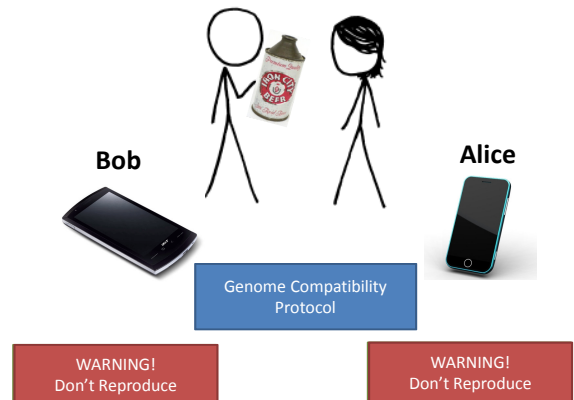


Secure Computation in the Real(ish) World

Carnegie Mellon
20 April 2011

David Evans
University of Virginia
<http://www.cs.virginia.edu/evans>
<http://www.MightBeEvil.com>

"Genetic Dating"



2



TheScientist News Current Issue Archive Sun

2 comments
Comment on this news story
By Kerry Grens

Forget mistletoe - what about DNA?

A new dating service matches singles using major histocompatibility complex genes



3

Genome Sequencing

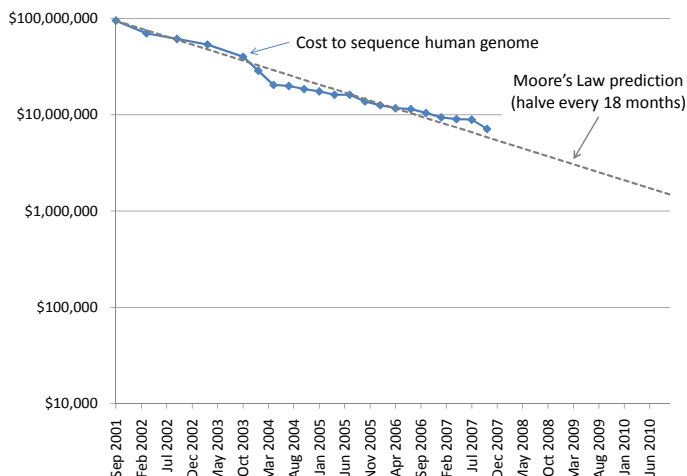
1990: Human Genome Project starts, estimate \$3B to sequence one genome (\$0.50/base)

2000: Human Genome Project declared complete, cost ~\$300M

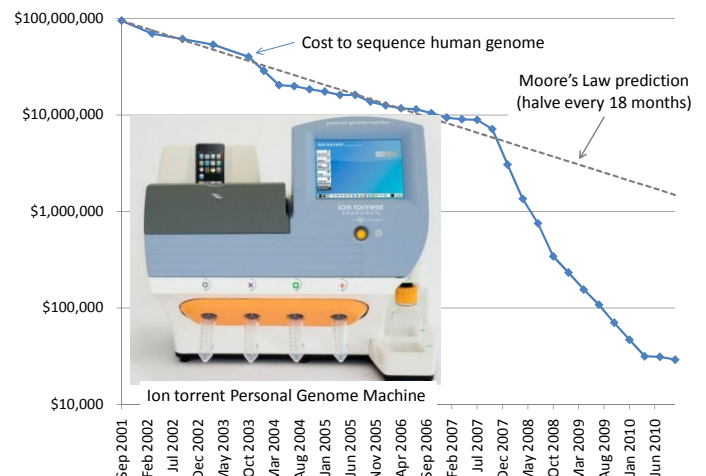


Whitehead Institute, MIT

4



Data from National Human Genome Research Institute: <http://www.genome.gov/sequencingcosts>



Data from National Human Genome Research Institute: <http://www.genome.gov/sequencingcosts>

