



**California Association of Criminalists Fall Meeting  
06 November 2012 – San Jose, California**

# Exploring the Capabilities of Mixture Interpretation using True Allele Software

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National Institute of Standards and Technology  
Gaithersburg, Maryland USA



# Which of the topics below would be your first choice for additional training?

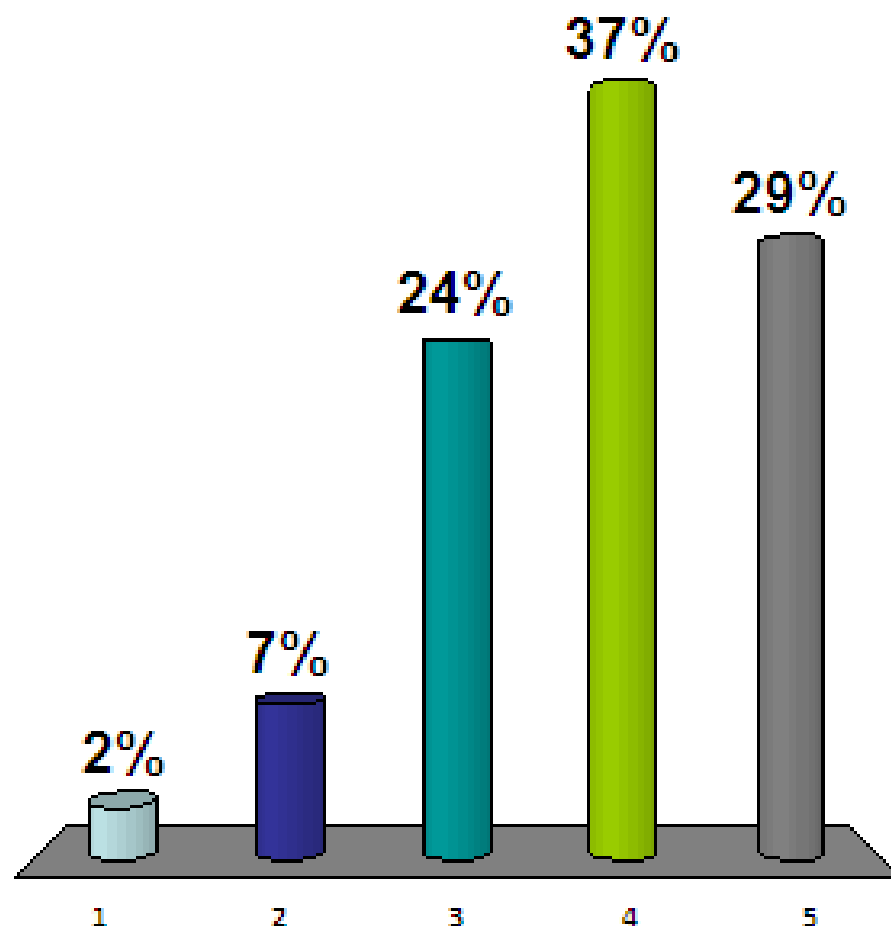
1. Relevant literature
2. How to validate thresholds
3. How to develop relevant SOPs

**4. Interpretation of low level mixtures**

**5. Statistics**

**2/3 want more information on these topics**

From one of the regional mixture workshops (Apr – June 2011)



# Stats Required for Inclusions

SWGDM Interpretation Guideline 4.1:

**“The laboratory **must perform statistical analysis** in support of any inclusion that is determined to be relevant in the context of a case, irrespective of the number of alleles detected and the quantitative value of the statistical analysis.”**

Buckleton & Curran (2008): “There is a considerable aura to DNA evidence. Because of this aura **it is vital that weak evidence is correctly represented as weak or not presented at all.**”

Buckleton, J. and Curran, J. (2008) A discussion of the merits of random man not excluded and likelihood ratios. *Forensic Sci. Int. Genet.* 2: 343-348.

# Statistics

A Tragedy in 400 Quadrillion Acts



“Though this be madness,  
yet there is method in't.”

— William Shakespeare, *Hamlet*

# Statistical Approaches with Mixtures

See Ladd *et al.* (2001) *Croat Med J.* 42:244-246

## “Exclusionary” Approach

Random Man Not Excluded  
(RMNE)

*Combined Prob. of Inclusion*  
(CPI)

*Combined Prob. of Exclusion*  
(CPE)

“Allele-centric”

## “Inferred Genotype” Approach

Random Match Probability  
[modified]  
(mRMP)

Likelihood Ratio  
(LR)

“Genotype-centric”

# A discussion of the merits of random man not excluded and likelihood ratios

John Buckleton<sup>a,\*</sup>, James Curran<sup>b</sup>

<sup>a</sup>*ESR, PB 92021, Auckland, New Zealand*

<sup>b</sup>*Department of Statistics, University of Auckland, PB 92019, Auckland, New Zealand*

Received 15 January 2008; received in revised form 29 April 2008; accepted 1 May 2008

We conclude that the two matters that appear to have real force are:

- (1) LR<sub>s</sub> are more difficult to present in court and
- (2) the RMNE statistic wastes information that should be utilised.

# Curran and Buckleton (2010)



**PAPER**

**CRIMINALISTICS; GENERAL**

*James M. Curran,<sup>1</sup> M.Sc.(Hons.), Ph.D. and John Buckleton,<sup>2</sup> Ph.D.*

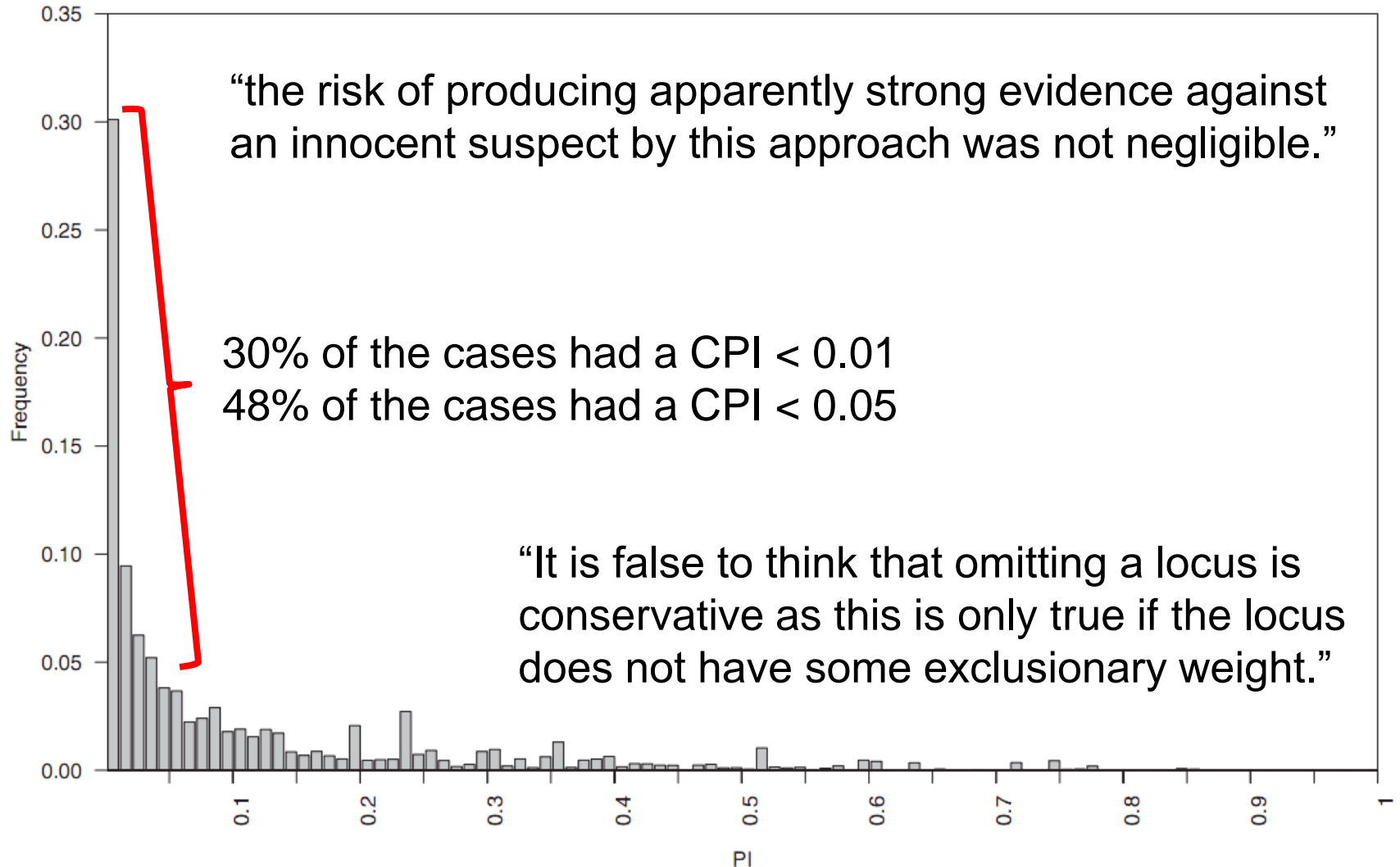
## Inclusion Probabilities and Dropout

Created 1000 Two-person Mixtures (Budowle *et al.* 1999 AfAm freq.).

Created 10,000 “third person” genotypes.

Compared “third person” to mixture data, calculated PI for included loci, ignored discordant alleles.

# Curran and Buckleton (2010)





# Review of Two Thresholds

Example values  
(empirically determined  
based on own internal  
validation)

200 RFUs

**Called Peak**

*(Greater confidence a sister  
allele has not dropped out)*

MIT

**Stochastic Threshold**

The value above which it is  
reasonable to assume that  
allelic dropout of a sister  
allele has not occurred

**Called Peak**

*(Cannot be confident  
dropout of a sister allele  
did not occur)*

50 RFUs

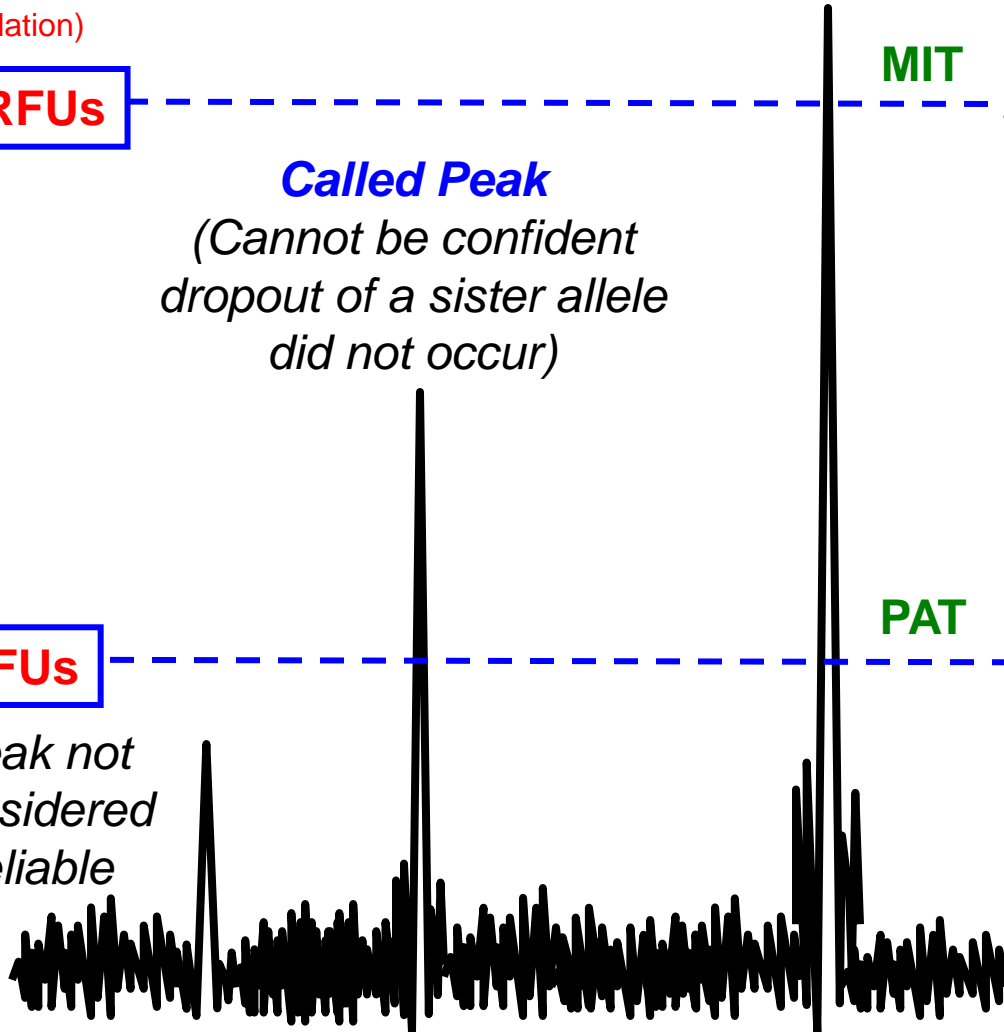
PAT

**Analytical Threshold**

Minimum threshold for data  
comparison and peak  
detection in the DNA typing  
process

*Peak not  
considered  
reliable*

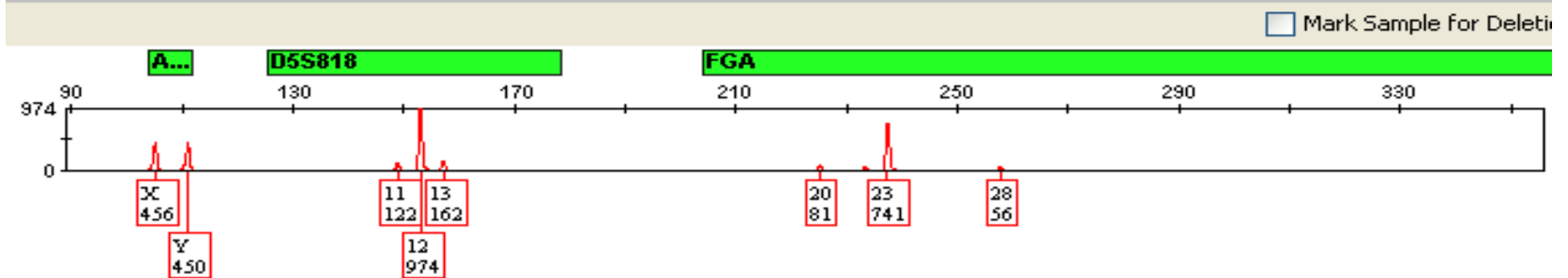
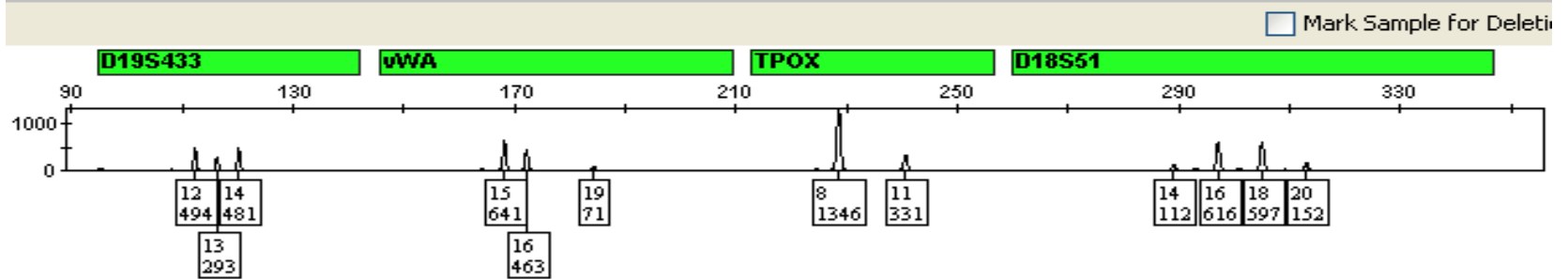
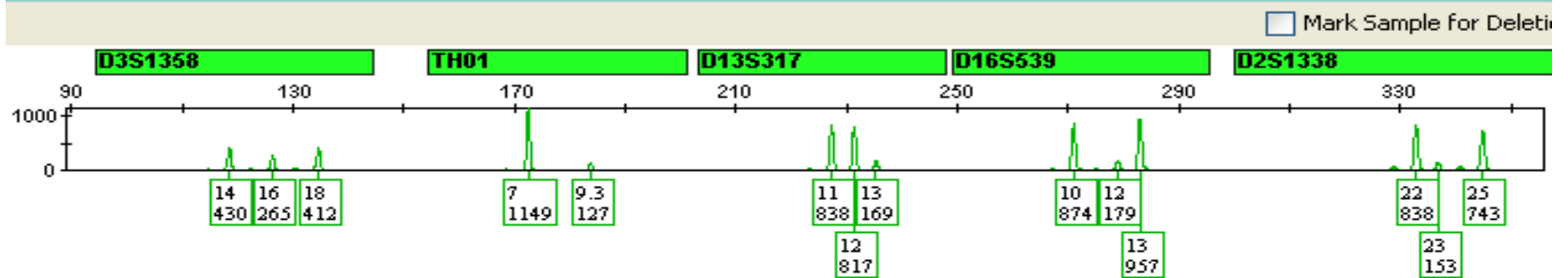
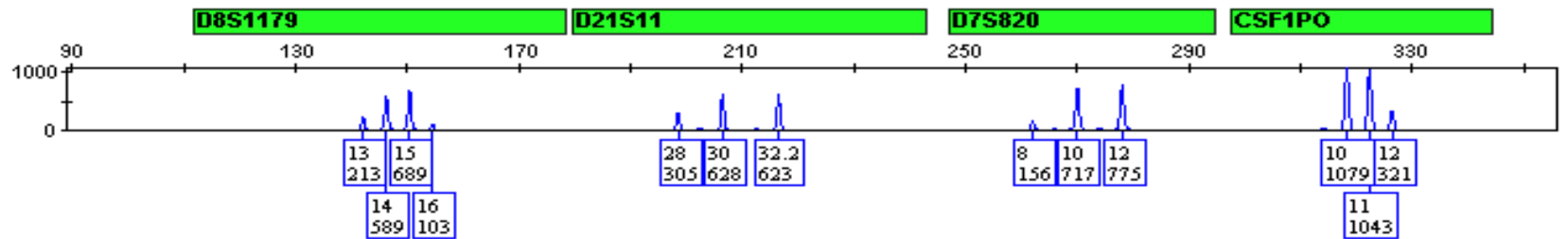
**Noise**



