# Familial Searching of Forensic DNA Databases 

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## Outline

- Case aided with familial searching
- Fundamentals of searching for relatives
- Research with New Zealand DNA database
- Ways to increase efficiency of familial searching


## Grim Sleeper Case

- 12 victims murdered in Los Angeles (1985-2007)
- Cases linked through firearms analysis

Over a 13 year gap in detected crimes, hence the "Sleeper" nickname

- DNA evidence recovered and searched against state and national database
- California Dept. of Justice initiated a research program to evaluate the use of familial searching
- Program was developed and validated using NIST population data from autosomal and Y-STR markers
- Data are freely available on the STRBase website


## Familial Search for the Grim Sleeper

- October 2008: First familial search of the California database (over 1.1 million profiles) yielded no strong possibilities
- June 30, 2010: Second familial search of the California database (over 1.3 million profiles) yielded one likely relative
- Database profile from Christopher Franklin (31 years old)
- Profile added in 2009 after a felony weapons possession charge
- Profiles from Grim Sleeper evidence and C. Franklin shared one allele at all 15 loci
- Both individuals shared the same Y-STR profile

CRIME \& COURTS
Arrest Made in L.A. 'Grim Sleeper' Killings


## Familial Searching

- Search unknown evidence profile against forensic DNA database to identify possible close relatives of the true offender
- For no suspect cases, cold cases, violent crimes to develop investigative leads
- Success in the United Kingdom (2004 - Jan. 2011)
- 179 cases submitted; 36 successes/81 cases completed (44.4\% success rate)
- Metadata (age, locality, ethnicity) increase success
- Familial searching programs in the U.S.
- Colorado: all forensic unknowns, 10 identifications, 1 conviction (as of June 2011)
- California: 13 searches, 2 arrests (as of March 2011)
- Virginia: validation completed (March 2011)
- Texas


## Fundamentals of Searching for Relatives

## Allele Sharing: Parent-Offspring

Single locus example


Probability of sharing alleles from a common ancestor (per locus)

$$
\begin{aligned}
& \operatorname{Pr}(0 \text { alleles })=0 \\
& \operatorname{Pr}(1 \text { allele })=1 \\
& \operatorname{Pr}(2 \text { alleles })=0
\end{aligned}
$$

$$
12,13
$$

1 allele shared between any parent and child

## Allele Sharing: Full Siblings

## Single locus example



Probability of sharing alleles from a common ancestor (per locus)

$$
\begin{aligned}
& \operatorname{Pr}(0 \text { alleles })=1 / 4 \\
& \operatorname{Pr}(1 \text { allele })=1 / 2 \\
& \operatorname{Pr}(2 \text { alleles })=1 / 4
\end{aligned}
$$

0 alleles shared between these full siblings

## Allele Sharing: Half Siblings*

Single locus example


0 alleles shared between
Probability of sharing alleles from a common ancestor (per locus)

$$
\begin{aligned}
& \operatorname{Pr}(0 \text { alleles })=1 / 2 \\
& \operatorname{Pr}(1 \text { allele })=1 / 2 \\
& \operatorname{Pr}(2 \text { alleles })=0
\end{aligned}
$$

## How is kinship assessed?

## Likelihood Ratio (LR)

Evaluate genotypes to give weight (strength) to compared relationships
$L R=\frac{\text { Probability of genotypes if } 1,2 \text { are full siblings }}{\text { Probability of genotypes if } 1,2 \text { are unrelated }}$

By the definition of a LR:
LR > 1 supports the numerator (alleged relationship)
$L R<1$ supports the denominator (unrelated)

Larger LR values provide more support for the alleged relationship

# Research with New Zealand DNA Database 

Ph.D. dissertation with<br>Bruce Weir and Mary-Claire King<br>UW Genome Sciences

## Statistical Modeling

- Assessed the effectiveness of searching for parent-offspring, full sibling, and half sibling (or equivalent) relationships
- Used the New Zealand DNA Database
- 80,000 subjects
- 10-locus profiles
- Performed 1,000 simulations by generating one true relative pair per search


## Statistical Modeling

| Database Profiles |
| :---: |
| Index1 <br> Index2 <br> Index3 <br> Index4 <br> Index5 <br> Index6 <br> Index7 <br> Index8 <br> $\downarrow$ |
| Allele frequencies |

## Statistical Modeling



## Statistical Modeling



## Statistical Modeling



## Challenge of Identifying True Relatives in a Database

| Rank | Index \# |
| :---: | :---: |
| 1 | Index5243 |
| 2 | Index1438 |
| 3 | Parent1 |
| 4 | Index45677 |
| 5 | Index39732 |
| 6 | Index134 |
| 7 | Index7701 |
| . | . |
| 412 | Index22093 |
| 413 | Index208 |

Likelihood Ratio (LR)

| 7048 |  |
| :---: | :---: |
| 5503 |  |
| 45 | - False positive |
| 3 |  |
| 0 |  |
| 0 |  |
| 0 |  |
| . | - False positive |
| . |  |
| 0 |  |
| 0 |  |
| 0 |  |

- Unrelated individuals may have higher LRs due to chance allele sharing
- Included in subsequent investigation $\rightarrow$ "False positive"
- True relatives will not always have the highest LR
- Potentially not included in subsequent investigation $\rightarrow$ "False negative"


## Evaluate how often a true relative will be found in a database search

1. Tracked the ordered rank of the true relative for each of the 1000 simulations
2. Calculated the cumulative frequency of true relatives (counts per rank/1000)
3. Think of frequencies as the empirical probability of finding true relative after investigating a certain proportion of individuals in the database