Year	Reported sequencing consumables cost	Estimated cost per 40-fold coverage
	\$10,000,000	\$57,000,000
	\$1,000,000	\$5,700,000
	\$250,000	\$330,000
	\$48,000	\$69,000
2009	\$8,005	\$3,700
2009	\$3,451	\$2,200
2009	\$1,726	\$1,500



Human Genome Project
 Drmanac, An...
 Carnevali, Igc...
 Krishna P. Par...
 Chernikoff, Alex...
 Brian Hauser, S...
George Church (Personal Genome Project)
 ... Coleen R. Hacker, Robert Hartlage,
 ... Vin Kong, Tom Landers, Catherine Le,
 Jia Liu, Celeste E. McBride, Matt Morenzi, Robert E. Morey, Karl Mutch, Helena Perazich, Kimberly Perry, Brock
 A. Peters, Joe Peterson, Charit L. Pethiyagoda, Kaliprasad Pothuraju, Claudia Richter, Abraham M. Rosenbaum,
 Shaunak Roy, Jay Shafto, Uladzislau Sharanovich, Karen W. Shannon, Conrad G. Sheppy, Michel Sun, Joseph V.
 Thakuria, Anne Tran, Dylan Vu, Alexander Wait Zaranek, Xiaodi Wu, Snezana Drmanac, Arnold R. Oliphant,
 William C. Banyai, Bruce Martin, Dennis G. Ballinger, George M. Church, Clifford A. Reid. *Science*, January 2010.

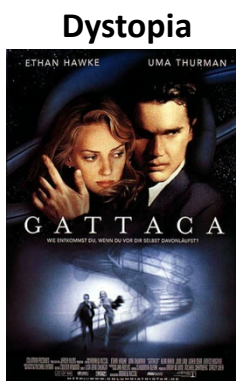


LETTERS

Steven Pinker (PGP-10)

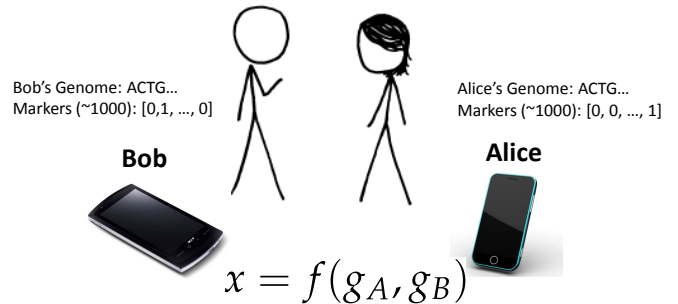
On Jim Watson's APOE status: genetic information is hard to hide

European Journal of Human Genetics (2009) 17, 147–149; doi:10.1038/ejhg.2008.198; published online 22 October 2008



Personalized Medicine

Secure Two-Party Computation



Can Alice and Bob compute a function of their private data, without exposing anything about their data besides the result?

Secure Function Evaluation

Alice (circuit generator) Picks $a \in \{0,1\}^s$
Bob (circuit evaluator) Picks $b \in \{0,1\}^t$
 Agree on $f(a,b) \rightarrow x$

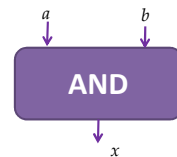
Garbled Circuit Protocol

Outputs $x = f(a,b)$ without revealing a to Bob or b to Alice.

Andrew Yao, 1982/1986

Yao's Garbled Circuits

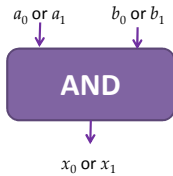
Inputs		Output
a	b	x
0	0	0
0	1	0
1	0	0
1	1	1



Computing with Meaningless Values?

Inputs		Output
a	b	x
a_0	b_0	x_0
a_0	b_1	x_0
a_1	b_0	x_0
a_1	b_1	x_1

a_i, b_i, x_i are random values, chosen by the circuit generator but meaningless to the circuit evaluator.