# 2-person Mixture



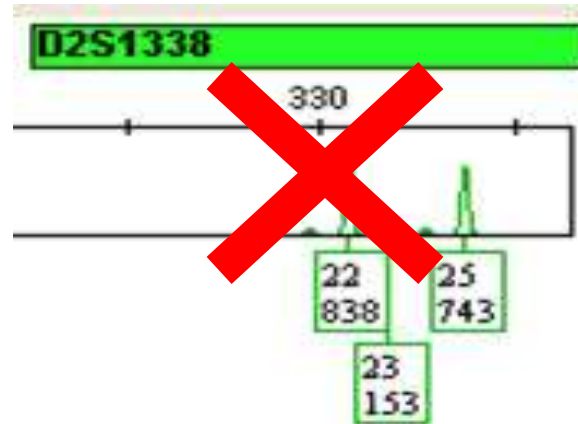
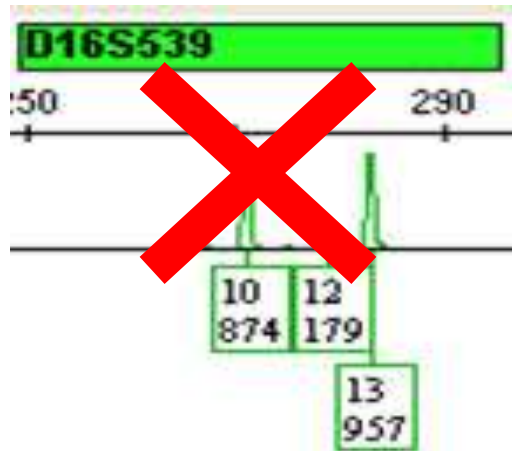
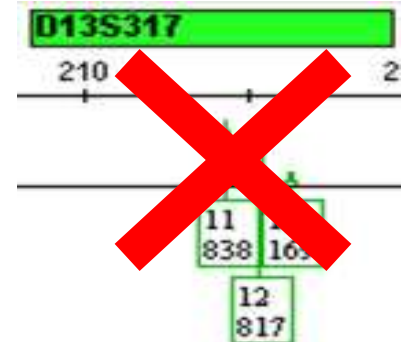
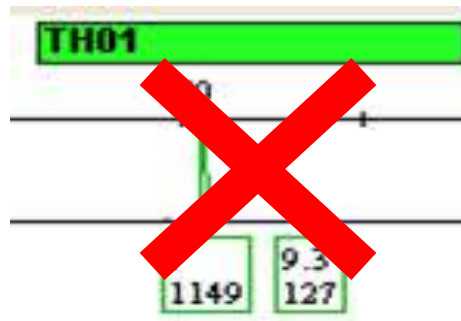
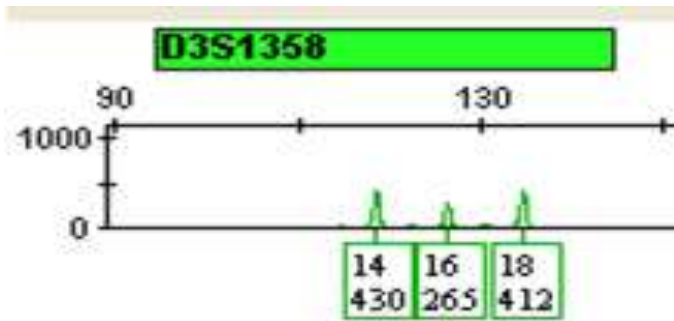
# 2-Person Mixture



# If CPI/CPE Stats are Used

Since exclusionary statistics cannot adjust for the possibility of dropout, and does not take the number of contributors into account, any loci with alleles below the stochastic threshold cannot be used in the CPI statistic.

# If CPI/CPE Stats are Used (ST = 200 RFU)



# Shakespeare on Allelic Drop-Out

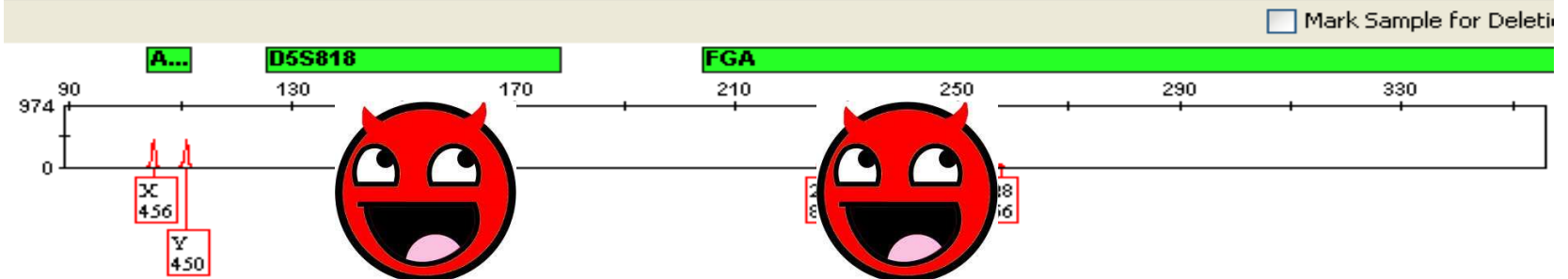
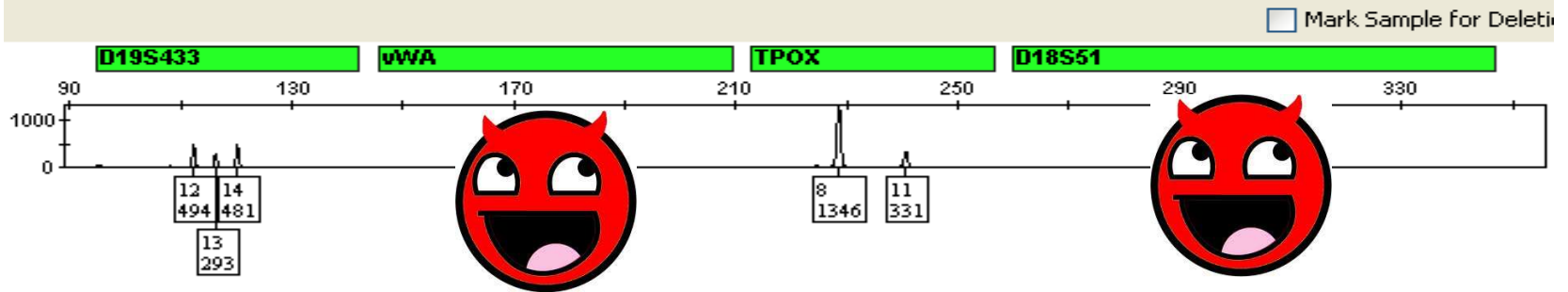
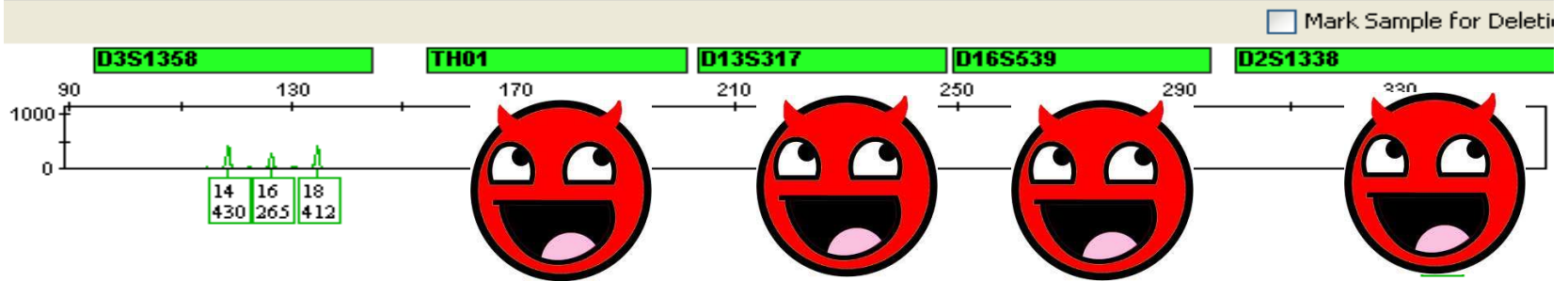
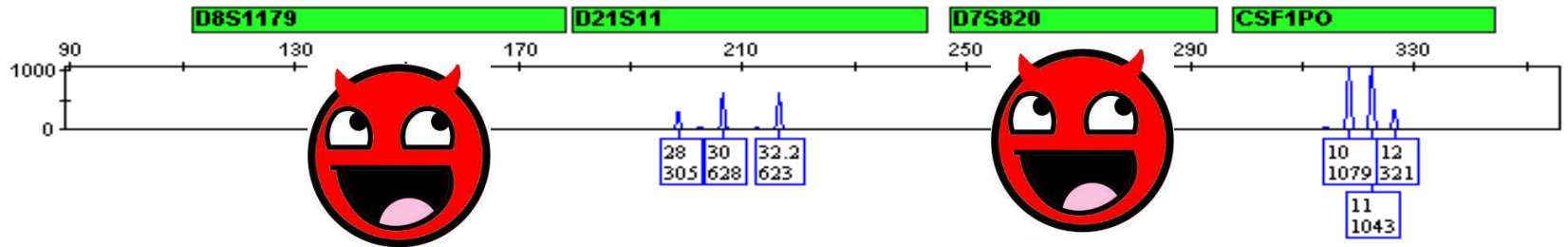


“Hell is empty and all the devils are here.”

— William Shakespeare,  
*The Tempest*



# If CPI/CPE Stats are Used



# If CPI/CPE Stats are Used

## Can use

D21

CSF

D3

D19

TPOX

## Cannot use

D8

D2

D7

vWA

TH01

D18

D13

D5

D16

FGA

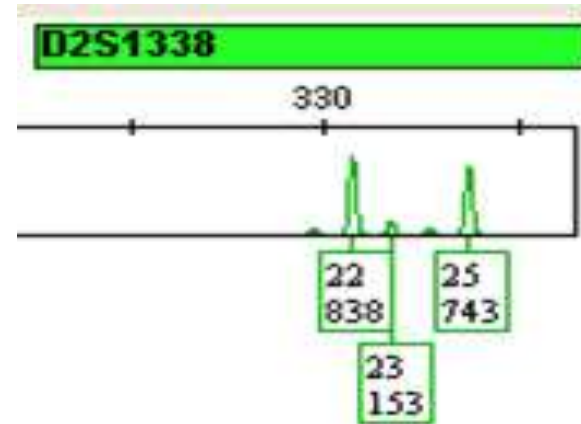
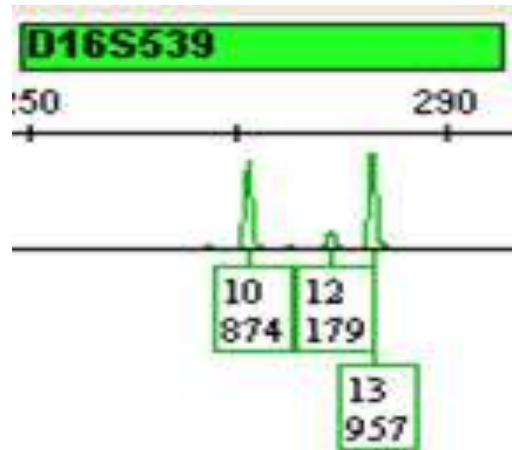
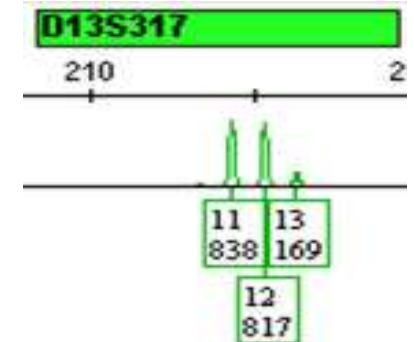
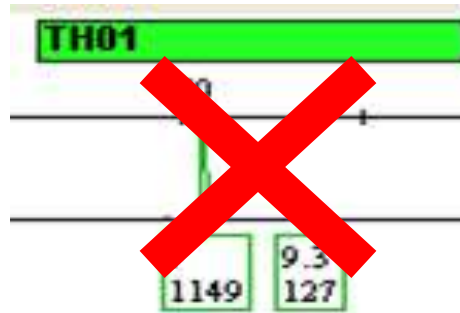
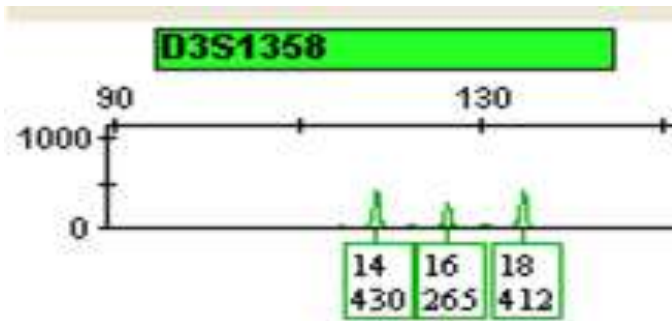
Impact: discarding 2/3 of the data



# If CPI/CPE Stats are Used

- CPI statistics using FBI Caucasian Frequencies
- 1 in 71 Caucasians included
- 98.59% Caucasians excluded

# If CPI/CPE Stats are Used (ST = 150 RFU)

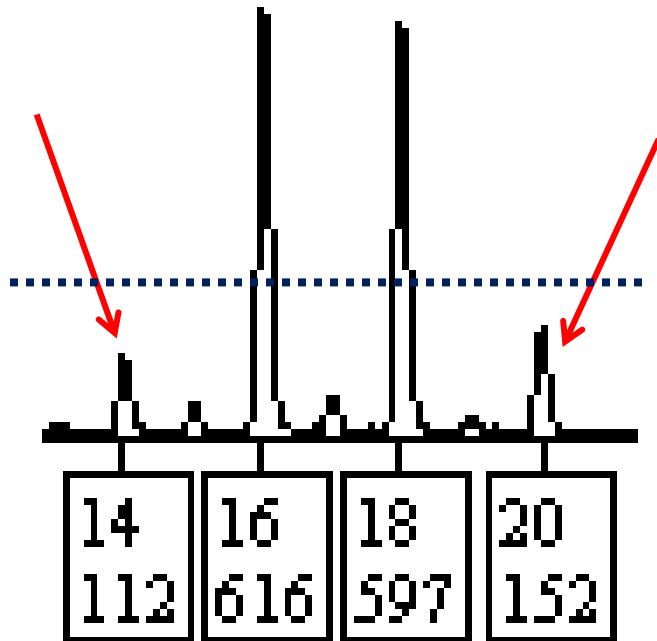


The impact of changing thresholds

# If mRMP/LR Stats are Used

- Since there is an assumption to the number of contributors, it is possible to use data that falls below the ST.

