



UNIVERSITÄT
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Sublinear Algorithms

Lecture 04: Measurements I



Last 3 lectures:
Streaming Algorithms

Main requirement: Maintain *sublinear* space
Often of interest: fast query and update time

Next 3 lectures:
Measurements

Access an object using queries/measurements,
e.g. linear combinations of its components

Main requirement:
Achieve *sublinear* measurement complexity

Sparse Recovery

Recovery of sparse structures using queries from a restricted class

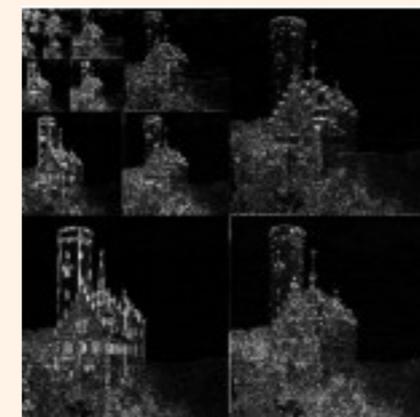
Principle: Often 1% of an object carries 99% of the information



Image from Hubble telescope



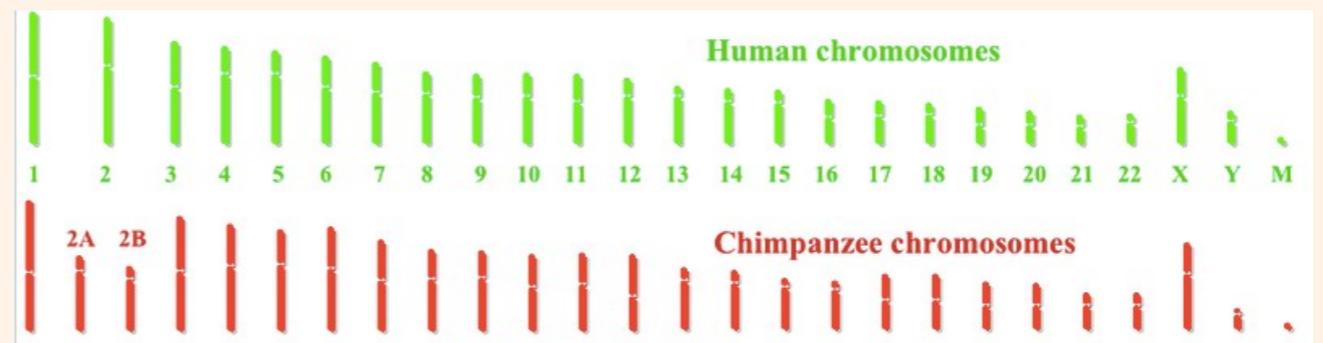
Dense Image



Sparse Transformation of a Dense Image



*Audio signals
are sparse*



*Difference between human and chimpanzee DNA
is a sparse vector*

Combinatorial Group Testing

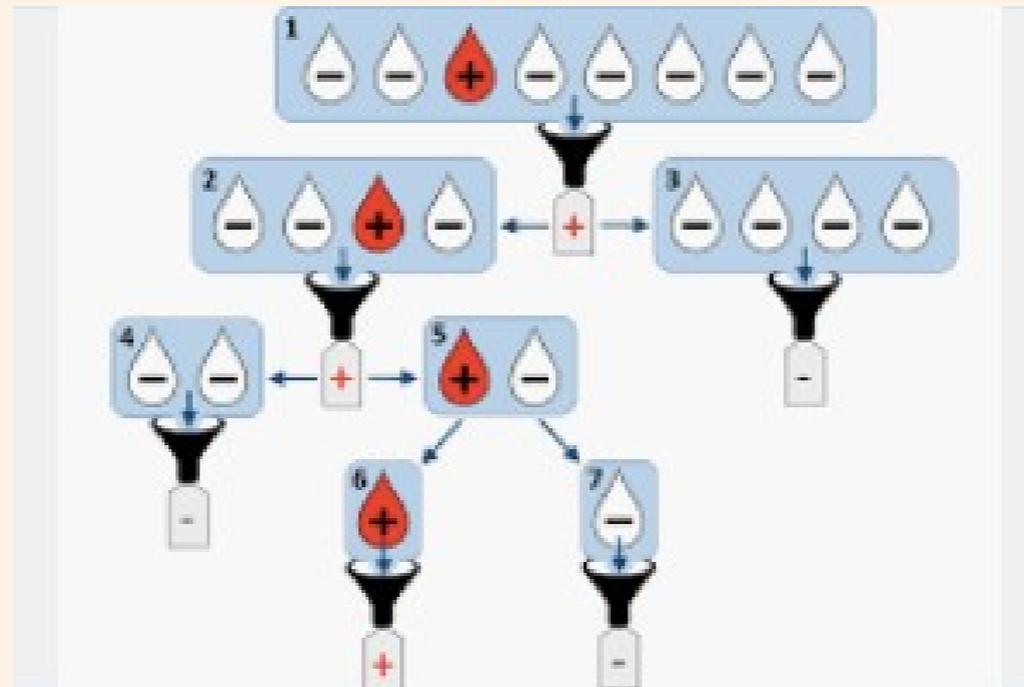
The syphilis problem

Dorfman (1942):

At most k of my n soldiers are infected with syphilis.
I need to find them using less than n tests!

Pool together patients.

If at least one is infected, then test outcome is “Positive”
If none of them is infected, then test outcome is “Negative”



Re-gained interest during COVID-19 pandemic

Combinatorial Group Testing

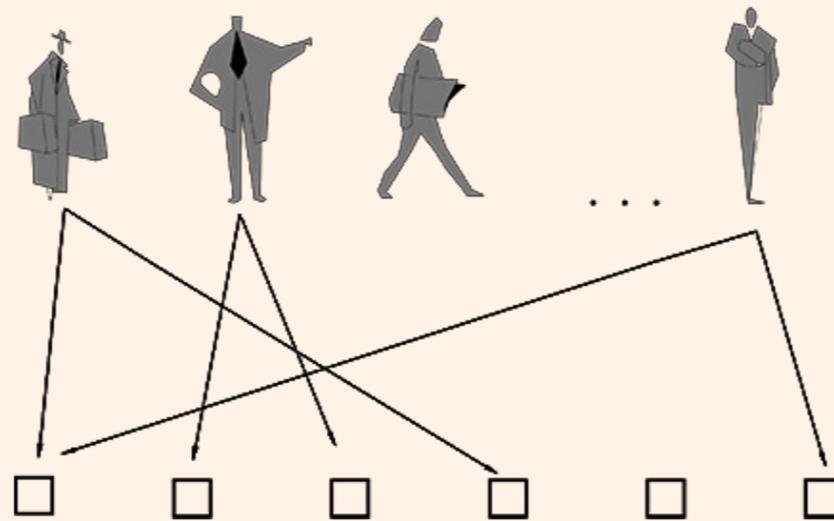
The basic set-up