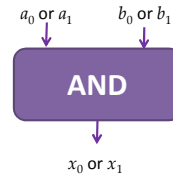


Computing with Garbled Tables

Inputs		Output
a	b	x
a_0	b_0	$Enc_{a_0,b_0}(x_0)$
a_0	b_1	$Enc_{a_0,b_1}(x_0)$
a_1	b_0	$Enc_{a_1,b_0}(x_0)$
a_1	b_1	$Enc_{a_1,b_1}(x_1)$

Bob can only decrypt one of these!

a_i, b_i, x_i are random values, chosen by the circuit generator but meaningless to the circuit evaluator.



Garbled And Gate
$Enc_{a_0,b_1}(x_0)$
$Enc_{a_1,b_1}(x_1)$
$Enc_{a_1,b_0}(x_0)$
$Enc_{a_0,b_0}(x_0)$

Garbled Circuit Protocol

Alice (circuit generator) Creates random keys: $a_0, a_1, b_0, b_1, x_0, x_1$
Bob (circuit evaluator)

And Gate
$Enc_{a_0,b_1}(x_0)$
$Enc_{a_1,b_1}(x_1)$
$Enc_{a_1,b_0}(x_0)$
$Enc_{a_0,b_0}(x_0)$

Sends a_i to Bob based on her input value

How does the Bob learn his own input wires?

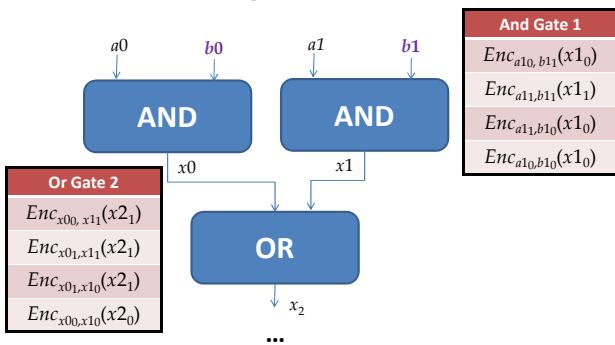
Primitive: Oblivious Transfer



Oblivious: Alice doesn't learn which secret Bob obtains
Transfer: Bob learns one of Alice's secrets

Rabin, 1981; Even, Goldreich, and Lempel, 1985; many subsequent papers

Chaining Garbled Circuits



We can do **any** computation privately this way!

Threat Model

Semi-Honest (Honest But Curious) Adversary

Adversary follows the protocol as specified (!)
 Curious adversary tries to learn more from protocol execution transcript

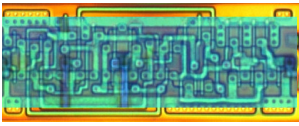
Garbled Circuits security proofs depend on this very weak model

General techniques for converting protocols secure in semi-honest model to resist malicious adversary.

Amount of information that could leak is probably small

Possibility to use software attestation to validate executing code?

Building Computing Systems

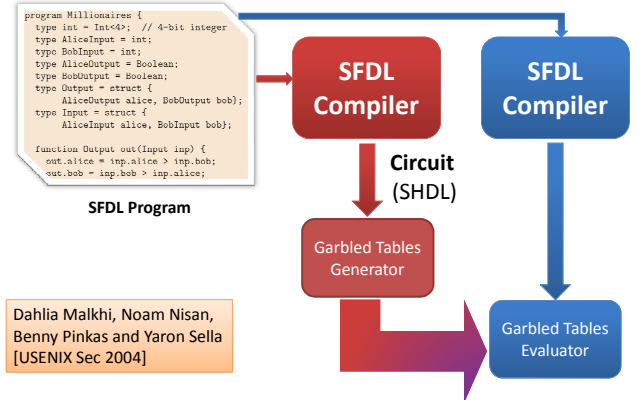


$Enc_{x_{00}, x_{11}}(x_{21})$
 $Enc_{x_{01}, x_{11}}(x_{21})$
 $Enc_{x_{01}, x_{10}}(x_{21})$
 $Enc_{x_{00}, x_{10}}(x_{20})$