# mRMP - D18S51



If Assume 2 Contributors....

Major

16,18

Minor

14,20

$$\text{mRMP}_{\text{minor}} = 2pq$$

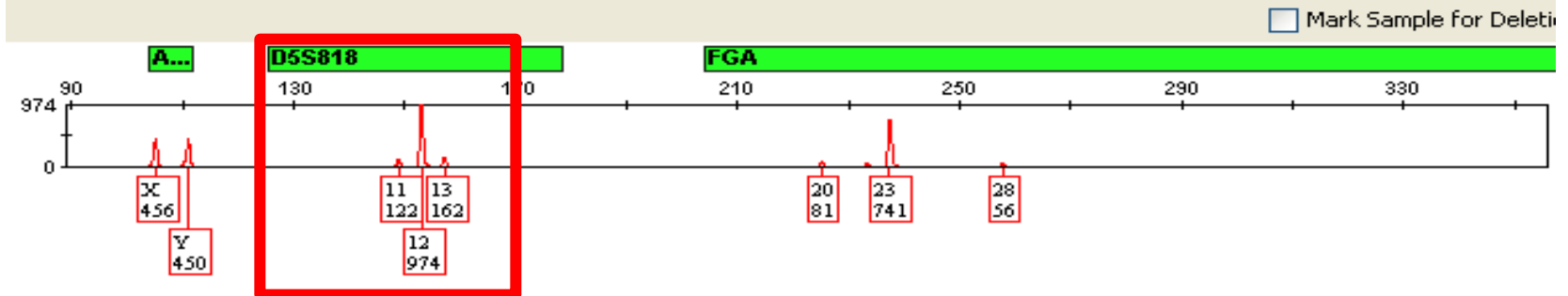
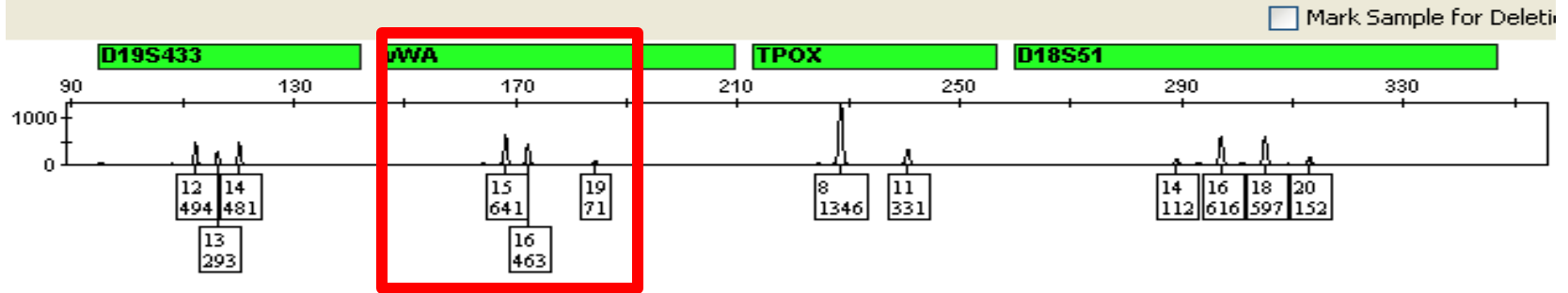
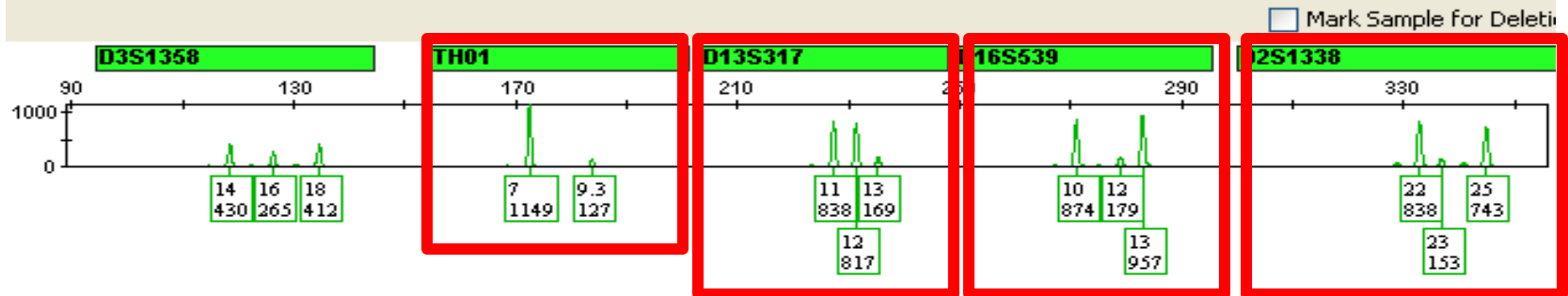
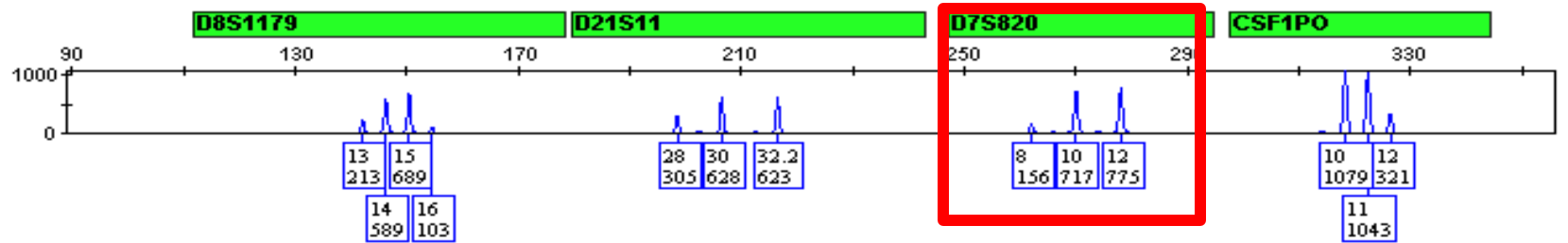
$$= 2 \times f(14) \times f(20)$$

$$= 2 \times (0.1735) \times (0.0255)$$

$$= 0.00884 \quad \text{or 1 in 113}$$

(LR = 113)

# Potential for Drop-out



# If mRMP/LR Stats are Used

## Can use

D8

D21

D18

D3

D19

TPOX

FGA

CSF

## Loci with potential D-out

D7

D2

TH01

vWA

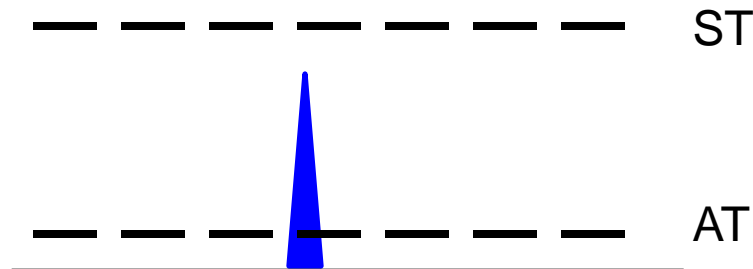
D13

D5

D16

# The “2p” Rule

- The “2p” rule can be used to statistically account for zygosity ambiguity – i.e. is this single peak below the stochastic threshold the result of a homozygous genotype or the result of a heterozygous genotype with allele drop-out of the sister allele?



“2p” or not “2p”... That is the question.

## Shakespeare on “2p”



“Drink sir, is a great provoker  
of three things....nose painting,  
sleep and urine.”

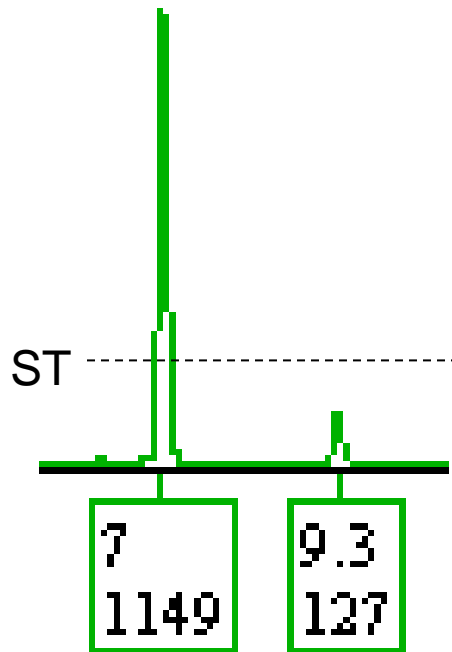
— William Shakespeare, *Macbeth*



# 2p – SWGDAM Guidelines

- 5.2.1.3.1. The formula  $2p$ , as described in recommendation 4.1 of NRCII, may be applied to this result.
- 5.2.1.3.2. Instead of using  $2p$ , the algebraically identical formulae  $2p - p^2$  and  $p^2 + 2p(1-p)$  may be used to address this situation without double-counting the proportion of homozygotes in the population.

# Macbeth/Duncan Profile - TH01



Major – 7, 7

Possible Minor Contributors

7, 9.3       $(2pq)$

9.3, 9.3       $p^2$

9.3, ?       $2p$  (or  $p^2 + 2p(1 - p)$ )

# Macbeth/Duncan Profile - TH01

$$\frac{P(E | H_1)}{P(E | H_2)} = \frac{V \ \& \ S}{V \ \& \ U} = \frac{f_7^2 + f_7(1-f_7)\theta \ \& \ 1}{f_7^2 + f_7(1-f_7)\theta \ \& \ 2p}$$

$$p^2 + 2p(1-p)$$

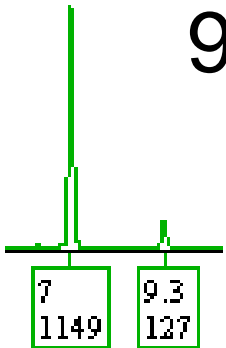
$$V = 7, 7$$

$$U = 7, 9.3$$

$$9.3, 9.3$$

$$9.3, ?$$

$$= \frac{1}{f_{9.3}^2 + 2f_{9.3}(1-f_{9.3})}$$



$$f_{9.3} = 0.3054$$

$$= 1 / 0.5175 = 1.93$$

# Macbeth/Duncan Profile - TH01

$$\frac{P(E | H_1)}{P(E | H_2)} = \frac{V \ \& \ S}{V \ \& \ U} = \frac{1}{p^2 + p(1-p)\theta + 2pq}$$

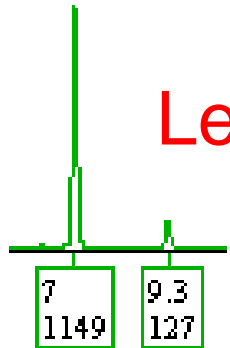
$$V = 7, 7$$

$$U = 7, 9.3$$

$$9.3, 9.3$$

$$= \frac{1}{f_{9.3}^2 + f_{9.3}(1-f_{9.3})\theta + 2f_{9.3}f_7}$$

Let  $ST = 125$  RFU



$$f_{9.3} = 0.3054$$

$$f_7 = 0.1724$$

$$= 1 / 0.2007 = 4.98$$

# Macbeth/Duncan Profile - TH01

	<u>LR</u>
ST = 200 (2p is used)	1.93
ST = 125 (2pq is used)	4.98

2p is conservative...

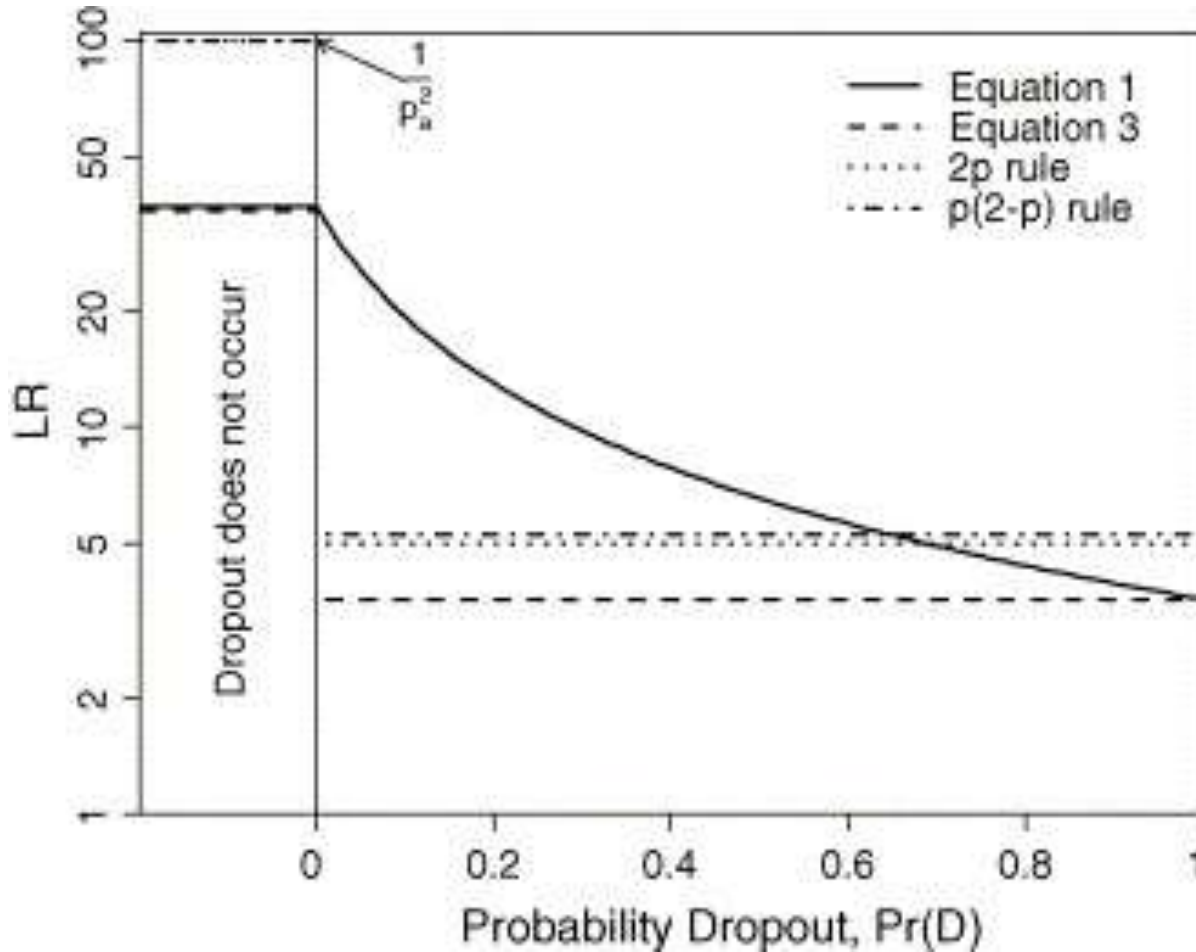
# The “2p” Rule

- “This rule arose during the VNTR era. At that time many smaller alleles “ran off the end of the gel” and were not visualised.”