Recover k -sparse $x \in \{\text{FALSE}, \text{RIGHT}\}^n$ using queries of the form $\bigvee_{i \in S} x_i$

Design measurements S_1, S_2, \dots, S_m

such that, given for all $j \in [m] : y_j = \bigvee_{i \in S_j} x_i$

we can recover $x \in \{\text{FALSE}, \text{RIGHT}\}^n$ if it is k -sparse



In matrix form (Boole Algebra): $y = Mx$ where $M_{j,i} = (i \in S_j)$

(Looks like linear sketching, right?)

Main Goal: Minimize number of measurements (tests)

Combinatorial Group Testing Guarantees

Non-uniform Group Testing: Recover a *fixed* k -sparse vector x with some target probability δ

Uniform Group Testing: Design measurements which allow recovery of *every* k -sparse vector x

$\Sigma_{k,n} :=$ set of $x \in \{\text{FALSE}, \text{TRUE}\}^n$ with at most k of the x_i equal to TRUE

Non-uniform Group Testing
(with δ failure probability)

$$\forall x \in \Sigma_{k,n} : \Pr \{ \text{FIND}(y) \neq x \} \leq \delta$$

Uniform Group Testing
(with δ failure probability)

$$\Pr \{ \exists x \in \Sigma_{k,n} : \text{FIND}(y) \neq x \} \leq \delta$$

Uniform is NOT *Deterministic*

Non-uniform Group Testing

$O(k \log n)$ measurements suffice

In every measurement S_j include i with probability $1/k$

No false negatives: If i is infected ($x_i = \text{True}$), then all the tests it participates in will be positive.