Digital Electronic Circuits	Garbled Circuits
Operate on known data	Operate on encrypted wire labels
One-bit logical operation requires moving a few electrons a few nanometers (hundreds of Billions per second)	One-bit logical operation requires performing (up to) 4 encryption operations (~100,000 gates per second)
Reuse is great!	Reuse is not allowed!
All basic operations have similar cost	Some logical operations "free" (XOR, NOT)

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Fairplay



Dahlia Malkhi, Noam Nisan, Benny Pinkas and Yaron Sella [USENIX Sec 2004]

20

(Un)Fairplay?

An alternative approach to our protocols would have been to apply Yao's generic secure two-party protocol to the recognition algorithm. This would have required expressing the algorithm as a circuit which computes and compares many Hamming distances, and then sending and computing that circuit. ... **We therefore believe that the performance of our protocols is significantly better than that of applying generic protocols.**

Margarita Osadchy, Benny Pinkas, Ayman Jarrous, Boaz Moskovich. SCIFI – A System for Secure Face Identification. Oakland 2010.

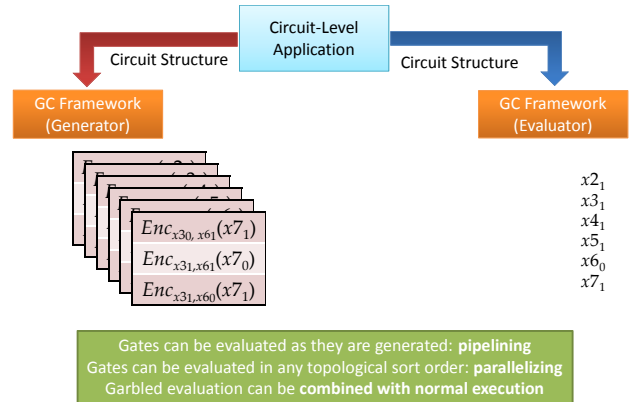


Protocol 1 (generic SMC) is very fast. Protocol 1 is ideal for small strings because the entire computation is performed in one round, but the circuit size is extremely large for longer strings. Our prototype circuit compiler can compile circuits for problems of size (200, 200) but uses almost 2 GB of memory to do so. **Significantly larger circuits would be constrained by available memory for constructing their garbled versions.**

Somesh Jha, Louis Kruger, Vitaly Shmatikov. Towards Practical Privacy for Genomic Computation. Oakland 2008.

21

Faster Garbled Circuits



22



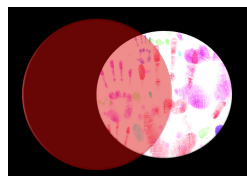
Privacy-Preserving Biometric Matching

Private Personal Genomics



Applications

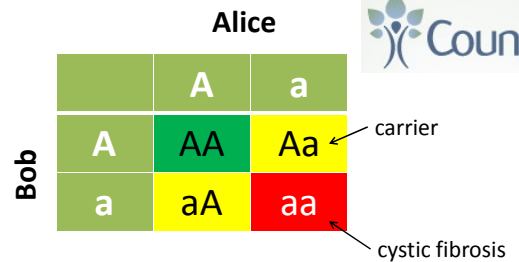
Private AES Encryption



Private Set Intersection

23

Heterozygous Recessive Risk

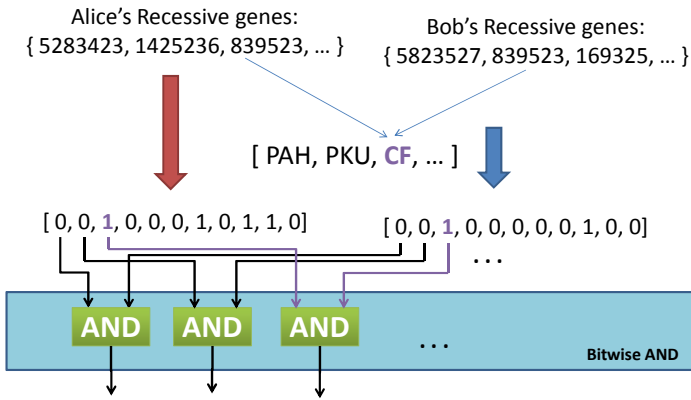


Alice's Heterozygous Recessive genes: { 5283423, 1425236, 839523, ... }
 Bob's Heterozygous Recessive genes: { 5823527, 839523, 169325, ... }