- Buckleton and Triggs (2006)

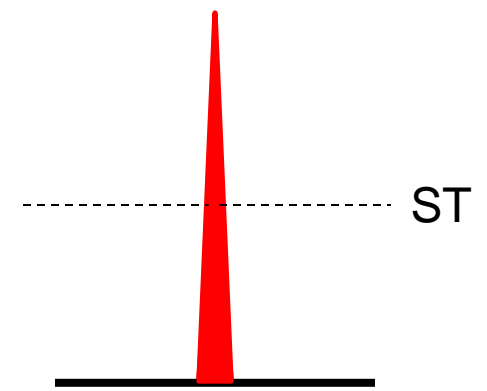
“Is the  $2p$  rule always conservative?”

# The “2p” Rule



Stain = AA

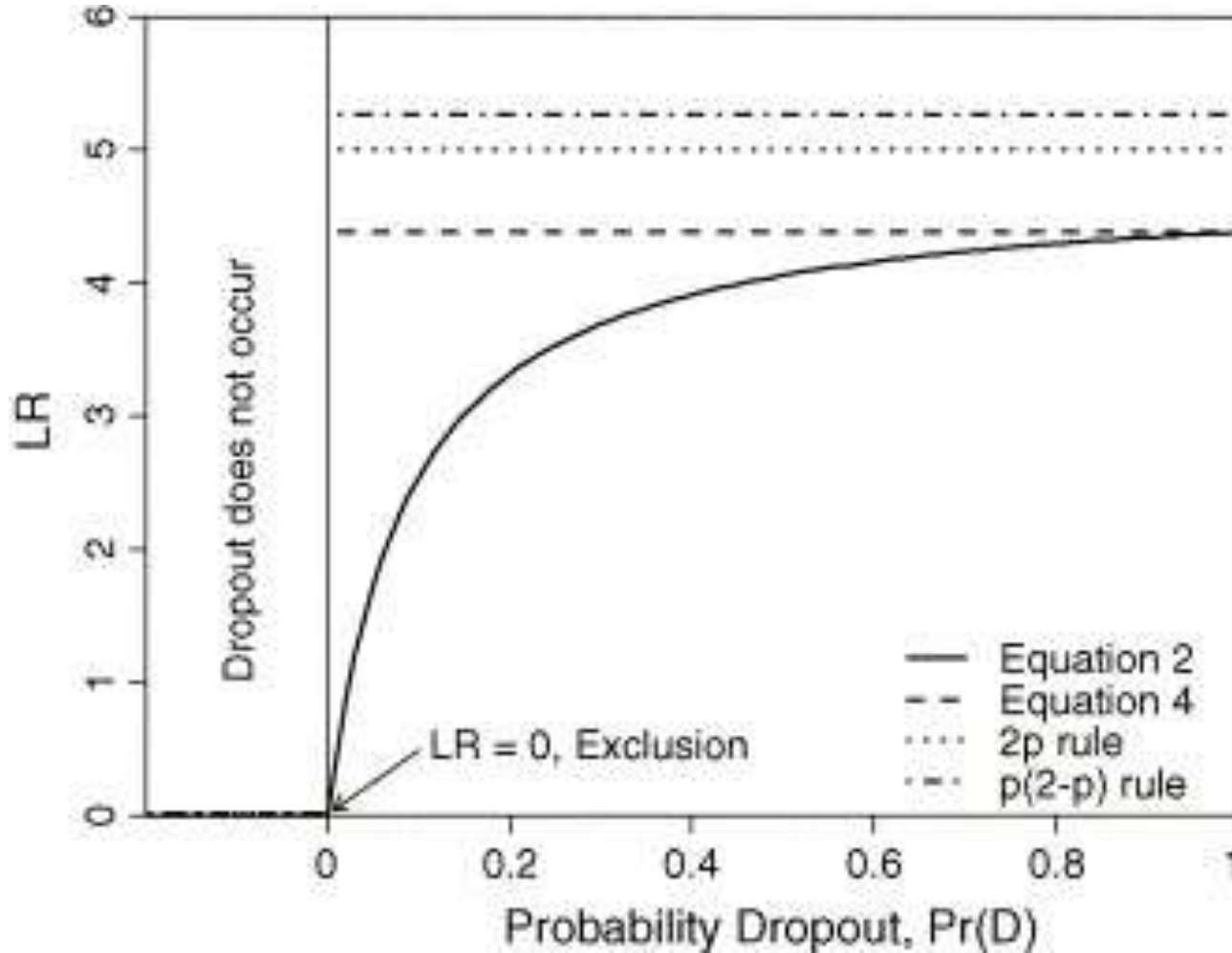
Suspect = AA



LR = 100

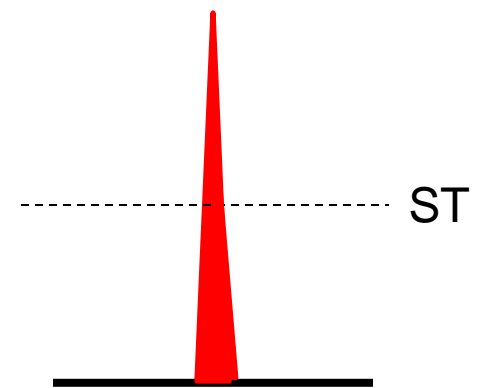
$$f(a) = 0.10 \quad 1/p^2 = 100 \quad 1/2p = 5$$

# The "2p" Rule



Stain = AA

Suspect = AB



Exclusion

$$f(a) = 0.10 \quad 1/2p = 5$$

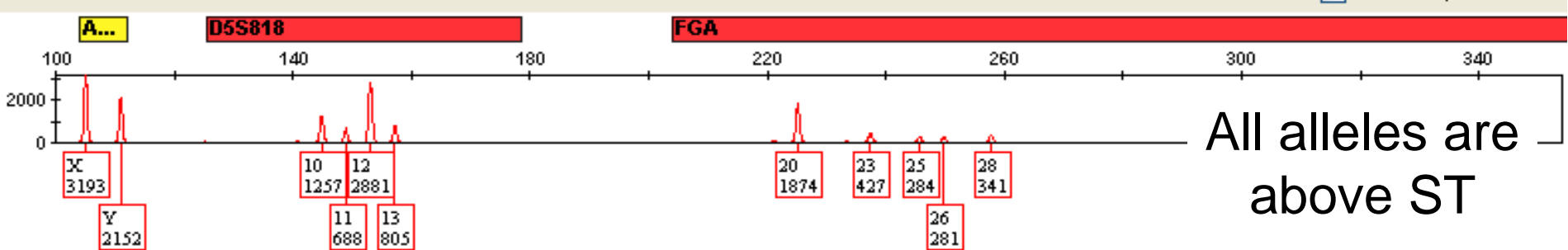
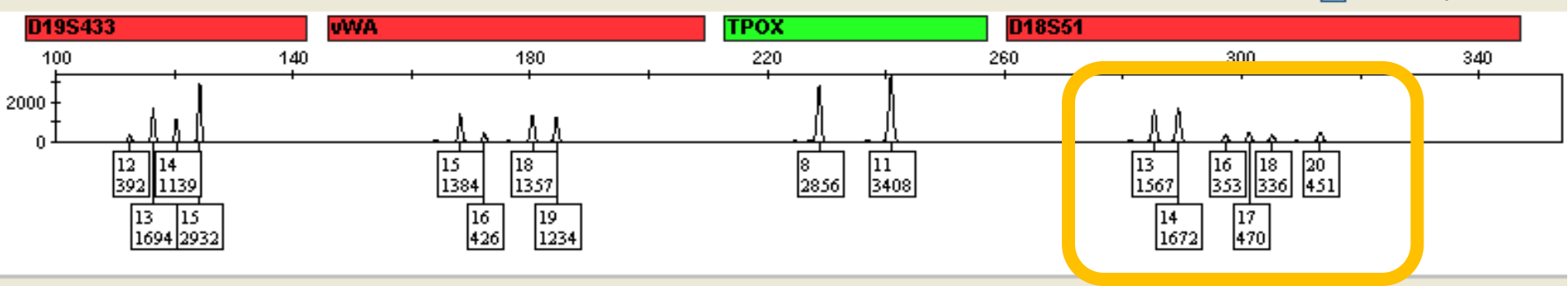
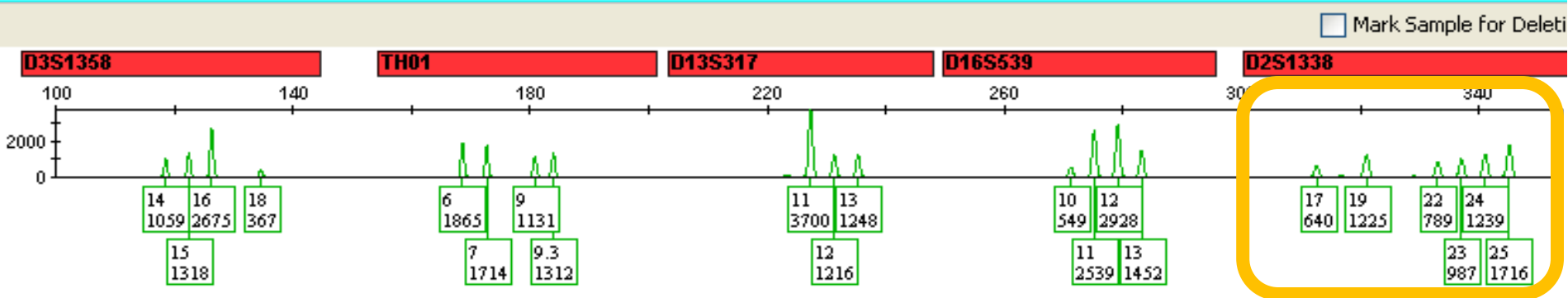
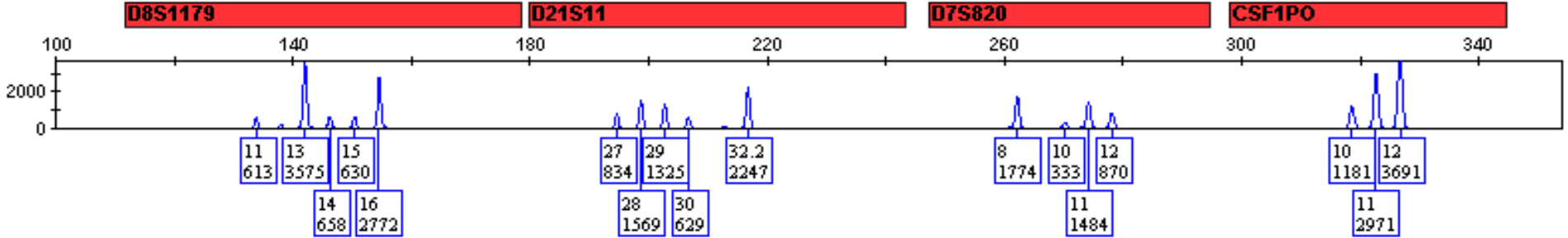


# Three Questions

- According to William Shakespeare, what were the last words of Julius Caesar before he died?
- Et tu, Brute?
- What is the capital of Bangladesh?
- Dhaka

# Three Questions

- How many people are in this mixture?



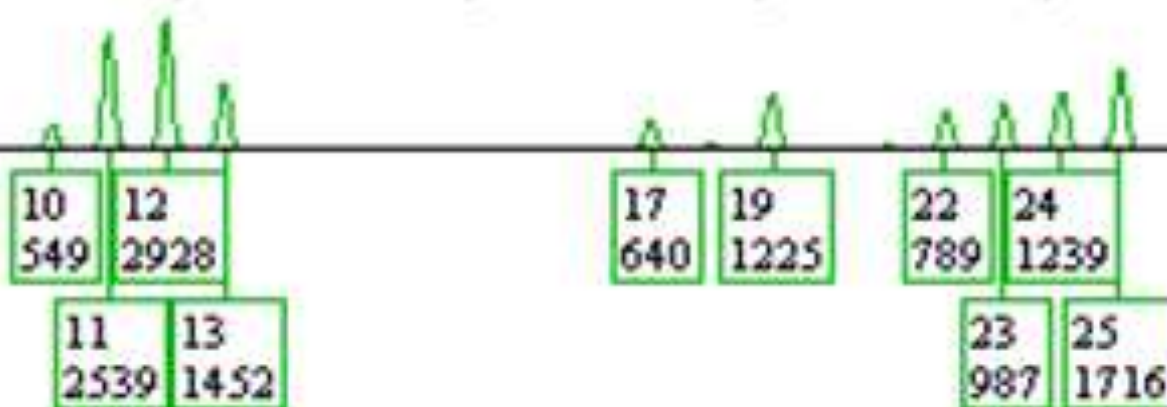
All alleles are above ST

**D16S539****D2S1338**

260

300

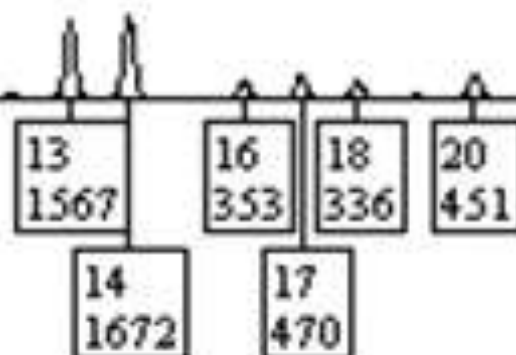
340

 Mark Sample for Deletion**D18S51**

260

300

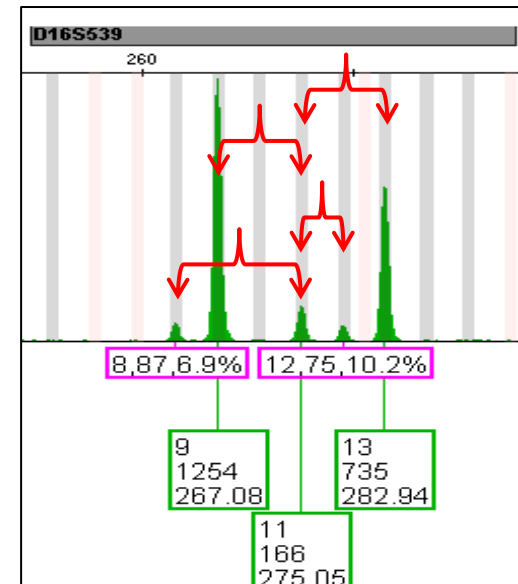
340



# Do You Have Uncertainty in Your Data?

- **If allele dropout is a possibility** (e.g., in a partial profile), then there is uncertainty in whether or not an allele is present in the sample...and therefore what genotype combinations are possible
- **If different allele combinations are possible** in a mixture, then there is uncertainty in the genotype combinations that are possible...

Possible allele pairing  
with the 11



# Uncertainty and Probability

- “Contrary to what many people think, **uncertainty is present throughout any scientific procedure.**”
  - Dennis V. Lindley, in his foreword to Aitken & Taroni (2004) *Statistics and the Evaluation of Evidence for Forensic Scientists, Second Edition*
- “It is now recognized that **the only tool for handling uncertainty is probability.**”
  - Dennis V. Lindley, in his foreword to Aitken & Taroni (2004) *Statistics and the Evaluation of Evidence for Forensic Scientists, Second Edition*

Is there a way forward?

# Next Issue of FSI-Genetics

Forensic Science International: Genetics xxx (2012) xxx–xxx



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Forensic Science International: Genetics

journal homepage: [www.elsevier.com/locate/fsig](http://www.elsevier.com/locate/fsig)



Editorial

Focus issue—Analysis and biostatistical interpretation of complex and low template DNA samples

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# Article in press...



