tags:
Passive tags and Active tags















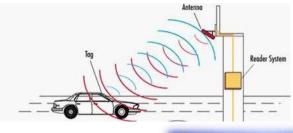


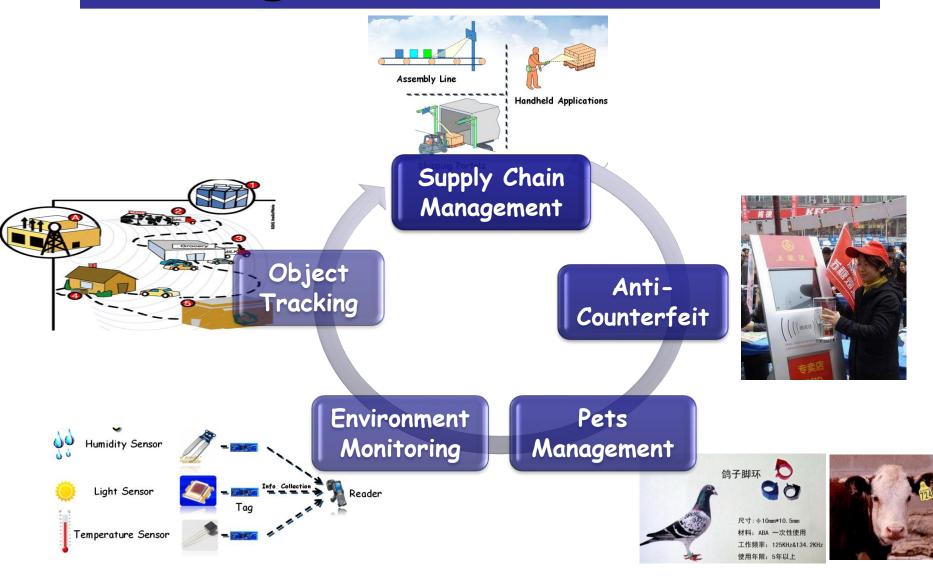


RFID Background



- Advantages of RFID over bar-code:
 - remote access
 non-line-of-sight reading
 multiple simultaneous accesses
 large rewritable memory





- The widely-used RFID tags impose serious privacy concerns.
- Reason: When C1G2 tags are interrogated by an RFID reader, no matter whether the reader is authorized or not, they blindly respond with their IDs and other stored information (such as manufacturer, product type, and price) in a broadcast fashion.

- What woman wants her dress size to be publicly readable by any nearby scanner?
- Who wants the medications and other contents of a purse to be scannable?
- Who wants his or her location to be tracked and recorded based on the unique ID number in their shoes or other clothing?
- An effective solution to this privacy issue is to use commercially available blocker tags.

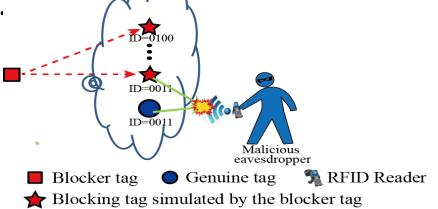


• What are blocker tags?

A blocker tag is an RFID device that is preconfigured with a set of known RFID tag IDs, which we call blocking IDs. The blocker tag behaves as if all tags with its blocking IDs are present.

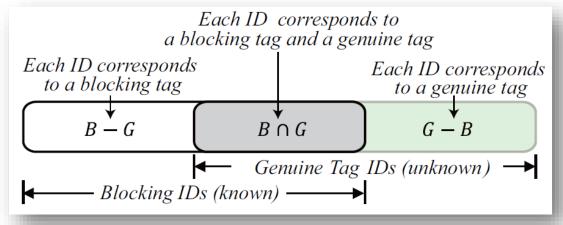
- How blocker tags protect the privacy?
- A blocker tag protects the privacy of the set of genuine tags whose IDs are among the blocking IDs of the blocker tag because any response from a genuine tag is coupled with the simultaneous response from the blocker tag; thus, the two responses always collide and attackers cannot obtain private information.