Goal: find the intersection of A and B

24

Bit Vector Intersection



25

Scaling

What if there are millions of possible diseases?

Length of bit vector:

number of possible values

(2^L where L is number of bits for each value)

Other private set intersection problems:

Do Alice and Bob have any friends in common?

Data mining problems: combine medical records across hospitals

Two companies want to do joint marketing to common customers

26

Pairwise Comparison

randomly permute A

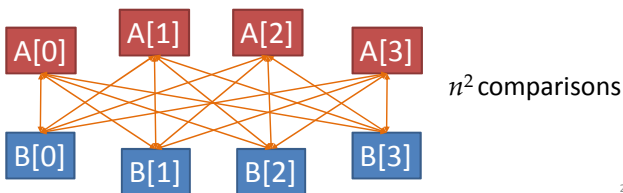
randomly permute B

for i in range(0, $n-1$):

for j in range(0, $n-1$):

if $A[i] = B[j]$ output $A[i]$

data-oblivious algorithm



27

Short-Circuit Pairwise Comparison

for i in range(0, $n-1$):

mask[i] = false

for i in range(0, $n-1$):

for j in range(0, $n-1$):

if not mask[i] and $A[i] = B[j]$:

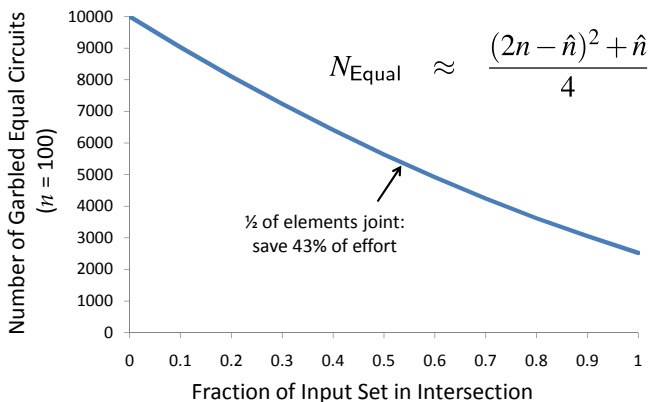
reveal $A[i]$ to both

mask[i] = true

break

28

Short-Circuit Analysis



29

Scaling

Other private set intersection problems:

Do Alice and Bob have any friends in common?

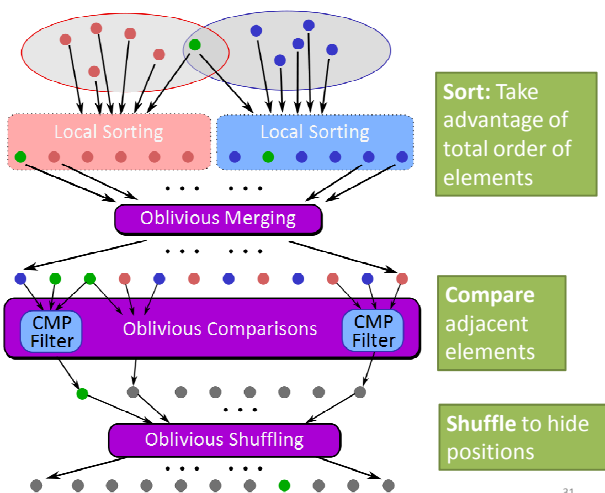
Data mining problems: combine medical records across hospitals

Two companies want to do joint marketing to common customers

This is still $O(n^2)$. Is there an $O(n \log n)$ solution?