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Forensic Science International: Genetics

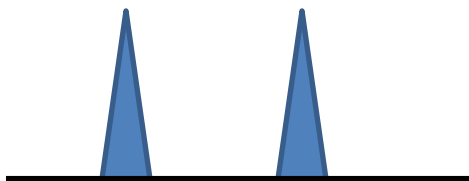
journal homepage: [www.elsevier.com/locate/fsig](http://www.elsevier.com/locate/fsig)



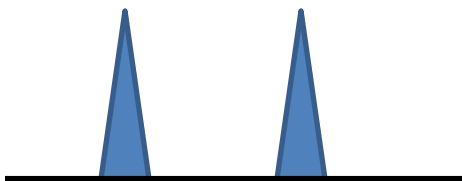
DNA commission of the International Society of Forensic Genetics:  
Recommendations on the evaluation of STR typing results that may  
include drop-out and/or drop-in using probabilistic methods

P. Gill<sup>a,b,\*</sup>, L. Gusmão<sup>c</sup>, H. Haned<sup>d</sup>, W.R. Mayr<sup>e</sup>, N. Morling<sup>f</sup>, W. Parson<sup>g</sup>, L. Prieto<sup>h</sup>,  
M. Prinz<sup>i</sup>, H. Schneider<sup>j</sup>, P.M. Schneider<sup>k</sup>, B.S. Weir<sup>l</sup>

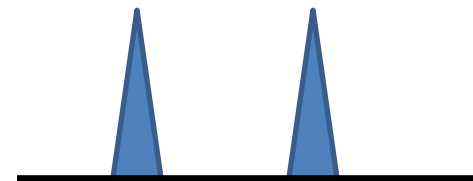
Suspect



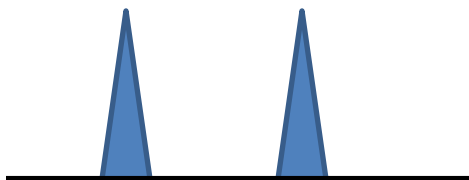
Suspect



Suspect



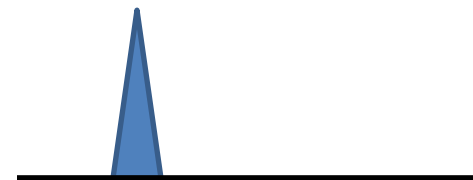
Evidence



Evidence



Evidence



$$LR = \frac{1}{2pq}$$

$$LR = \frac{0}{2pq}$$

$$LR = \frac{?}{2pq}$$

“2p”

$$p^2 + 2p(1 - p)$$

Whatever way uncertainty is approached, probability is the *only* sound way to think about it.



-Dennis Lindley

# Haned *et al.*

Forensic Science International: Genetics xxx (2012) xxx–xxx



Contents lists available at SciVerse ScienceDirect

Forensic Science International: Genetics

journal homepage: [www.elsevier.com/locate/fsig](http://www.elsevier.com/locate/fsig)



## Exploratory data analysis for the interpretation of low template DNA mixtures

H. Haned<sup>a,\*</sup>, K. Slooten<sup>a,b</sup>, P. Gill<sup>c,d</sup>

<sup>a</sup> Netherlands Forensic Institute, Department of Human Biological traces, The Hague, The Netherlands

<sup>b</sup> VU University Amsterdam, Amsterdam, The Netherlands

<sup>c</sup> Norwegian institute of Public Health, Oslo, Norway

<sup>d</sup> University of Oslo, Norway

# Mitchell *et al.*

Forensic Science International: Genetics xxx (2012) xxx–xxx



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## Validation of a DNA mixture statistics tool incorporating allelic drop-out and drop-in

Adele A. Mitchell\*, Jeannie Tamariz, Kathleen O'Connell, Nubia Ducasse, Zoran Budimlija, Mechthild Prinz, Theresa Caragine

Department of Forensic Biology, Office of Chief Medical Examiner of The City of New York, 421 E 26th Street, New York, NY 10016, United States

PAPER

*J Forensic Sci*, 2011  
doi: 10.1111/j.1556-4029.2011.01859.x  
Available online at: [onlinelibrary.wiley.com](http://onlinelibrary.wiley.com)

## CRIMINALISTICS

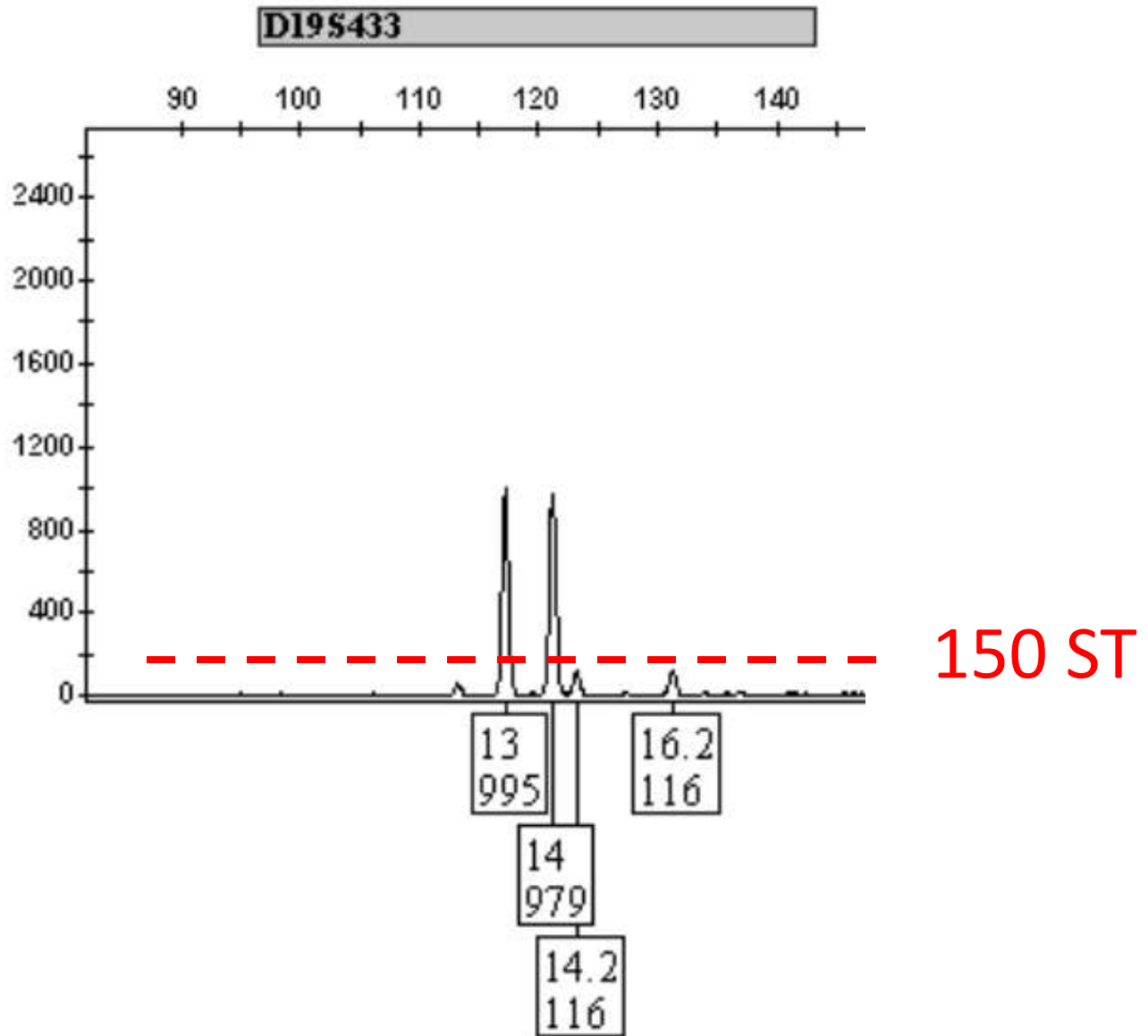
*Mark W. Perlin,<sup>1</sup> M.D., Ph.D.; Matthew M. Legler,<sup>1</sup> B.S.; Cara E. Spencer,<sup>1</sup> M.S.; Jessica L. Smith,<sup>1</sup> M.S.; William P. Allan,<sup>1</sup> M.S.; Jamie L. Belrose,<sup>2</sup> M.S.; and Barry W. Ducean,<sup>3</sup> Ph.D.*

Validating TrueAllele<sup>®</sup> DNA Mixture Interpretation<sup>\*,†</sup>

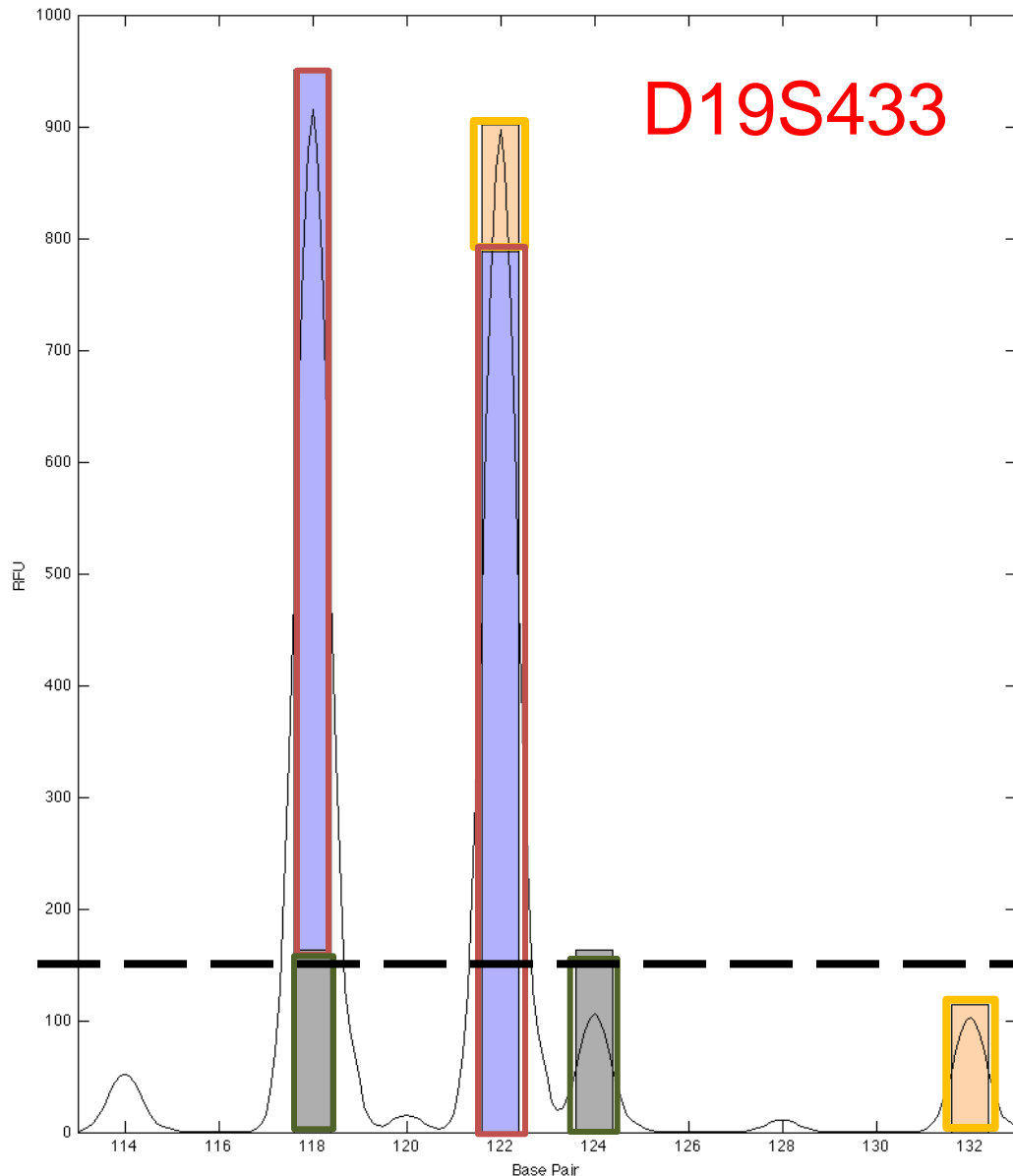
- Quantitative computer interpretation using Markov Chain Monte Carlo testing
- Models peak uncertainty and infers possible genotypes
- Results are presented as the Combined LR



# 3 Person Mixture



# Review of One Replicate (of 50K)



D19S433

3P mixture,  
2 Unknowns,

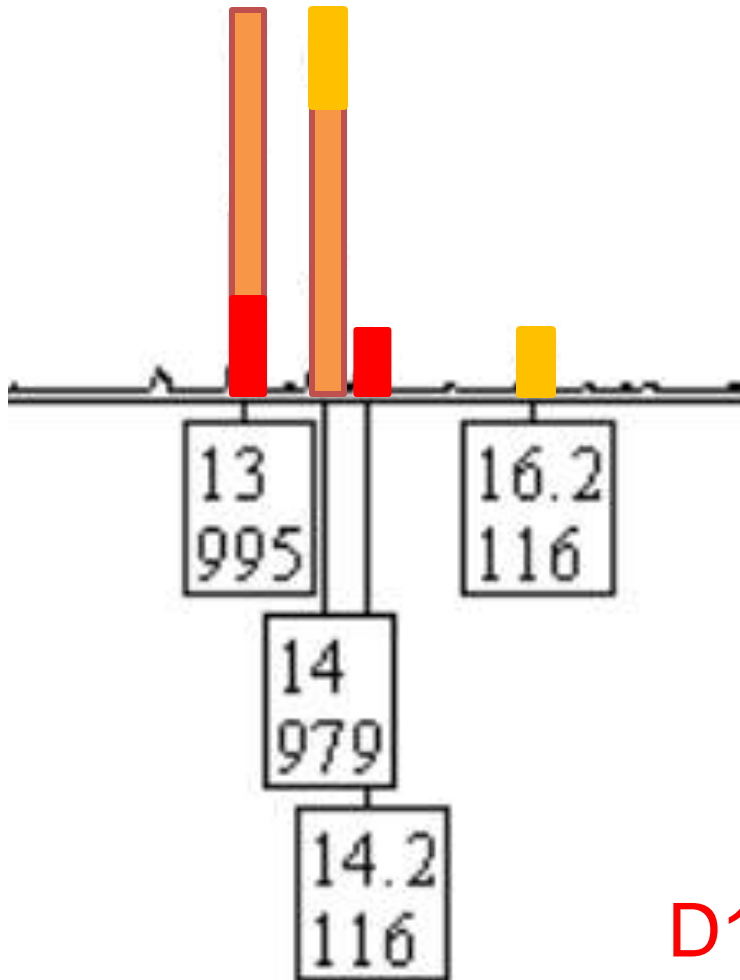
Conditioned  
on the Victim  
(major)