The genuine tag <mark>always</mark> collides with the blocking tag having the same ID



Problem Formulation

- We are concerned with the problem of RFID (population size) estimation with the presence of blocker tags.
- **Problem Definition:** given (1) a set of unknown genuine tags *G* of unknown size *g*, (2) a blocker tag with a set of known blocking IDs *B*, (3) a required confidence interval $\alpha \in (0.1]$, and a required reliability $\beta \in [0,1)$, we want to use one or more readers to estimate the number of genuine tags in *G*, denoted as \hat{g} , so that $P\{|\hat{g} - g| \leq g\alpha\} \geq \beta$



Problem Formulation

- To the best of our knowledge, this paper is the first to investigate RFID estimation with the presence of a blocker tag.
- None of the existing estimation schemes considers the presence of a blocker tag. Furthermore, none of them can be easily adapted to solve this problem.

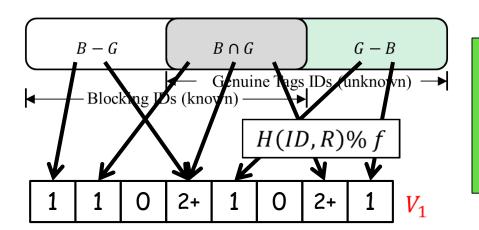
Problem Formulation

- How about turning off the blocker tag and then using prior RFID estimation schemes to estimate the number of genuine tags?
- Turning off the blocker tag will give attackers a time window to breach privacy, especially for the scenarios in which RFID estimation schemes are being continuously performed for monitoring purposes.

- <u>R</u>FID <u>E</u>stimation scheme with <u>B</u>locker tags
- The communication protocol used by REB is the standard *framed slotted Aloha* protocol.

- Detailed Steps:
- **Step1**: the reader broadcasts a value *f* and a random number *R* to query all tags (including blocker tags), where *f* is the number of slots in the forthcoming frame. Then, each tag computes a hash *H*(*ID*,*R*)%*f* to select a slot to respond.

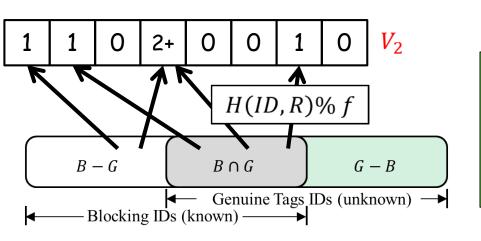
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- O represents no tag responds
- 1 represents only one tag responds
- 2+ represents two or more tags simultaneously respond and create a collision

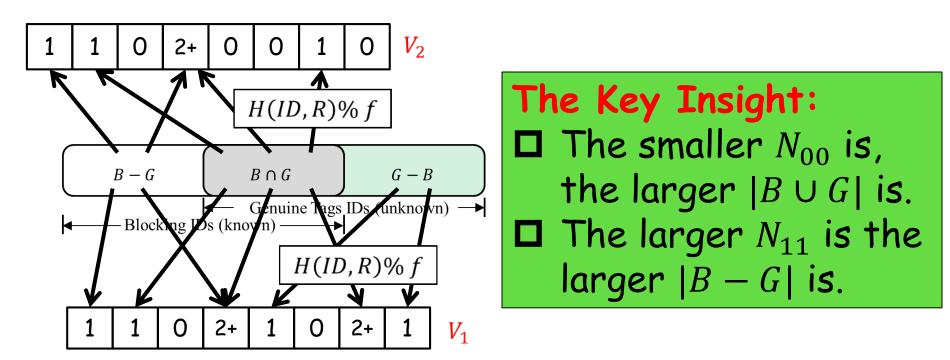
 Step2: As we know the blocking IDs, we can virtually execute the framed slotted Aloha protocol using the same frame size f and random number R for the blocking IDs; thus, we get another vector.

