# Bounds for Group Testing Algorithms to Determine the Number of Defectives 

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## Group Testing



Robert Dorfman's paper in 1943 introduced the field of (Combinatorial) Group Testing. The motivation arose during the Second World War when the United States Public Health Service and the Selective service embarked upon a large scale project. The objective was to weed out all syphilitic men called up for induction. However, syphilis testing back then was expensive and testing every soldier individually would have been very cost heavy and inefficient.

We can combine blood samples and test a combined sample together to check if at least one soldier has syphilis.


Positive
1101
$0110 \quad 1$

0101
$x_{1}, x_{2}, x_{3}, x_{4}$, $x_{5}, x_{6}, x_{7}, x_{8}$, $x_{9}, x_{10}, x_{11}, x_{12}$ $x_{13}, x_{14}, x_{15}, x_{16}$

Detecting the Defective items
Determining the Number of Defective items
Estimate the Number of Defective items

1010
Negative

$$
0101
$$



1010
0101

$$
f=x_{3} \vee x_{10}
$$



Positive
0010
1010
0001

## Many other applications

Soldiers $\Rightarrow$ Items
Sick soldiers $\Rightarrow$ defective items
$n$ \#items
d \#defective items


## Detecting the Defective items

| Algorithm | Lower Bound | Upper Bound |
| :---: | :---: | :---: |
| Schlaghoff and Triesch 2005 and Cheng et al. 2014, 2015 and many others |  |  |
| Adaptive Deterministic or Randomized (poly) |  |  |
| Folklore Result |  |  |
| Non-Adaptive Randomized MC (poly) |  |  |
| A.G. D'yachkov, V.V. Rykov 1982 Porat and Rothschild 2011 |  |  |
| Non-Adaptive Deterministic (poly) |  |  |

A. De Bonis, L. Gasieniec, U. Vaccaro 2005

Two-round Deterministic (exp)

## M. Cheraghchi 2013

Two-round Deterministic (poly)

$$
n \text { tests } \rightarrow d \log n-d^{2} \log n \text { tests }
$$

## Determining the Number of Defective items



## Determining the Number of Defective items



## Monte Carlo Adaptive - Upper Bound

0

$d \leq Y \leq 2 d$ with failure probability at most $\frac{\delta}{2}$ - can be done in $d+\log d \log 1 / \delta$

$$
\begin{aligned}
r=\frac{Y^{2}}{\delta} \text { sets } \quad \operatorname{Pr}[\text { Fail }] & =1-\left(1-\frac{1}{r}\right)\left(1-\frac{2}{r}\right) \cdots\left(1-\frac{d-1}{r}\right) \\
& \leq \frac{1}{r}+\frac{2}{r}+\cdots+\frac{d-1}{r} \leq \frac{d^{2}}{2 r} \leq \frac{\delta}{2}
\end{aligned}
$$

Schlaghoff and Triesch 2005 and Cheng et al. 2014, 2015

$$
d \log \frac{n}{d}+O(d)
$$

$$
d \log \frac{r}{d}=d \log \frac{Y^{2}}{d \delta}=d \log \frac{4 d}{\delta}=d \log \frac{d}{\delta}+O(d)
$$



## Estimating the Number of Defective items

$$
d / \Delta \leq x \leq \Delta d \quad \Delta=1+\Omega(1)
$$

Algorithm
Lower Bound
Upper Bound
Damaschke and Sheikh Muhammad 2010
Non-Adaptive MC Rand.

## Bshouty 2019

Non-Adaptive MC Rand.

## Bshouty 2019

Non-Adaptive Deterministic or LV Rand.

## Estimating the Number of Defective items

Given some upper bound $d \leq \boldsymbol{D}$

$$
d / \Delta \leq x \leq \Delta d
$$

| Algorithm | Lower Bound | Upper Bound |
| :---: | :---: | :---: |
| Damaschke and Sheikh Muhammad 2010 |  |  |
| Non-Adaptive MC Rand. |  |  |

## Bshouty 2019

Non-Adaptive MC Rand.

## Deterministic Estimating the Number of Defective items

## Bshouty Haddad-Zaknoon $2020 \quad$ Given some upper bound $d \leq \boldsymbol{D} \quad d / \Delta \leq x \leq \Delta d$

| Bounds | Adaptive/ <br> Non-Adapt. | Result | Explicit/ <br> Non-Expl. | Ref. |
| :---: | :---: | :---: | :---: | :---: |
| Lower B. | Non-Adapt. | $\frac{D}{\Delta^{2}} \log \frac{n}{D}$ | - | $[1]$ |
| Lower B. | Adaptive | $\frac{D}{\Delta^{2}} \log \frac{n}{D}$ | - | Ours |
| Upper B. | Adaptive | $\frac{D}{\Delta^{2}}\left(\log \frac{n}{D}+\log \Delta\right)$ | Explicit | Ours |
| Upper B. | Non-Adapt. | $\frac{\log D}{\log \Delta} D \log n$ | Non-Expl. | $[1]$ |
| Upper B. | Non-Adapt. | $\frac{D}{\Delta^{2}}\left(\log \frac{n}{D}+\log \Delta\right)$ | Non-Expl. | Ours |
| Upper B. | Non-Adapt. | $\frac{D^{1+o(1)}}{\Delta^{2}} \log n$ | Explicit ${ }^{1}$ | Ours |
| Upper B. | Non-Adapt. | $\frac{D}{\Delta^{2}} \cdot$ Quazipoly $(\log n)$ | Explicit | Ours |

[1] Abhishek Agarwal, Larkin Flodin, and Arya Mazumdar. Estimation of sparsity via simple measurements. In 2017 IEEE International Symposium on Information Theory (ISIT), pages 456-460. IEEE, 2017.