Good fit of the  
data to the model

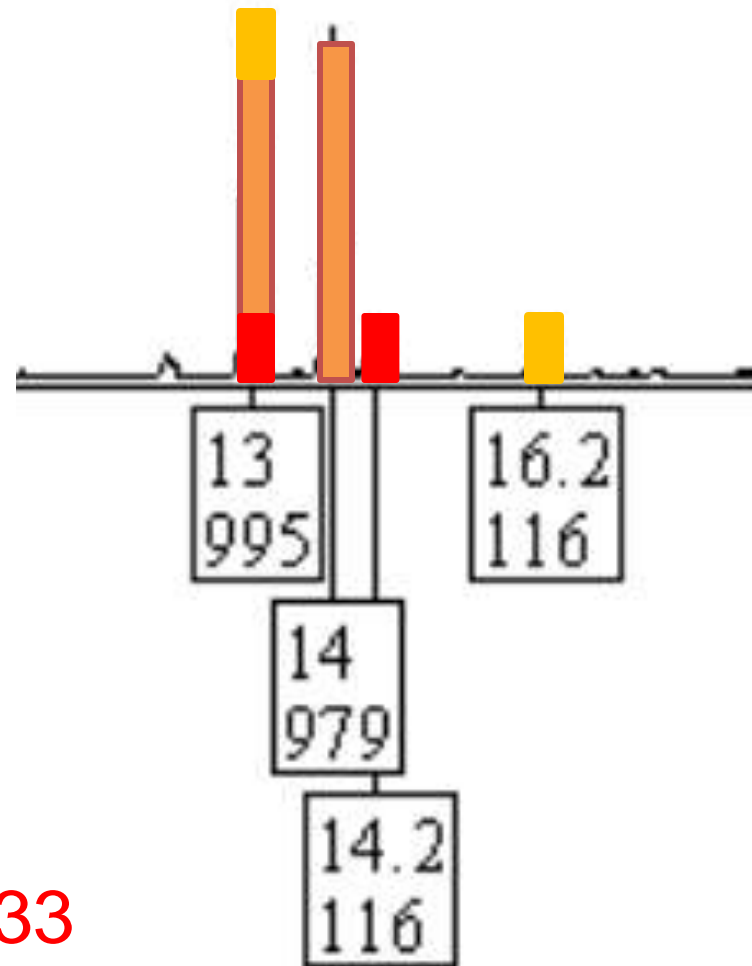
150 RFU



# Alternative Explanations of the Data

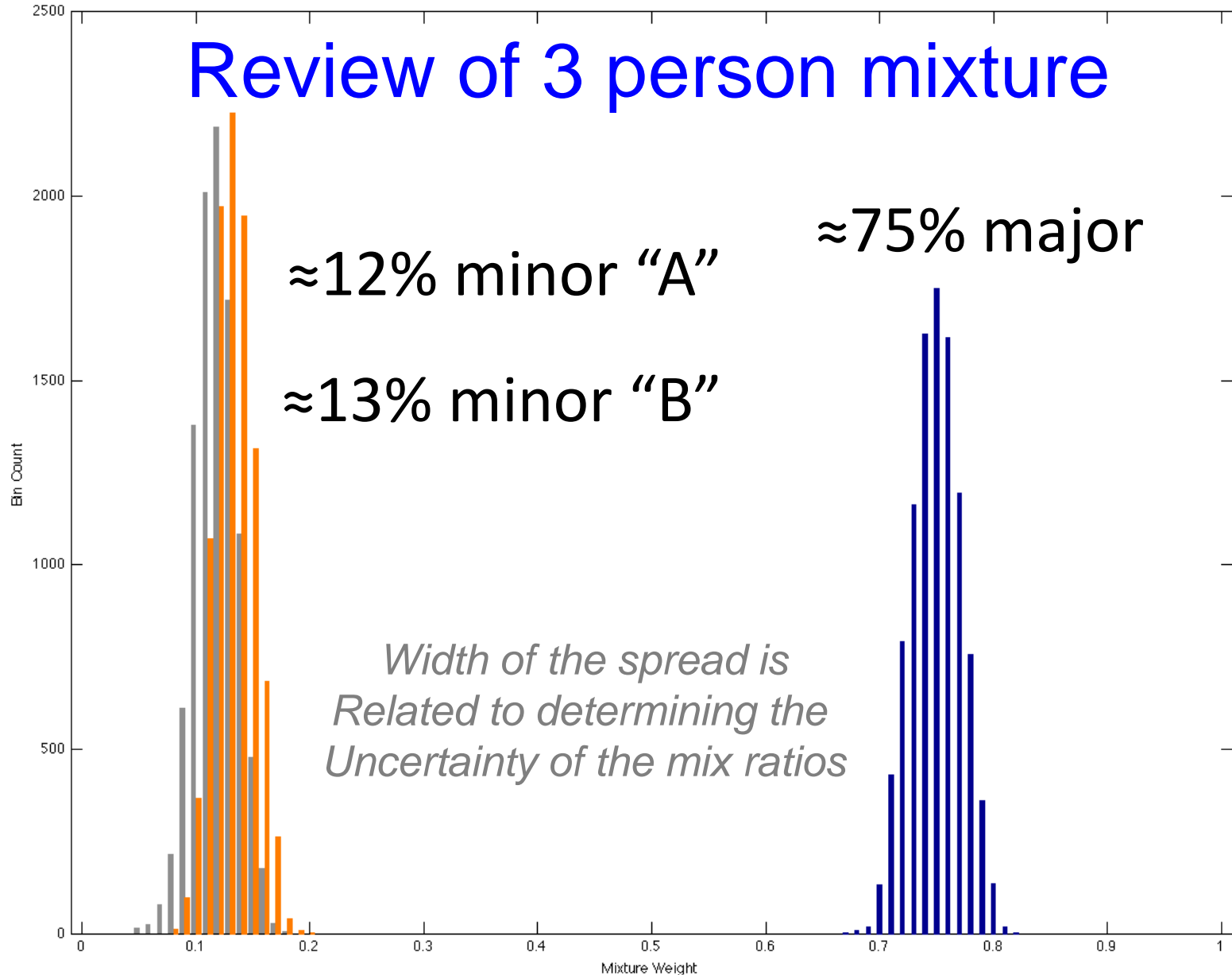


D19S433



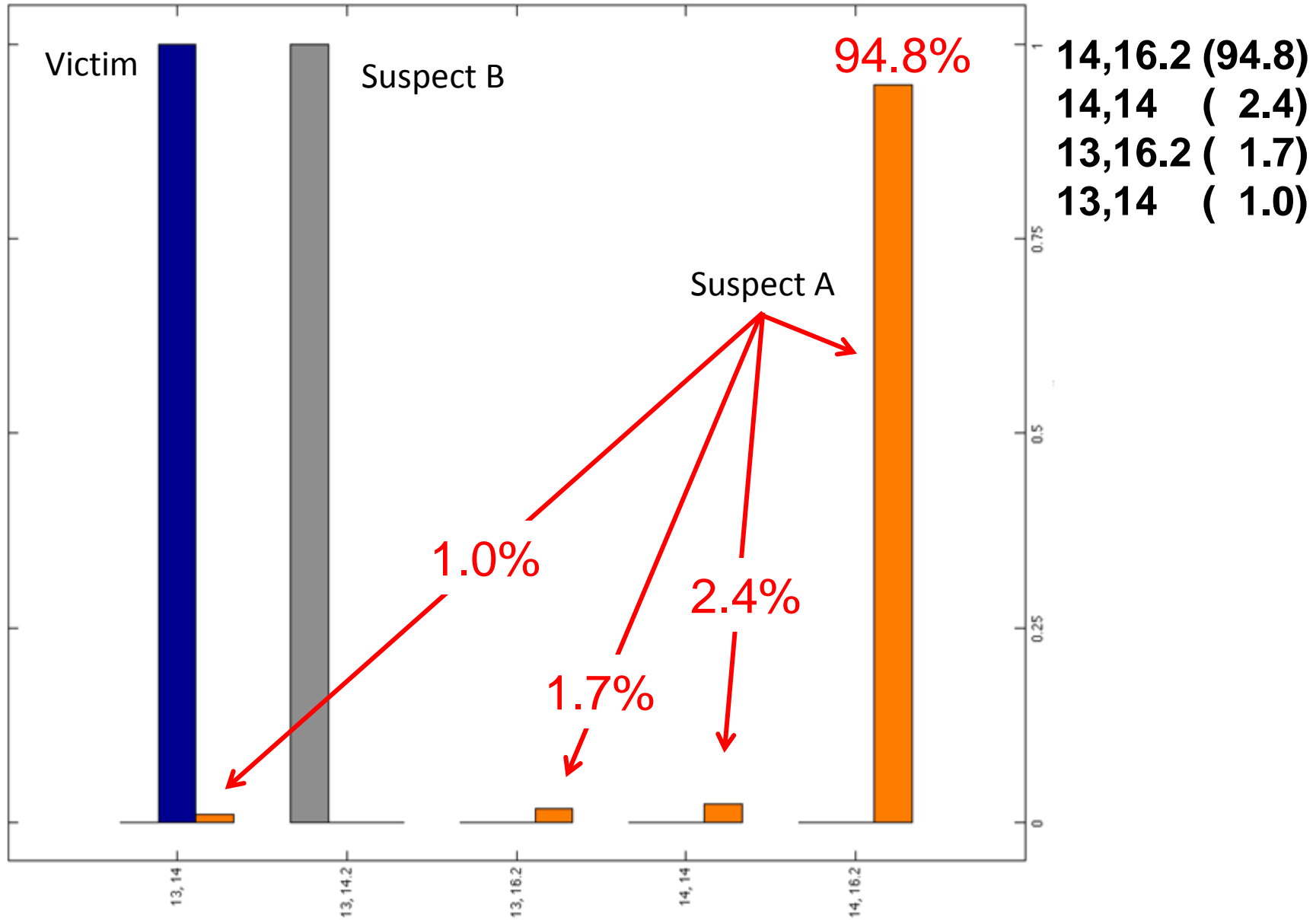
# Review of 3 person mixture

Bin Count



Mixture Weight

Genotype Probability



D19S433

Genotypes

# Determining the LR for D19S433

Suspect A = 14, 16.2

$$H_p = 0.967$$

	Allele Pair	Probability Before Conditioning
→	14, 16.2	0.967
	14, 14	0.003
	13, 16.2	0.026
	13, 14	0.001

$$LR = \frac{0.967}{\quad}$$

# Determining the LR for D19S433

Suspect A = 14, 16.2

$H_P = 0.967$

Allele Pair	Probability Before Conditioning	Genotype Frequency	Probability * Genotype Freq
14, 16.2	0.967	0.0120	0.01164
14, 14	0.003	0.0498	0.00013
13, 16.2	0.026	0.0131	0.00034
13, 14	0.001	0.1082	0.00009
		sum	0.0122

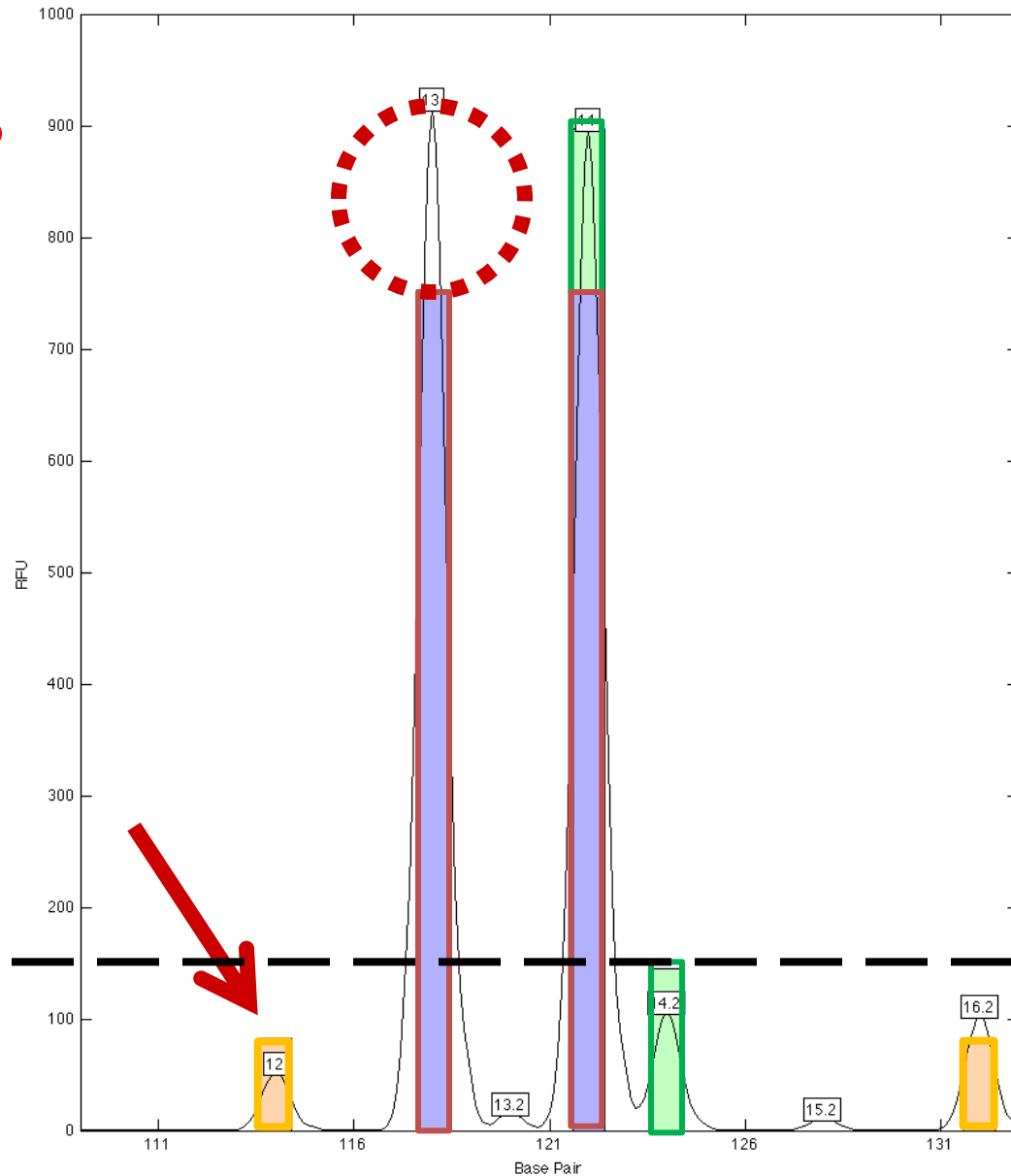
$$LR = \frac{0.967}{0.0122} = 79.26 \quad H_D$$

# Combined LR = 5.6 Quintillion

locus	allele pair x	Likelihood l(x)	Genotype Probability Distribution			Weighted Likelihood		Likelihood Ratio	
			Questioned q(x)	Reference r(x)	Suspect s(x)	Numerator l(x)*s(x)	Denominator l(x)*r(x)	LR	log(LR)
CSF1PO	11, 12	0.686	0.778	0.1448	1	0.68615	0.1292	5.31	0.725
D13S317	9, 12	1	1	0.0291	1	0.99952	0.02913	34.301	1.535
D16S539	9, 11	0.985	0.995	0.1238	1	0.98451	0.12188	8.036	0.905
D18S51	13, 17	0.999	1	0.0154	1	0.99915	0.01543	64.677	1.811
D19S433	14, 16.2	0.967	0.948	0.012	1	0.96715	0.01222	79.143	1.898
D21S11	28, 30	0.968	0.98	0.0872	1	0.96809	0.08648	11.194	1.049
D2S1338	23, 24	0.998	1	0.0179	1	0.99831	0.01787	55.866	1.747
D3S1358	15, 17	0.988	0.994	0.1224	1	0.98759	0.12084	8.14	0.911
D5S818	11, 11	0.451	0.394	0.0537	1	0.45103	0.07309	6.17	0.79
D7S820	11, 12	0.984	0.978	0.0356	1	0.98383	0.03617	27.198	1.435
D8S1179	13, 14	0.203	0.9	0.1293	1	0.20267	0.02993	6.771	0.831
FGA	21, 25	0.32	0.356	0.028	1	0.31986	0.01906	16.783	1.225
TH01	7, 7	0.887	0.985	0.1739	1	0.88661	0.15588	5.687	0.755
TPOX	8, 8	1	1	0.1375	1	1	0.13746	7.275	0.862
vWA	15, 20	0.998	0.996	0.0057	1	0.99808	0.00569	174.834	2.243