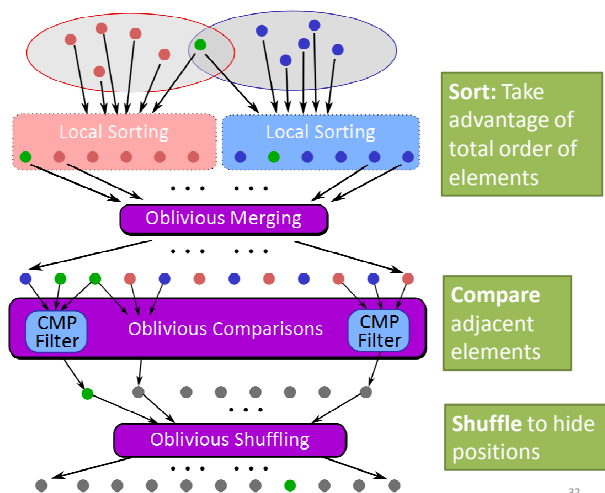
30

Sort-Compare-Shuffle



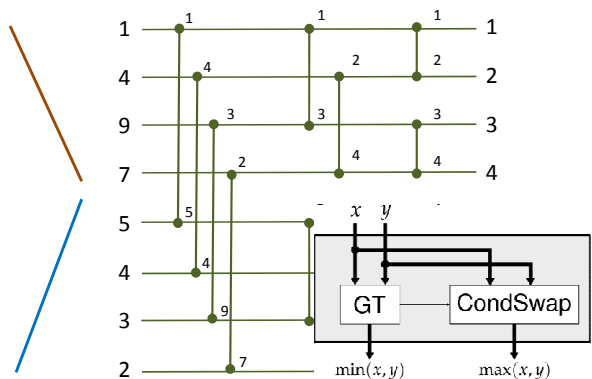
31

Sort-Compare-Shuffle

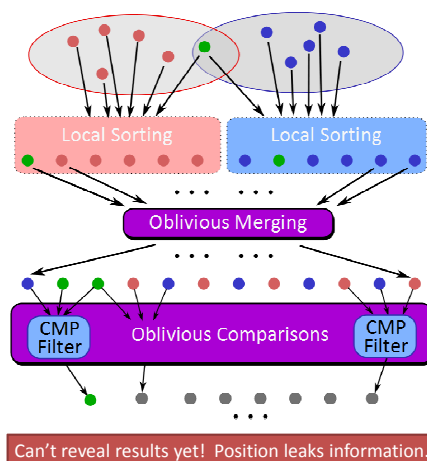
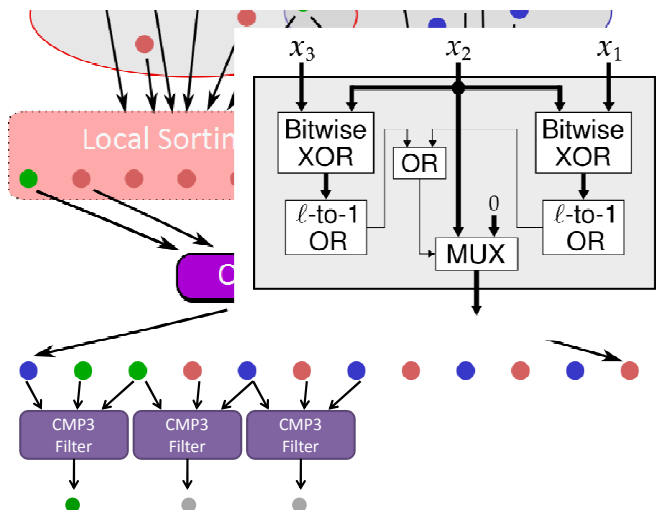
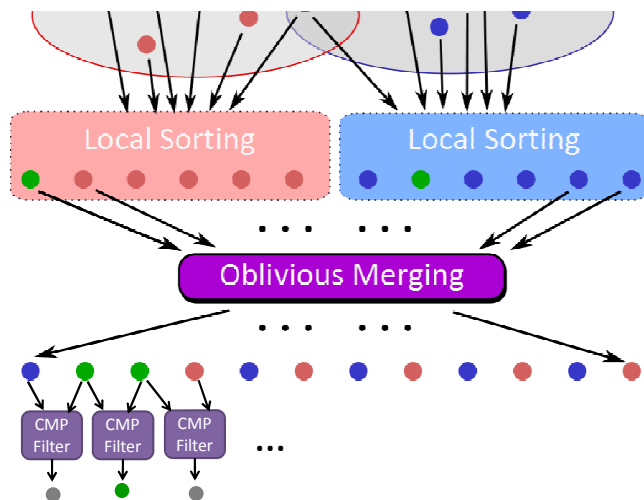