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- O represents no tag chooses this slot.
- 1 represents only one tag chooses this slot.
- 2+ represents two or more tags choose this common slot.

• **Step3**: we count two numbers: N_{00} , which is the number of slot *i* such that both $V_1[i] = 0$ and $V_2[i] = 0$, and N_{11} , which is the number of slots *i* such that both $V_1[i] = 1$ and $V_2[i] = 1$.



- We theoretically proved that N_{00} monotonously decreases with the increase of $|B \cup G|$; and N_{11} monotonously increases with the increase of |B G|.
- Therefore, from the observed values of N_{00} and N_{11} , we can estimate $|B \cup G|$ and |B G|, respectively. Then, we can calculate the number of genuine tags, i.e., $|G| = |B \cup G| |B G|$.

- **Practical Issue:** The frame size should be set as no more than 512. To scale to a large tag population, the reader uses a persistence probability $p \in (0, 1]$ to virtually extend the frame size f to f/p, but actually terminates the frame after the first f slots.
- Fundamentally, each tag participates in the actual frame of f slots with a probability p.

- Functional Estimator:
- $\hat{g} = -\frac{f}{p} ln \left(\frac{N_{00}}{f} \right) \frac{fN_{11}}{pN_{00}}$, where f is the observed frame size, p is the persistence probability, N_{00} is the number of persistent empty slots, N_{11} is the number of persistent singleton slots.

• Variance of the Estimator:

• $Var(\hat{g}) = \frac{1}{fp^2} e^{\frac{up}{f}} (b'^2 p^2 + f^2 - b'fp) - \frac{f}{p^2}$, where f is the observed frame size, p is the persistence probability, $u = |B \cup G|$, and b' = |B - G|.

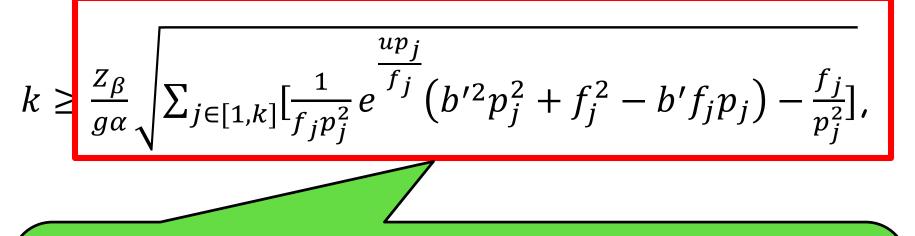
- Refined Estimation with k Frames:
- We repeat k independent frames with different seeds, and use the average estimation result $\widehat{g_k}' = \frac{1}{k} \sum_{j \in [1,k]} \widehat{g_j}$ to refine the estimation of REB, where $\widehat{g_j}$ is the estimate derived from the *j*-th frame.

- Termination Condition:
- If the frame number k satisfies: $k \ge k$

$$\frac{Z_{\beta}}{g\alpha} \sqrt{\sum_{j \in [1,k]} \left[\frac{1}{f_j p_j^2} e^{\frac{up_j}{f_j}} \left(b'^2 p_j^2 + f_j^2 - b' f_j p_j\right) - \frac{f_j}{p_j^2}}\right]},$$

where f_j and p_j are the frame size and persistence probability used in the *j*-th frame.