# Review of One Replicate (of 50K)

D19S433



3P mixture,  
3 Unknowns

Poor fit of the  
data to the  
model

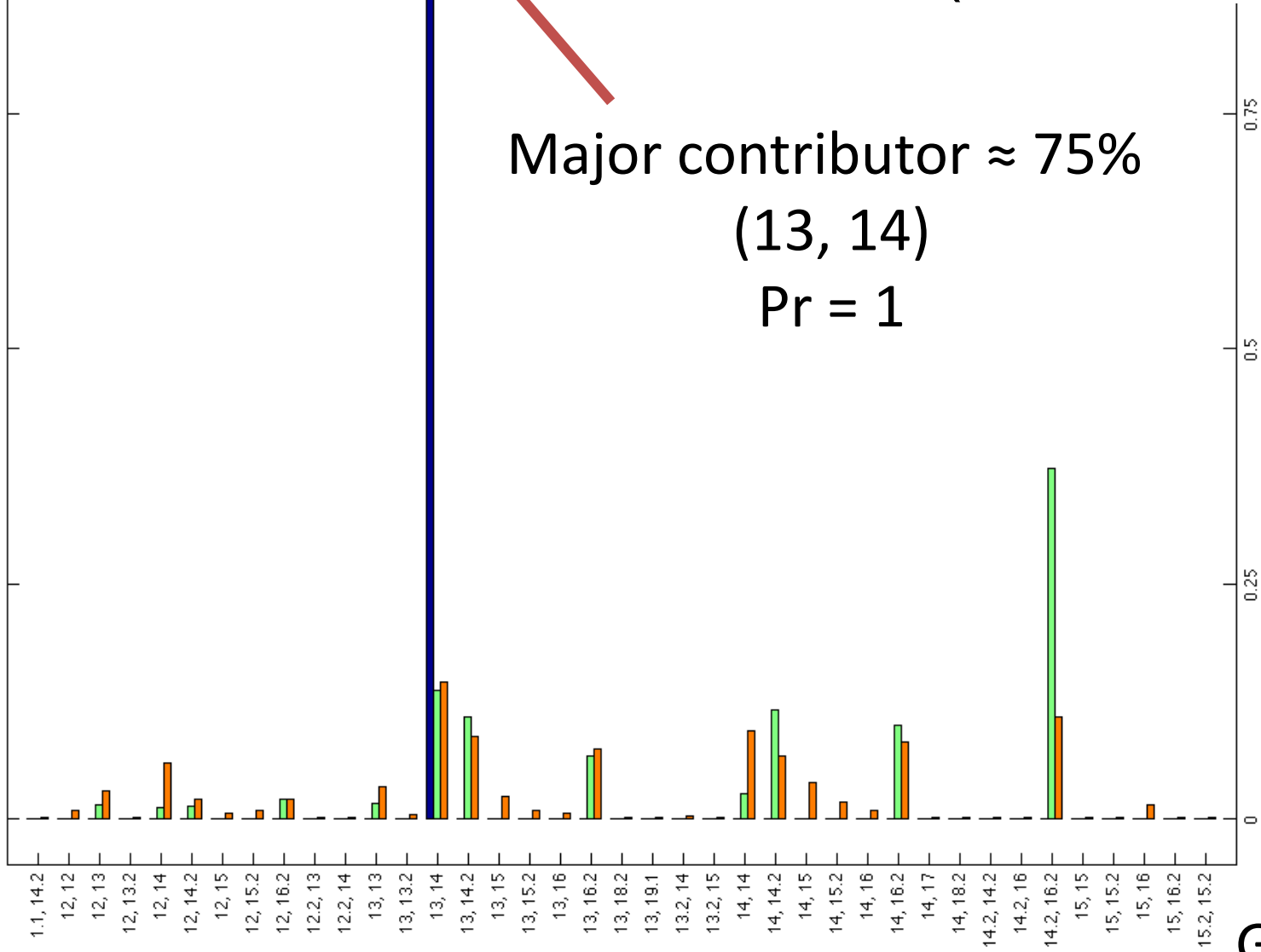
150 RFU

# No Conditioning (3 Unknowns)

D19S433

Genotype Probability

Major contributor  $\approx 75\%$   
(13, 14)  
Pr = 1



Genotypes

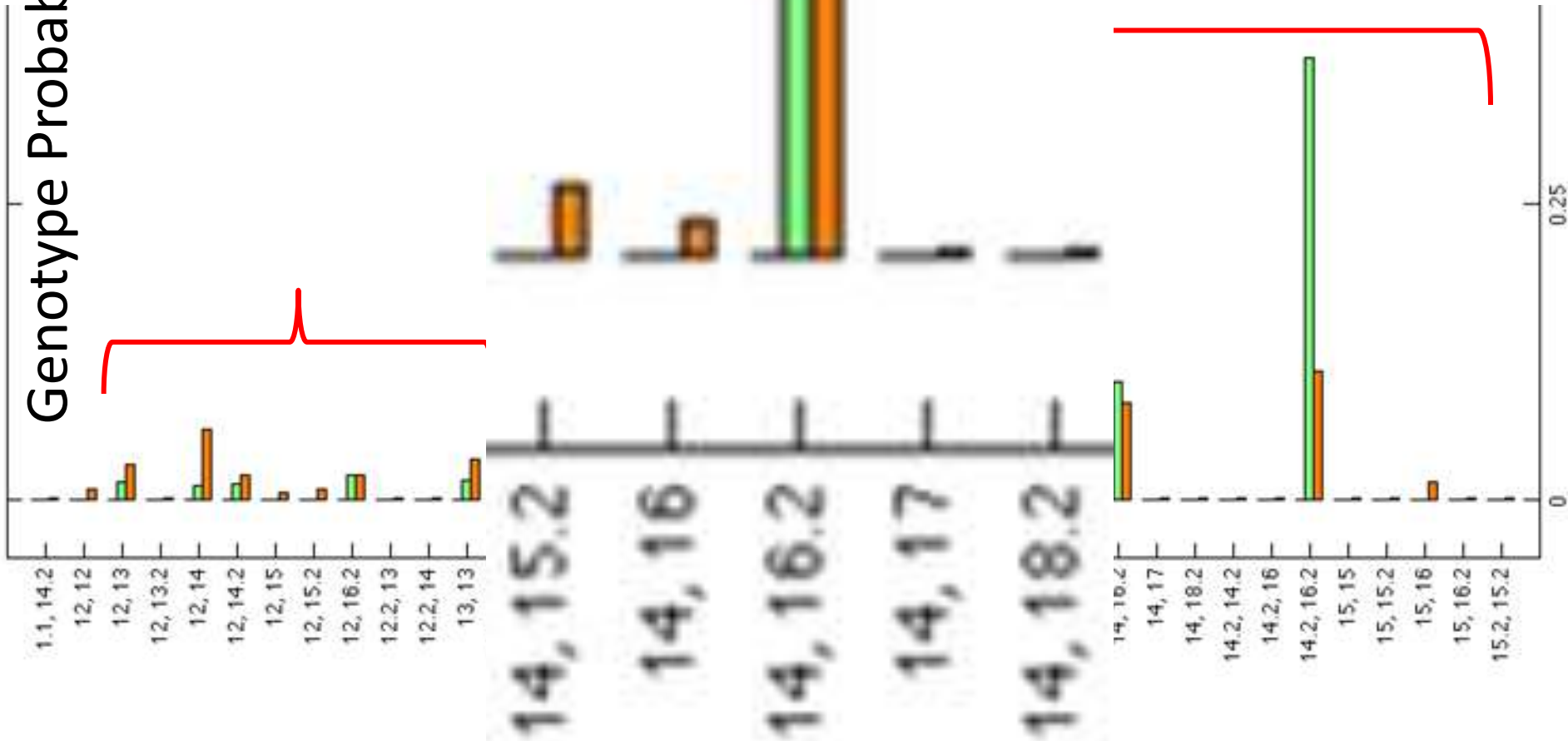


# No Conditioning (3 Unknowns)

D19S433

Genotype Probability

8.1% ns for the two contributors



locus	allele pair	L	Q	R	S	L*S	L*R	LR	log(LR)
D19S433	13 , 14	0.002	0.146	0.1082			0.00020		
	14.2, 16.2	0.270	0.109	0.0044			0.00118		
	14 , 14	0.002	0.093	0.0498			0.00008		
	13 , 14.2	0.017	0.088	0.0392			0.00068		
	14 , 16.2	0.013	0.081	0.0120	1	0.01295	0.00016		
	13 , 16.2	0.018	0.074	0.0131			0.00023		
	14 , 14.2	0.009	0.067	0.0361			0.00031		
	12 , 14	0.002	0.059	0.0498			0.00012		
	14 , 15	0.001	0.038	0.0343			0.00002		
	13 , 13	0.001	0.034	0.0587			0.00007		
	12 , 13	0.002	0.029	0.0541			0.00010		
	13 , 15	0.001	0.024	0.0373			0.00002		
	12 , 16.2	0.017	0.021	0.0060			0.00010		
	12 , 14.2	0.013	0.020	0.0180			0.00023		
	14 , 15.2	0.001	0.018	0.0275			0.00003		
	15 , 16	0.002	0.015	0.0006			0.00000		
	13 , 15.2	0.001	0.009	0.0299			0.00003		
	12 , 15.2	0.003	0.009	0.0137			0.00004		
	14 , 16	0.000	0.009	0.0017			0.00000		
	12 , 12	0.004	0.009	0.0125			0.00004		
	12 , 15	0.001	0.006	0.0172			0.00001		
	13 , 16	0.000	0.006	0.0019			0.00000		
	13 , 13.2	0.001	0.004	0.0261			0.00003		
	13.2, 14	0.001	0.003	0.0240			0.00002		
	13.2, 15	0.001	0.002	0.0083			0.00001		
	14 , 18.2	0.002	0.002	0.0017			0.00000		
	13 , 19.1	0.019	0.002	0.0000			0.00000		
	12 , 13.2	0.002	0.002	0.0120			0.00003		
	14.2, 16	0.001	0.002	0.0006			0.00000		
	12.2, 13	0.001	0.002	0.0168			0.00002		
	13 , 18.2	0.002	0.001	0.0019			0.00000		
	12.2, 14	0.001	0.001	0.0155			0.00001		
	14.2, 14.2	0.004	0.001	0.0065			0.00003		
	15 , 15	0.000	0.001	0.0059			0.00000		
	15 , 15.2	0.000	0.001	0.0095			0.00000		
	14 , 17	0.001	0.001	0.0000			0.00000		
	15 , 16.2	0.000	0.001	0.0042			0.00000		
	15.2, 15.2	0.001	0.001	0.0038			0.00000		
	1.1, 14.2	0.072	0.001	0.0097			0.00069		
						0.01295	0.00385	3.367	0.527

Suspect "A"  
Genotype

39 probable  
genotypes

D19S433

Suspect A = 14, 16.2

$$H_P = 0.013$$

Allele Pair	Probability	Genotype Frequency	Prob * GenFreq
13,14	0.002	0.1082	0.00020
14.2, 16.2	0.270	0.0044	0.00118
14, 14	0.002	0.0498	0.00008
13, 14.2	0.017	0.0392	0.00068
14, 16.2	0.013	0.0120	0.00016
13, 16.2	0.018	0.0131	0.00023
etc...	etc...	etc...	etc...
		<b>Sum</b>	<b>0.00385</b>

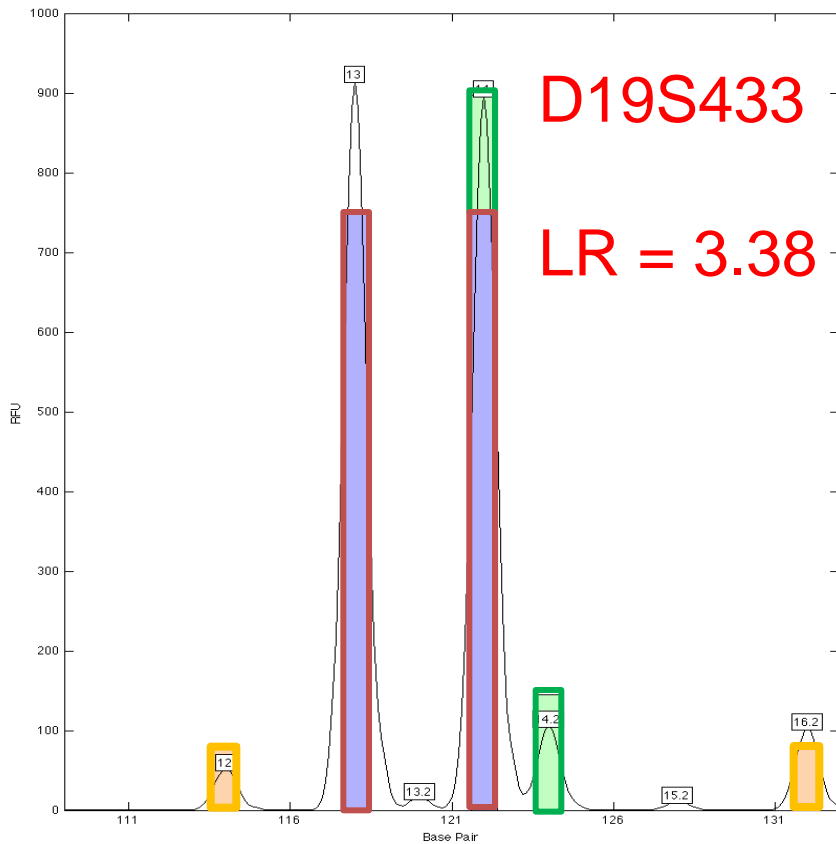
$$LR = \frac{0.013}{0.00385} = 3.38$$

$H_D$

D19S433

*No Conditioning (3 Unknowns)*

# No Conditioning

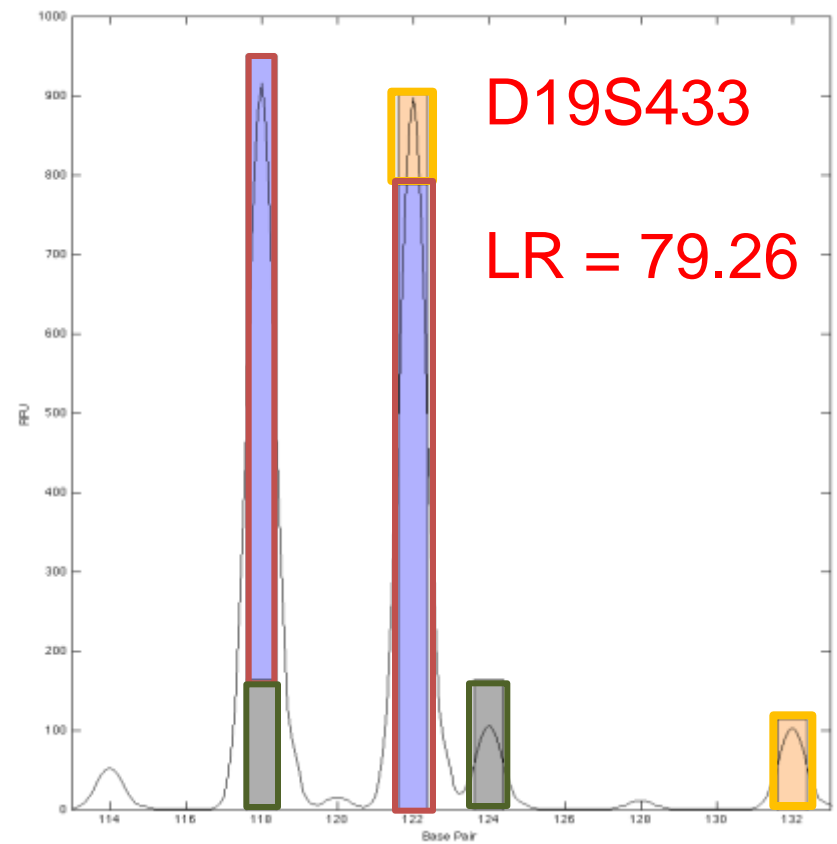


Profile - Combined  $\log(\text{LR})$

Suspect A  $\log(\text{LR}) = 8.03$

Suspect B  $\log(\text{LR}) = 7.84$

# Conditioned on Victim



Profile - Combined  $\log(\text{LR})$

Suspect A  $\log(\text{LR}) = 18.72$

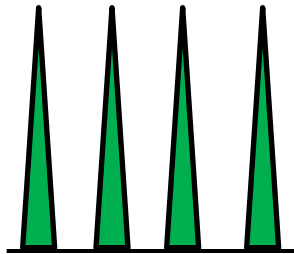
Suspect B  $\log(\text{LR}) = 19.45$

# Mixture Data Set

- Mixtures of pristine male and female DNA amplified at a total concentration of 1.0 ng/ $\mu$ L using Identifiler (standard conditions).
- Mixture ratios ranged from 90:10, 80:20, 70:30, 60:40, 50:50, 40:60, 30:70, 20:80, and 10:90
- Each sample was amplified twice.

