Learning Your Identity and Disease from Research Papers: Information Leaks in Genome Wide Association Study

> Rui Wang, Yong Li, XiaoFeng Wang, Haixu Tang, Xiaoyong Zhou





### **Presentation Overview**

- Brief Introduction: Genomes, SNP, GWAS
- Privacy Implications of GWAS
- Authors' Attacks
- Defense
- Implementation
- Conclusion

### Genome

- Complete set of genes in a single organism
- Entirety of an organism's hereditary information
- Human Genome Project (HGP)<sup>1</sup> produced a reference sequence of the human genome

### Single Nucleotide Polymorphisms (SNP)

 DNA sequence variations that occur when a single nucleotide (A,T,C,or G) in the genome sequence is altered<sup>1</sup>



<sup>1</sup> http://www.ornl.org/sci/techresources/Human\_Genome/faq/snps.shtml <sup>2</sup> Picture: http://science.marshall.edu/murraye/341/Images/416px-Dna-SNP\_svg.png

## Single Nucleotide Polymorphisms (SNP)

- Variation must occur in 1% population to be considered a SNP
- SNP contains a major allele (0) and a minor allele (1)
- Large amount of information
  - Individual frequency (1 or 0)
  - or SNP pairs of allele (00, 01, 10, 11)

# Genome-wide Association Studies (GWAS)

- GWAS developed to leverage genome data to discover:
  - Genetic variations (SNPs)
  - Common diseases
- Data widely available
  - HapMap (<u>http://hapmap.ncbi.nlm.nih.gov/</u>)
- Individuals' disease susceptibility

### Privacy Implications for GWAS DBs

- Privacy enforced through individuals' consent
- Individuals' disease susceptibility
  - Insurance
  - Profiling
  - Dating ... or perhaps "Dataing"



### Existing Database Attacks of GWAS

- Homer's Attack
  - Individual's blood compared to a target population
  - If distribution of risk alleles match, individual ID'd
- Subverting database anonymization
  - By analyzing the remaining data, feature information can be used to ID the individual
  - Ex: Blonde hair, blue eyes
- Database connections



### Paper Framework

- Preexisting attacks
- Novel identification attacks on GWAS statistics
  - Smaller reference populations
- Implementation of attacks
- Study of the attack countermeasures
- Attack results and evaluations

# Attack 1: From Statistics to Allele Frequencies





## Attack I



Figure 1: Recover allele frequencies.

 How likely one SNP can be used to infer some of the subjects other SNPs



### Attack I

• Allow for a range of acceptable boundaries by using inequalities:

 $L < r^{2} < U$ 

- Result is *positive* (true) if the signs hold, or *negative* (false) otherwise
- If false, then infers that the sign's may need to be recovered (switched)

## Attack II: A Statistic Attack



## Attack II

- Establish a reference group
  - SNP sequences from group of individuals
  - Same genetic background of the case group
- Derived from HMAP studies
- High confidence when results in *linkage distribution* (LD)
  - Combinations of alleles or genetic markers occur more or less frequently in a population than would be expected from a random formation

### Attack II

- Assumes a null hypothesis that the victim is not in the case group
- *T<sub>r</sub>* is the statistic designed to make the presence of an victim in the case group *valid*
- Given a positive result of T<sub>r</sub>, an individual's SNP can be distinguished from the group therefore identifying the individual



### Attack II

 Since single allele correlations are *not* normally completely independent, cannot assume null hypothesis



 Result is the similarity between the case group's r<sup>2</sup> and the victim's r<sup>2</sup>

# Attack III: Integer Programming Attack



### Attack III

 Given allele frequencies for the surrounding regions of a SNP site (*locus*)

### Haplotypes

- Specific combination of alleles across multiple neighboring SNP sites in a locus
- Each individual has two haplotypes inherited from the parents
- Population level some haplotypes are more common than others.



### Attack III



- "Divide and Conquer"
- Instead of computing every block derived from haplotypes merge haplotypes based on strong correlation between two SNPs

### Defense

- Low-precision statistics
  - Downgrade the *linkage distribution* (LD)
  - Limiting the accuracy in comparing the victim's LD
  - Using allele frequencies still restored over 50% of pairwise frequencies and all the signs

### Thresholds

- Publish less data  $\rightarrow$  less informative
- Sufficient information for recovering signs, attack still works

### Noise

Mitigates attack, but data becomes less useful

### Implementation

- (1) Infer allele frequencies for individual SNPs and SNP from statistics (GWAS)
- (2) Propagate the marker SNP frequencies to other SNPs by using r<sup>2</sup>

### • Result:

- Recovered all SNP frequencies
- Half of pairwise frequencies
- Most of the signs for r

### Evaluations

- (1) Infer allele frequencies for individual SNPs and SNP from statistics (GWAS)
- (2) Propagate the marker SNP frequencies to other SNPs by using r<sup>2</sup>
- Result:
  - Recovered all SNP frequencies
  - Half of pairwise frequencies
  - Most of the signs for r

### **Evaluations**

- Using Markov model against GWASs
  - Low-precision attacks
    - 79% statistical power retained
  - Threshold defense
    - 85% statistical power retained
- Integer-programming attack
  - Run on 100 individuals
  - Within 12 hours successfully restored 174 SNPs for all 100 participants

### Implementation

00001

- Case = red dots, References (Ref) = green dots
- Tests: Test = blue dots, Test1 = black dots



### Conclusion

- GWAS is a burgeoning field with a lot of attention placed upon the privacy, defense, and attacks of the studies' data
- This paper presents two new techniques that can lead to identification of victims in a GWAS
- Key: Form a \*small set of statistics\* routinely published in GWAS studies



### Questions...

• ...for the authors?

# Attack I: Correlation and Recovery of SNP Alleles

• High r<sup>2</sup> value (0.93) =

$$r^{2} = \frac{(C_{00}N - C_{*0}C_{0*})^{2}}{C_{0*}C_{1*}C_{*0}C_{*1}}$$

• Quick rundown...