## Ordered Rank Proportion of Database

= rank/database size
$\left.\begin{array}{|c|c|}\hline 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ - \\ - \\ 412 \\ 413\end{array} \quad \begin{array}{|c}0.002 \\ 0.004 \\ 0.006 \\ 0.008 \\ 0.010 \\ 0.012 \\ 0.014 \\ . \\ . \\ \hline\end{array}\right]$ Top 1\%

| 108 |
| :---: |
| 88 |
| 83 |
| 46 |
| 47 |
| 34 |
| 27 |
| . |
| . |
| 0 |
| 1 |

\(\left.\begin{array}{|c|}\hline 0.180 <br>
0.088 <br>
0.083 <br>
0.046 <br>
0.047 <br>
0.034 <br>
0.027 <br>
. <br>
. <br>
0 <br>

0.001\end{array}\right] \quad\)| 0.444 |
| :--- |
| cumulative |
| frequency |

"In an example database of 500 profiles, the probability of finding the true parentoffspring is 0.444 if the top $1 \%$ of LR values are investigated after familial searching."

## Probability of finding a true relative given the proportion of the NZ database investigated (10 STR loci, $n=80,000$ )

"Simulations indicate that the probability of finding the true parent-offspring is approximately 0.95 if the top $0.1 \%$ of $L R$ values are investigated."


## Probability of finding a true relative given the proportion of the NZ database investigated (10 STR loci, $n=80,000$ )

"Simulations indicate that the probability of finding the true full sibling is approximately 0.95 if the top $3 \%$ of LR values are investigated."


## Probability of finding a true relative given the proportion of the NZ database investigated (10 STR loci, $n=80,000$ )

"Simulations indicate that the probability of finding the true half sibling is approximately 0.95 if the top $44 \%$ of LR values are investigated."


## Trade-off between False Positives and False Negatives

1. Set LR threshold to filter ranked list of potential relatives.
2. What is the probability that a true relative is in this filtered list (PT)?
3. How many false positives will be included in filtered list (FP)?

| LR Threshold | Parent-Offspring |  | Full Siblings |  | Half Siblings |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | PT | FP | PT | FP | PT | FP |
| 100 | 0.95 | 18 | 0.70 | 5 | 0.03 | 2 |
| 10 | 1.00 | 45 | 0.86 | 60 | 0.27 | 184 |
| 1 | 1.00 | 79 | 0.95 | 440 | 0.73 | 4,570 |

Increasing the LR threshold makes familial searching less efficient but reduces the number of false positive leads

## The Reality of Full Sibling Searches



Range of likelihood ratios for true brothers illustrates chance for false leads, even with additional loci

## Kinship Statistics for True Full Siblings

| Comparison | Likelihood Ratio |  |  |
| :--- | ---: | ---: | ---: |
|  | 13 STRs |  | 19 STRs |
|  | 571 | 21,239 |  |
| Brother 1 | 2703 | 1360 |  |
| Brother 2 | 2 | 19,991 |  |
| Brother 3 | 1 | 2 |  |
| Brother 4 | 2 | 2 |  |

Ways to Increase the Efficiency of Familial Searching

## More Success for Within-State Searches

Number of profiles that would have to be investigated for $90 \%$ chance of finding true relative*

| Relationship | Median state database <br> $\mathrm{n}=100,000$ | National database <br> $\mathrm{n}=10,000,000$ |
| :--- | :---: | :---: |
| Parent-offspring | 37 | 3,700 |
| Full Siblings | 134 | 13,400 |
| Half Siblings | 17,441 | $1,744,100$ |

## $87 \%$ of CODIS hits are within state

* Extrapolated results from searching the New Zealand database with 13 CODIS loci


## Filter on Y-chromosome

- Missouri has 12 Y-STRs typed on 45,000 (20\%) database samples
- Females are "noise" - contribute to false positives
- Up to $20 \%$ of database profiles are female
- Not investigated if additional testing (Y-STR) is required
- Remove female profiles prior to familial search or follow-up with non-genetic information
- What about mitochondrial DNA?
- Sequencing costs are prohibitive
- Linear arrays have low resolution
- Incorporate LR of Y-STR match into search statistic

$$
\text { "Odds" }=\mathrm{LR}_{\text {autosomal STR }} * \mathrm{LR}_{\mathrm{Y}-\mathrm{STR}} * 1 / \mathrm{N}
$$

Myers et al., Searching for first-degree familial relationships in California's offender DNA database. FSI Genetics (in press)


## Database Longevity Leads to Parent-Offspring Searches

Demographic of male inmates held in custody in U.S. state or federal prison or in local jails, by age, as of June 30, 2009

| Age | Proportion of <br> Male Profiles |
| :--- | :---: |
| $18-19$ | 0.07 |
| $20-24$ | 0.14 |
| $25-29$ | 0.15 |
| $30-34$ | 0.16 |
| $35-39$ | 0.14 |
| $40-44$ | 0.12 |
| $45-49$ | 0.09 |
| $50-54$ | 0.06 |
| $55-59$ | 0.04 |
| $60-64$ | 0.02 |
| 65 or older | 0.01 |

## Conclusions

- Using science, a cost-benefit analysis is necessary to balance effort to find relatives against spending limited resources on false leads
- National database searches are not efficient due to large number of false positives
- Not yet a way to effectively follow up on female profiles
- Database age will increase the utility of parent-offspring searches and the efficiency of familial searching


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