32

Bitonic Sorting



33



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Oblivious Shuffling

Homomorphic Encryption Shuffling Protocol

Add random mask, permute, exchange and reveal

Expensive

Sort

Simple...but expensive

Random Permutation

A Permutation Network

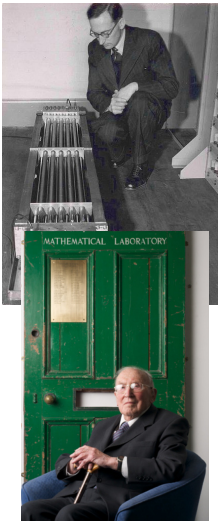
ABRAHAM WAKSMAN

Stanford Research Institute, Menlo Park, California

ABSTRACT. In this paper the construction of a switching network capable of $n!$ -permutation of its n input terminals to its n output terminals is described. The building blocks for this network are binary cells capable of permuting their two input terminals to their two output terminals.

The number of cells used by the network is $(n \cdot \log_2 n - n + 1) = \sum_{k=1}^n (\log_2 k)$. It could be argued that for such a network this number of cells is a lower bound, by noting that binary decision trees in the network can resolve individual terminal assignments only and not the partitioning of the permutation set itself which requires only $(\log_2 n!) = (\sum_{k=1}^n \log_2 k)$ binary decisions.

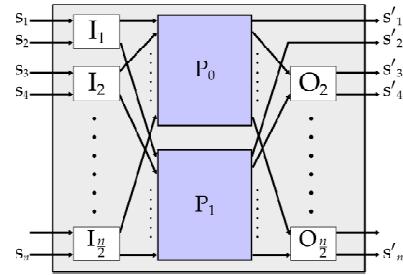
Journal of the ACM, January 1968



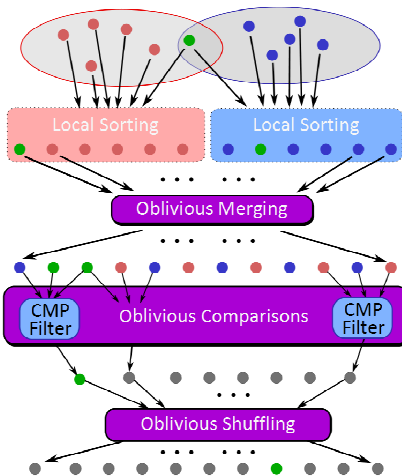
I do not imagine that many of the Turing lecturers who will follow me will be people who were acquainted with Alan Turing. ... Although a mathematician, Turing took quite an interest in the engineering side of computer design... **Turing's contribution to this discussion was to advocate the use of gin, which he said contained alcohol and water in just the right proportions ...**

Sir Maurice Wilkes (1913-29 Nov 2010), *Computers Then and Now* (1967 Turing Award Lecture)

Waksman Network



Same circuit can generate any permutation: select a random permutation, and pick swaps $n \log n - n + 1$ gates