False positive: A non-infected i ($x_i = \text{False}$) such that all the tests it participates in turn out to be positive.

Consider test j in which i participates in

$$\Pr \{y_j = 1\} = 1 - \left(1 - \frac{1}{k}\right)^k \leq 1 - \frac{1}{e}$$

Why?

$$\Pr \{i \text{ false positive}\} = \left(1 - \frac{1}{e}\right)^{c \log n} \leq \frac{1}{\text{poly}(n)} + \text{union bound}$$

Find() : Keep every i which participates only in positive tests.

Recovery Time	$O(n \log n)$
Failure Probability	$\frac{1}{\text{poly}(n)}$

Uniform Group Testing, or One Matrix to rule them all

Find(): Keep every i which participates only in positive tests.

Disjunct matrices

A set of measurements with the following combinatorial property

$$\forall T \subseteq [n], |T| = k + 1 \text{ ; } \exists j \in [m] \text{ and } i \in T : \\ T \cap S_j = \{i\}$$

Disjunct matrices allow uniform group testing

Proof: Assume there exists a false positive i .

Let T be the set of infected individuals along with i .

Then T violates the disjunctness property.

Some bounds on measurements:

$$\Omega(k^2 \log_k n) \\ O\left(k^2 \min\left\{\log n, (\log_k n)^2\right\}\right)$$

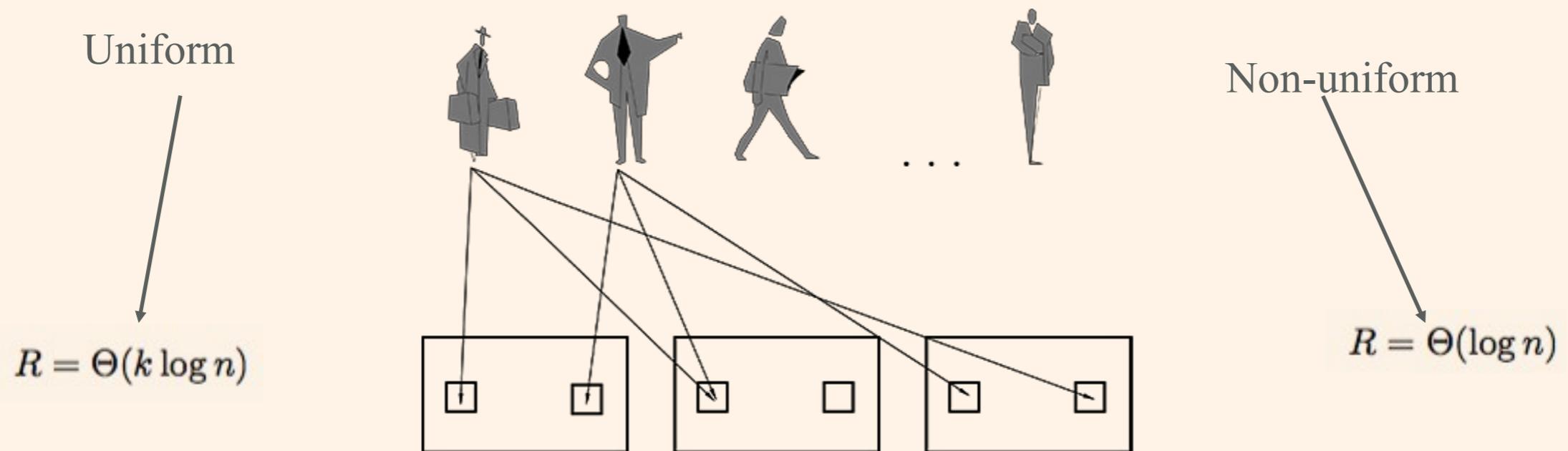
Gap since the 60s:
connected to a major question in
Coding Theory