Avoiding Premature Termination:

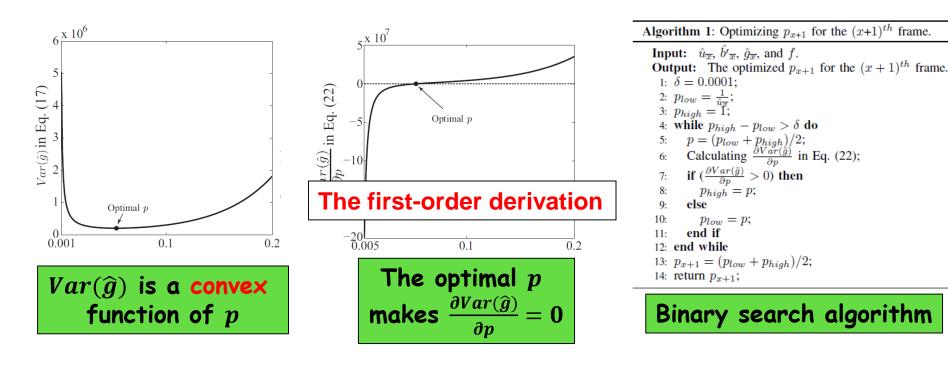


If we directly use the estimated values $\hat{b}', \hat{u}, \hat{g}$ to calculate the R.H.S. of this inequality, k may have a chance to be larger than it, which is not true and REB will have a premature termination.

- δ -sigma method to avoid premature termination.
- When calculating the R.H.S. of the termination inequality, we use the upper/lower bounds on b', u, g.
- Upper bounds: $\hat{x} \uparrow = \hat{x} + \delta \sqrt{Var(\hat{x})}$;
- Lower bounds: $\hat{x} \downarrow = \hat{x} \delta \sqrt{Var(\hat{x})}$,
- Here, x could be b', u, or g.
- Three-sigma rule indicates $\delta = 3$ is large enough.

- Optimization: frame size f and persistence probability p.
- For the first frame, we simply set f = 512 and $p = \frac{512}{\hat{u}}$, where \hat{u} is the number of total tags that can be fast estimated by the existing estimation protocols, e.g., ART [Mobicom 12].
- For the other frames, we can leverage the information obtained from previous frames to optimize f and p.

- Optimization: the Persistence Probability p
- For a fixed frame size f, the goal of optimizing p is to minimize the estimation variance $Var(\hat{g})$.



- Optimization: the frame size f
- We target finding an optimal f to minimize the expected remaining execution time.

The remaining

• Minimize
$$(f + 1) \times y$$
 execution time

• **s.t.**
$$x + y \ge \frac{Z_{\beta}}{g\alpha} \sqrt{\sum_{j \in [1,x]} Var(\widehat{g_j}) + yVar(\widehat{g})}$$

•
$$f \in \{2,4,8,16,\dots,512\}$$

• Here, x is the number of frames that have already been executed. y is the number of frames that need to be further executed.

• 1. Verifying the Optimized f and p.

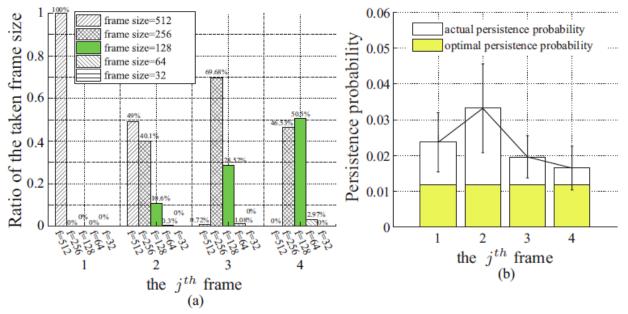


Fig. 3. Verifying the optimized settings of f and p. |B - G| = 5000, $|B \cap G| = 5000$, |G - B| = 5000. $\alpha = 10\%$, $\beta = 90\%$. (a) Verifying the optimized f. (b) Verifying the optimized p.

The values of f and p approach their overall optimal values after a few frames.

• 2. Estimation Reliability.