## Estimate the Number of Defective items

$$
(1-\epsilon) d \leq Y \leq(1+\epsilon) d
$$

| Algorithm | Lower Bound | Upper Bound |
| :--- | :--- | :--- |
| Cheng and Xu (2014) |  |  |
| Adaptive Rand. MC with |  |  |
|  | Ron and Tsur (2014) |  |
| Adaptive Rand. MC with |  |  |

## Falahatgar et al. (2016)

Adaptive Randomized MC

+ Expected number queries

Adaptive Randomized MC with

## Ron and Tsur (2014)

Adaptive Randomized MC with

## Falahatgar et al. (2016)

Adaptive Randomized MC +
Expected number of queries

| Bshouty, Bshouty-Hurani, Haddad, Hashem, Khoury, Sharafy 2018 |  |  |
| :--- | :--- | :--- |
| Adaptive Deterministic and Las |  |  |
| Vegas |  |  |
| Adaptive Randomized MC + |  |  |
| Expected number of queries |  |  |
| Adaptive Randomized MC |  |  |

Adaptive Randomized MC with

## Ron and Tsur (2014)

Adaptive Randomized MC with

## Falahatgar et al. (2016)

Adaptive Randomized MC +
Expected number of queries

| Bshouty, Bshouty-Hurani, Haddad, Hashem, Khoury, Sharafy 2018 |  |  |
| :--- | :--- | :--- |
| Adaptive Deterministic and Las |  |  |
| Vegas |  |  |
| Adaptive Randomized MC + |  |  |
| Expected number of queries |  |  |
| Adaptive Randomized MC |  |  |

## Deterministic Adaptive - Upper Bound



Schlaghoff and Triesch 2005 and Cheng et al. 2014, 2015

$$
d \log \frac{n}{d}+O(d)
$$

$$
\begin{array}{ccc}
d \log \frac{(1-\epsilon) n}{d} & & \frac{Y}{1-\epsilon} \geq d \\
Y & Y \leq d & Y \geq(1-\epsilon) d
\end{array}
$$

$$
(1-\epsilon) d \leq Y \leq d
$$

Adaptive Randomized MC with

## Ron and Tsur (2014)

Adaptive Randomized MC with

## Falahatgar et al. (2016)

Adaptive Randomized MC +
Expected number of queries

| Bshouty, Bshouty-Hurani, Haddad, Hashem, Khoury, Sharafy 2018 |  |  |
| :--- | :--- | :--- |
| Adaptive Deterministic and Las |  |  |
| Vegas |  |  |
| Adaptive Randomized MC + |  |  |
| Expected number of queries |  |  |
| Adaptive Randomized MC |  |  |

## Adaptive Deterministic and Las Vegas - Lower Bound



$$
\text { Expect }=d \log \frac{(1-\epsilon) n}{d}
$$

Yao MinMax Theorem


## OPEN PROBLEMS

| Algorithm | Lower Bound | Upper Bound |
| :--- | :--- | :--- | :--- | :--- |
| Randomized MC + |  |  |
| Expected number of queries |  |  |

## Thank You

## Thank You

| Algorithm | Lower Bound | Upper Bound |  |
| :--- | :--- | :--- | :--- |
| Randomized MC with | Cheng and Xu (2014) |  |  |
| Randomized MC with |  |  |  |

Monte Carlo adaptive algorithm

$$
(1-\epsilon) d \leq D \leq(1+\epsilon) d
$$



| Algorithm | Lower Bound | Upper Bound |  |
| :--- | :--- | :--- | :--- |
| Randomized MC with | Cheng and Xu (2014) |  |  |
| Randomized MC with |  |  |  |



$$
a_{i}=2^{i}
$$

Falahatgar et al. (2016) $\quad a_{i}=2^{2^{i}} \quad d^{\frac{1}{2}} \leq D_{1}, D_{2}<d^{2} \quad b_{1}=\left(D_{1} D_{2}\right)^{1 / 2}$
$\log \log d$
$\log \log d$
$2 \log \log d$

Ours

$$
a_{i}=2^{2^{i^{2}}}
$$

$d^{2^{-2} \sqrt{\text { log log }}} \leq D_{1}, D_{2}<d^{2^{2 \sqrt{\log \log d}}}$
$b_{1}=\left(D_{1} D_{2}\right)^{1 / 2}$
$\sqrt{\log \log d}$
$\log \log d+\sqrt{\log \log d}$
$\log \log d$


$$
\log \log d+O\left(\frac{1}{\epsilon^{2}} \log \frac{1}{\delta}\right)
$$

$$
(1-\delta) \log \log d+O\left(\frac{1}{\epsilon^{2}} \log \frac{1}{\delta}\right)
$$