Group Testing and the CountMin sketch

Pick (random) hash functions $h_r : [n] \rightarrow [2k]$ for $r \in [R]$

Pick measurements $h_r^{-1}(b) = \{i \in [n] : h_r(i) = b\}$

#measurements $2kR$



In CountMin sketch $R = \log(1/\delta)$ drives down the failure probability

You may think of $k = 1/\epsilon$

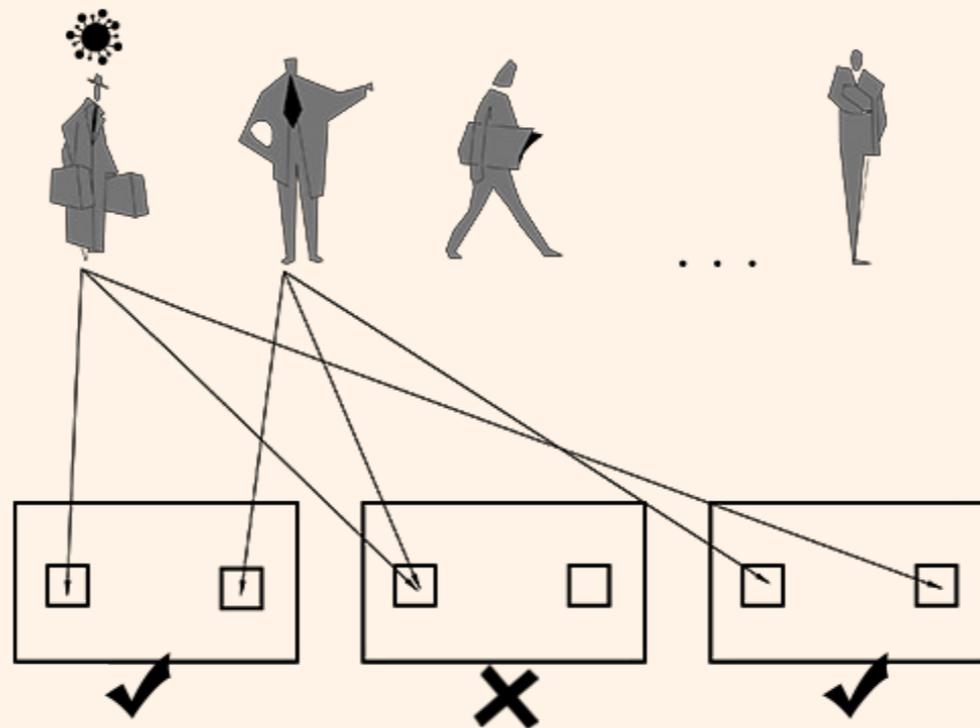
Analysis of disjunctness

Find(): Keep every i which participates only in positive tests.

Fix (T, i) , T a k -element subset of $[n]$ and i an element outside of T

$$\Pr \{h_r(i) \in h_r(T)\} \leq \frac{k}{2k} = \frac{1}{2}$$

$$\Pr \{h_r(i) \in h_r(T), \forall r \in [R]\} = \left(\frac{1}{2}\right)^R$$



$$\binom{n}{k} \approx e^{k \log(n/k)}$$

$$\Pr \{\text{disjunctness fails}\} \leq n \cdot \binom{n}{k} \cdot \left(\frac{1}{2}\right)^R \leq \frac{1}{3} \quad \text{since } R = \Theta(k \log n)$$

#measurements: $O(k^2 \log n)$

Recovery Time: $O(nk \log n)$

Two-stage group testing

We are allowed *two* rounds of adaptivity

Narrow down the set of possible infected individuals to $2k$,
then perform a test on each one.

The correct combinatorial construct
List-disjunct matrices

Written in terms of hash functions

$$\forall S, T \subseteq [n], S \cap T = \emptyset, |S| = k, |T| = k + 1, \\ \exists j \in T, r \in [R] : h_r(j) \notin h_r(S)$$

Find(): Keep every i which participates only in positive tests.



At most k false negatives

$O(k \log(n/k))$ measurements suffice (and is optimal)

Analysis of List-Disjunctness

Fix (S, T) .