Private Set Intersection Protocol

Gates to generate and evaluate

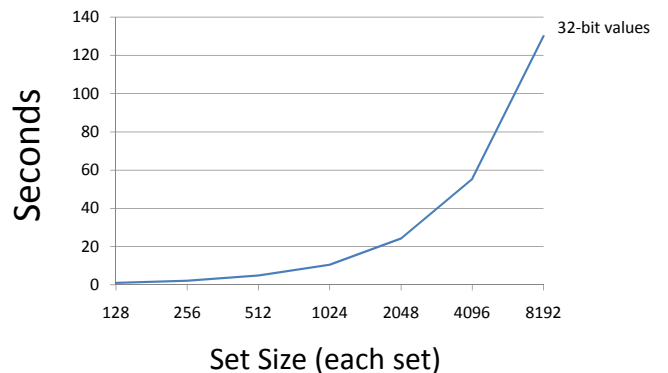
Free

$$n \log(2n) \times 2\sigma$$

$$(3\sigma - 1)(n - 1) + (2\sigma - 1)$$

$$n \log n - n + 1$$

Private Set Intersection Results



Some Other Results

	Problem	Best Previous Result	Our Result	Speedup
USENIX Security 2011	Hamming Distance (Face Recognition, Genetic Dating) – two 900-bit vectors	213s [SCIFI, 2010]	0.051s	4176
	Levenshtein Distance (genome, text comparison) – two 200-character inputs	534s [Jha+, 2008]	18.4s	29
	Smith-Waterman (genome alignment) – two 60-nucleotide sequences	[Not Implementable]	447s	-
NDSS 2011	AES Encryption	3.3s [Henecka, 2010]	0.2s	16.5
	Fingerprint Matching (1024-entry database, 640x8bit vectors)	~83s [Barni, 2010]	18s	4.6

Scalable: 1 Billion gates evaluated at ~100,000 gates/second on laptop

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Demo!

Private Set Intersection
on
Android
Devices

<http://MightBeEvil.com/mobile/>
Peter Chapman and Yan Huang

44



Yan Huang

(UVa Computer Science PhD Student)

Funding: **NSF, MURI** (AFOSR)
Android toys: **Google**



Peter Chapman

(UVa BACS 2012)



Aaron Mackey

(UVA Public Health Genomics)



Jonathan Katz

(University of Maryland)

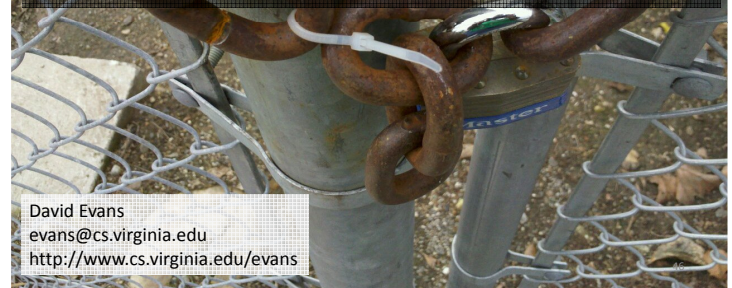


Lior Makla

(UMd / Intel)

Much of the early engineering development of digital computers was done in universities. A few years ago, the view was commonly expressed that universities had played their part in computer design, and that the matter could now safely be left to industry. I do not think that it is necessary that work on computer design should go on in all universities, but I am glad that some have remained active in the field. Apart from the obvious functions of universities in spreading knowledge, and keeping in the public domain material that might otherwise be hidden, universities can make a special contribution by reason of their freedom from commercial considerations, including freedom from the need to follow the fashion.

Sir Maurice Wilkes (June 1913-Nov 2010), 1967 Turing Award Lecture



David Evans
evans@cs.virginia.edu
<http://www.cs.virginia.edu/evans>

Introduction to Computing

Explorations in Language, Logic, and Machines
Spring 2010

www.computingbook.org

David Evans
University of Virginia

Shameless Plug

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