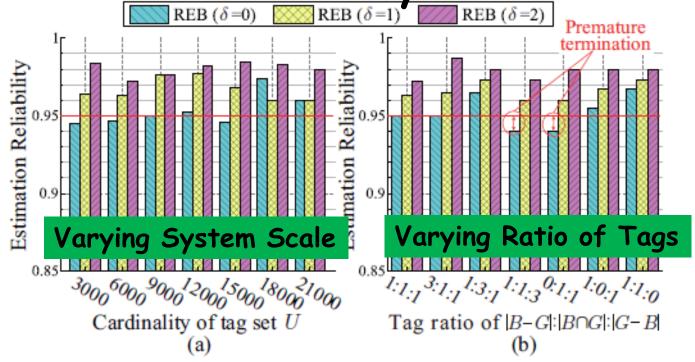


Fig. 4. Evaluating the reliability of REB. $\alpha = 5\%$, $\beta = 95\%$. (a) Tag ratio |B - G|: $|B \cap G|$: |G - B| is fixed to 1 : 1 : 1, and u varies from 3000 to 21000. (b) u is fixed to 9000, and tag ratio varies.

Our REB ($\delta = 1$) can meet the required accuracy under different simulation settings

• 3. Time Efficiency: Impact of |U|

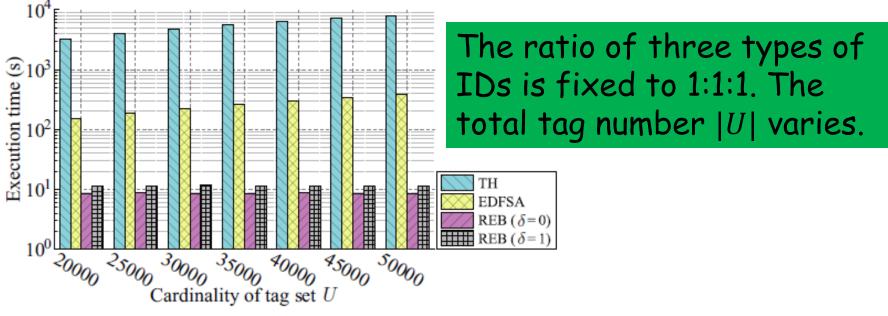
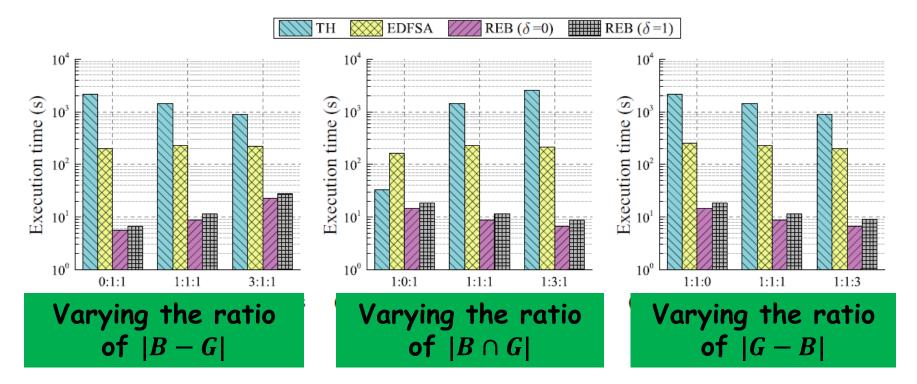


Fig. 5. Evaluating the time-efficiency of protocols with varying u. Tag ratio of |B - G|: $|B \cap G|$: |G - B| is fixed to 1:1:1 and $\alpha = 5\%$, $\beta = 95\%$.

When |U|=50000, our REB runs 33x faster than the fastest tag identification protocol.

• 4. Time Efficiency: Impact of Tag Ratio



Our REB persistently runs tens of times faster than the existing protocols.

Conclusion

- We take the first step to address the problem of RFID estimation with Blocker tags.
- The proposed REB protocol is compliant with the commodity EPC C1G2 standard, and does not require any modifications to off the-shelf RFID tags.
- REB can guarantee any degree of estimation accuracy specified by the users.
- Extensive simulation results reveal that REB is tens of times faster than the fastest identification protocol with the same accuracy requirement.

Thanks for your attention!

Q & A