Fix i in T .

$$\Pr \{h_r(i) \in h_r(S)\} \leq \frac{k}{2k} = \frac{1}{2}$$

Probability that all i in T appear as false positive in one repetition:

$$\Pr \{h_r(i) \in h_r(S), \forall i \in T\} \leq \left(\frac{1}{2}\right)^{k+1}$$

Probability that there exists a false negative

$$\Pr \{h_r(i) \in h_r(S), \forall i \in T, r \in [R]\} \leq \left(\frac{1}{2}\right)^{(k+1)R}$$

And now a union-bound over all pairs (S, T)

$$\binom{n}{k} \cdot \binom{n-k}{k+1} \cdot \left(\frac{1}{2}\right)^{(k+1)R} \quad \binom{n}{k} \approx e^{k \log(n/k)}$$

...smaller than $1/3$ by choosing $R = \Theta(\log(n/k))$

Compact Representations

All the above constructions can be analyzed using
 $O(k)$ -wise independent hash functions

Can store measurements in $O(k R \log n)$ bits of space

For disjunct matrices and non-uniform group testing,
2-wise independence suffices (in problem set)



Polynomial time derandomization for disjunct matrices

Practical Considerations

-Constraint 1: Cannot pick arbitrarily large groups

Additional Parameter: *Maximum group size*

Sparse Group Testing

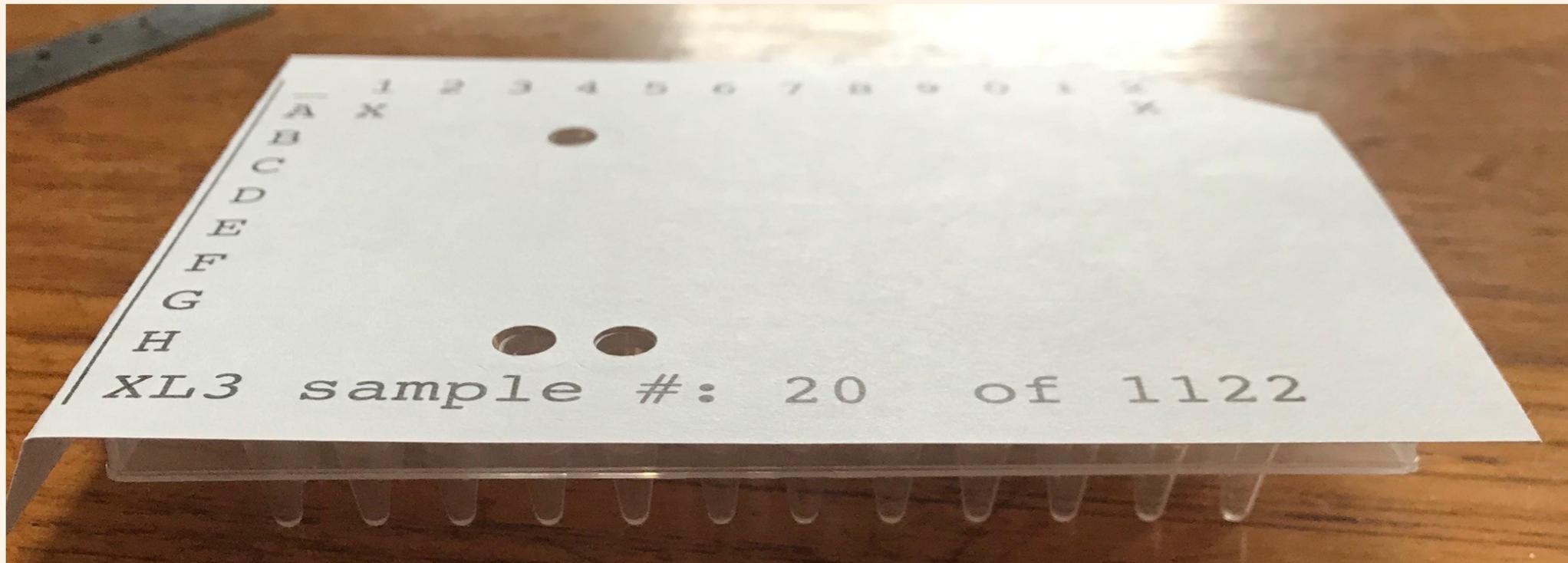
-Constraint 2: Some tests might fail

Error-Correcting Group Testing

-Constraint 3: Some times measurements must belong to a constrained ensemble

Application-dependent group testing

Bonus



Origami Assay paper template for group testing design,
designed during COVID-19 pandemic

Provides paper templates to guide the technician
on how to allocate patient samples across the test wells

Disclaimer: I do not understand how it works

Outline

Non-uniform Group Testing

Uniform Group Testing:

Disjunct Matrices and List-Disjunct Matrices

Group Testing from CountMin sketches

Thank you