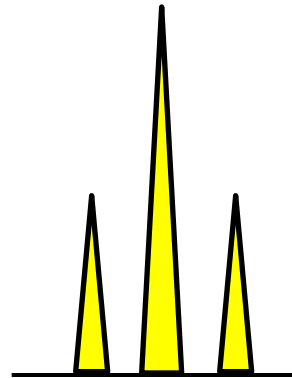
# Mixture Data Set

- Three different combinations:



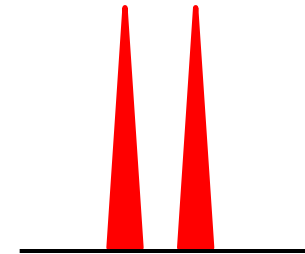
“Low” Sharing

4 alleles – 10 loci  
3 alleles – 5 loci  
2 alleles – 0 loci  
1 allele – 0 loci



“Medium” Sharing

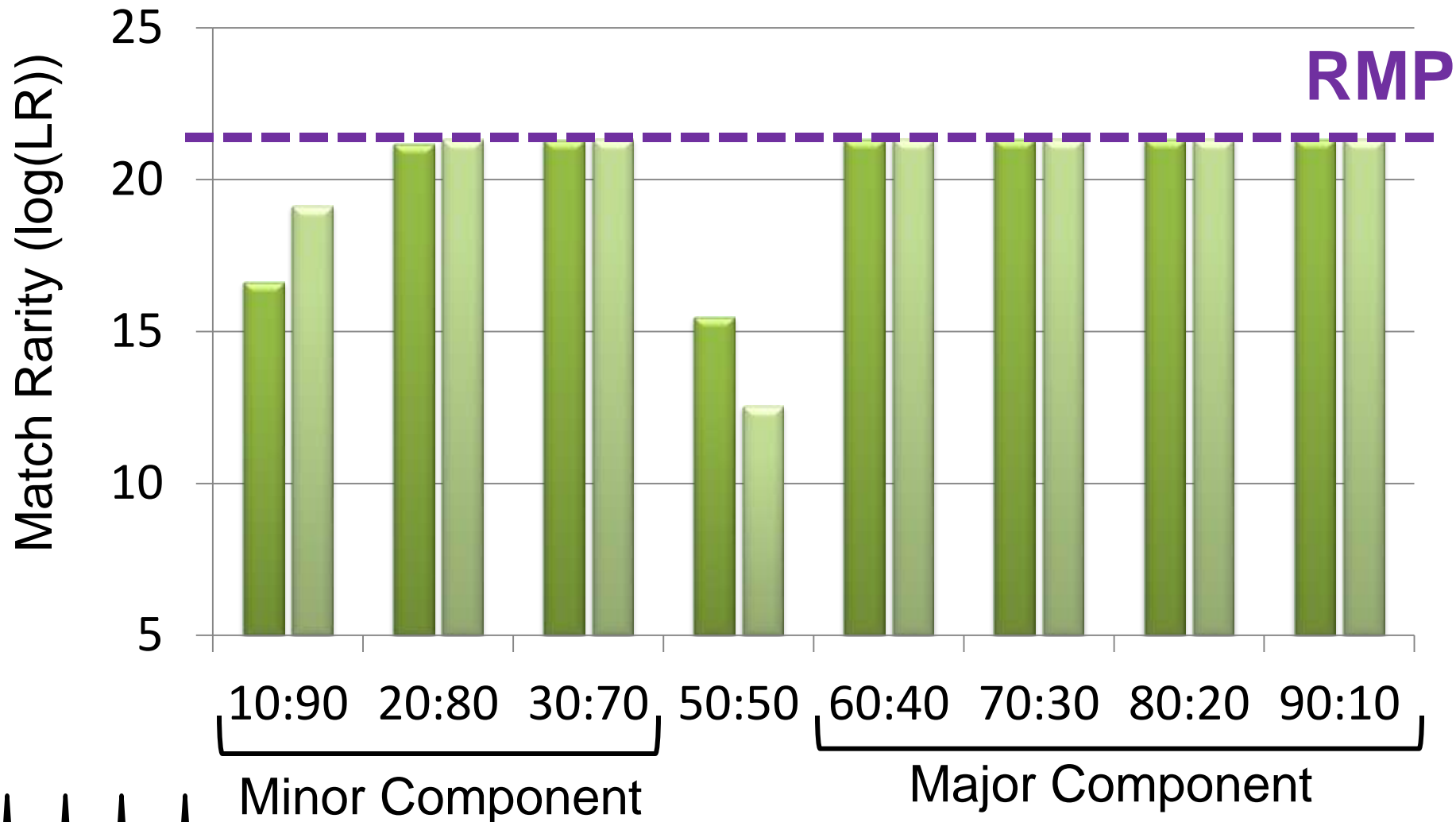
4 alleles – 3 loci  
3 alleles – 8 loci  
2 alleles – 4 loci  
1 allele – 0 loci



“High” Sharing

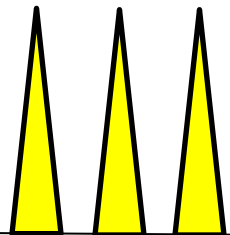
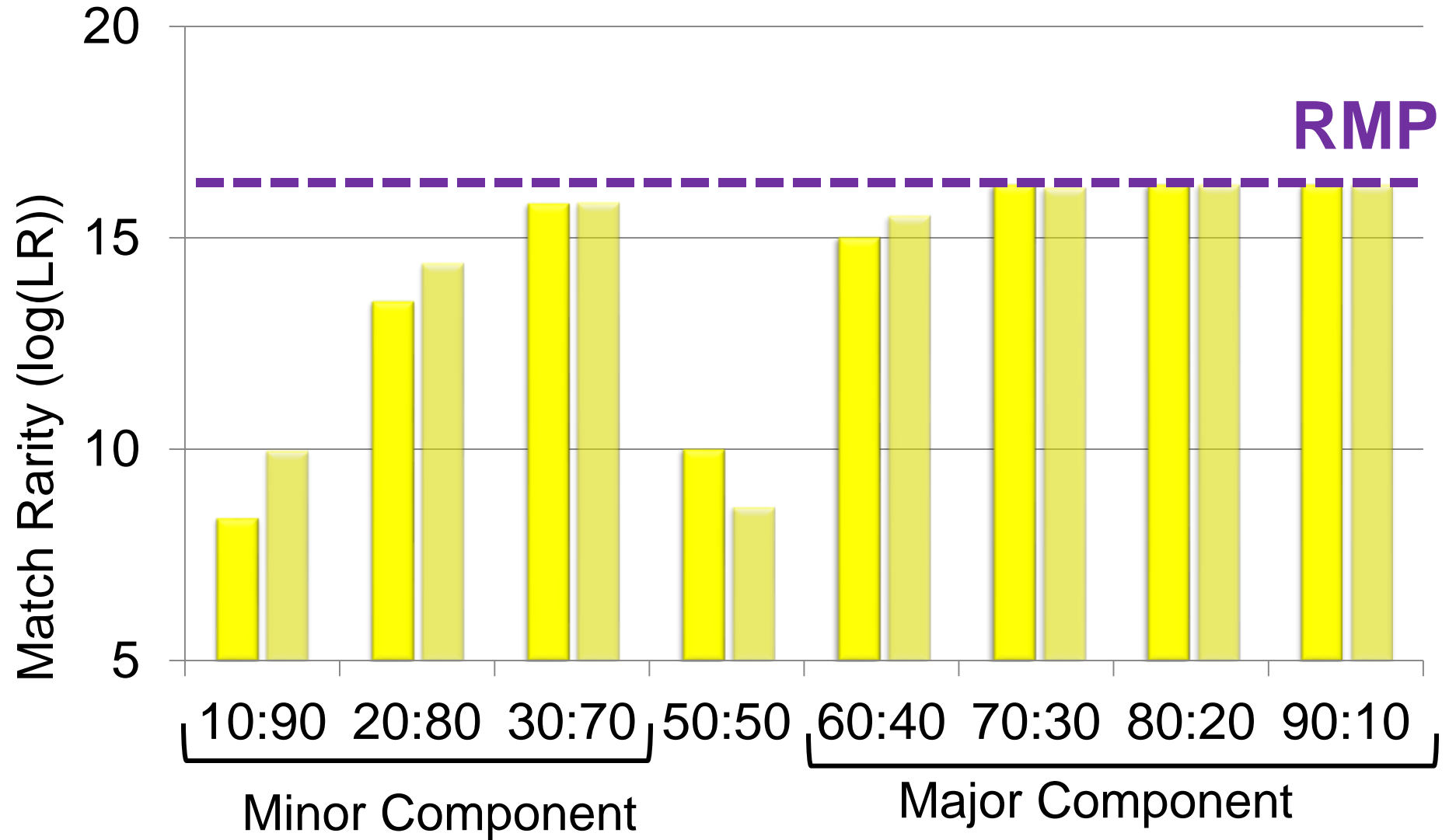
4 alleles – 0 loci  
3 alleles – 6 loci  
2 alleles – 8 loci  
1 allele – 1 loci

# Match Score in Duplicate Runs



“Easy” for  
Deconvolution

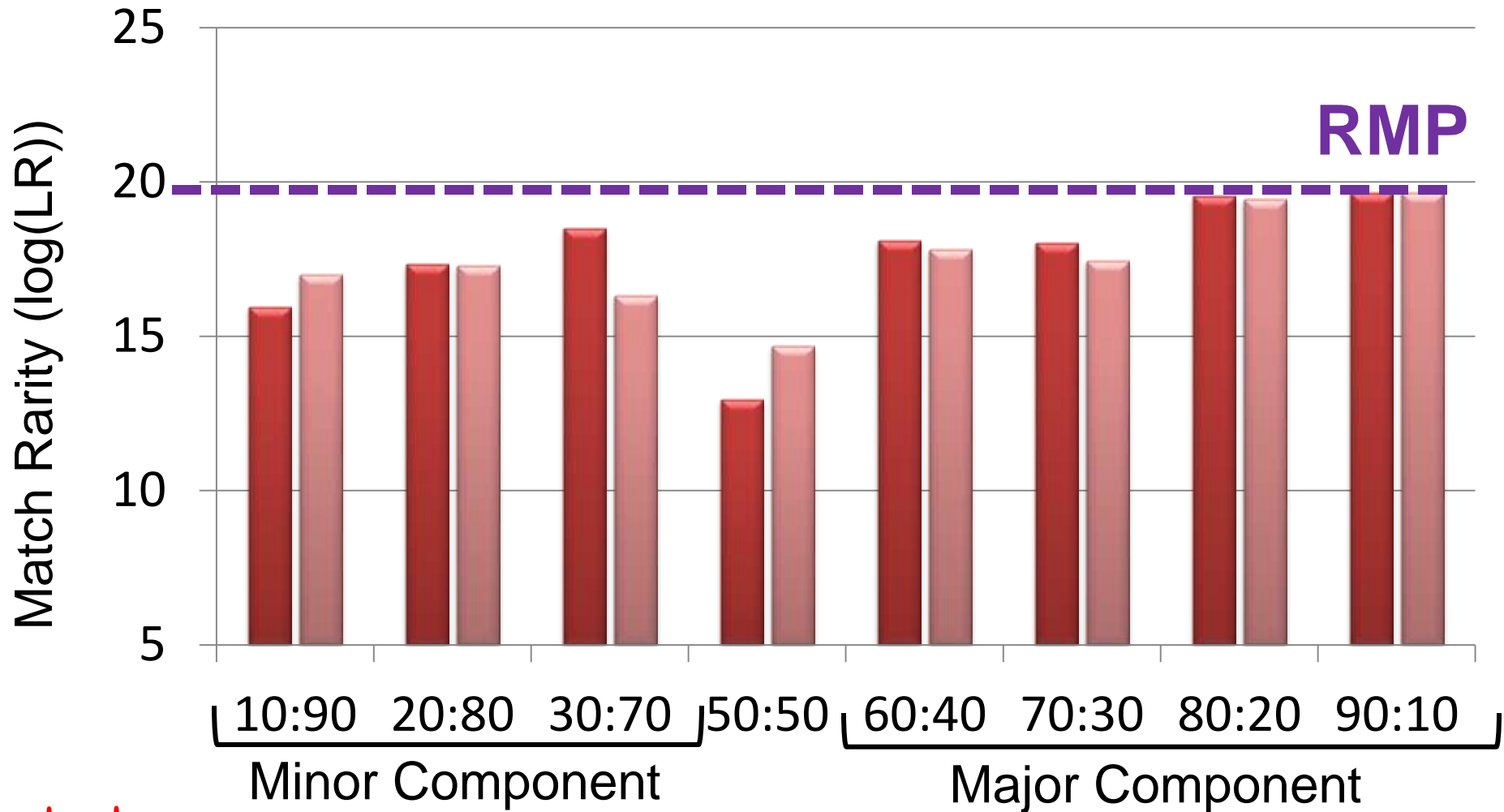
# Match Score in Duplicate Runs



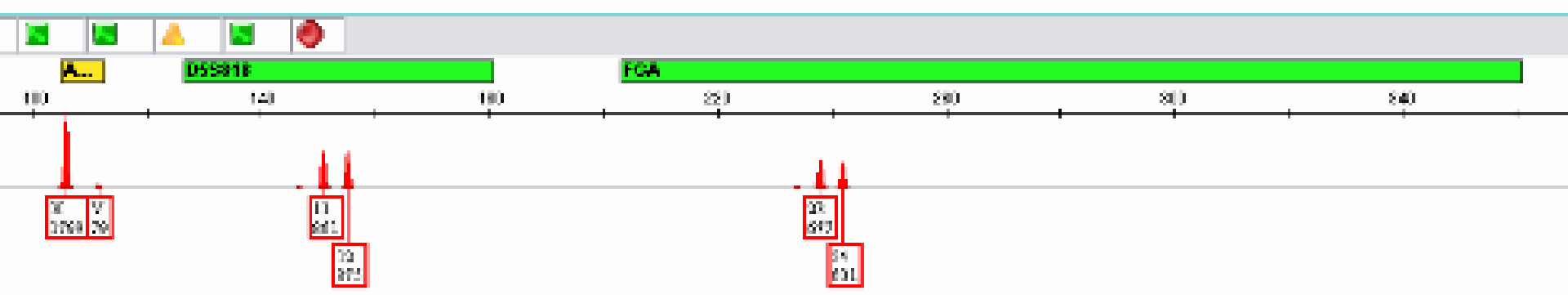
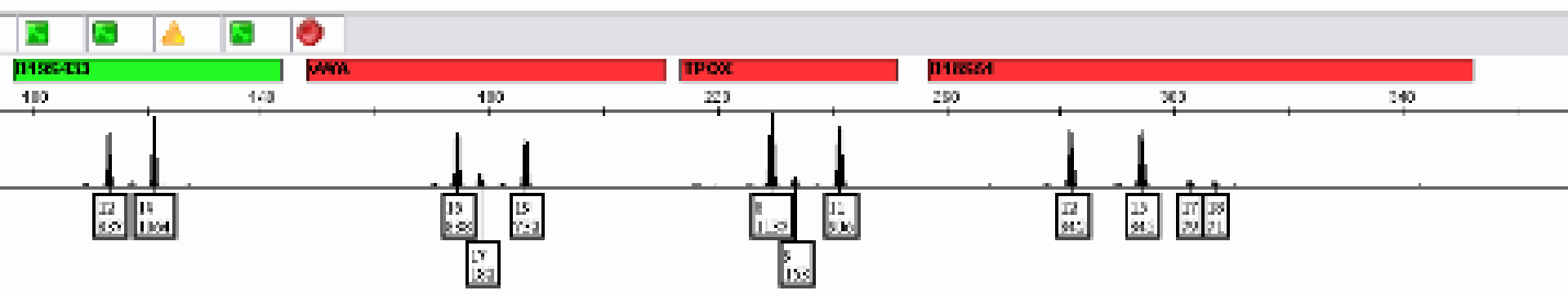
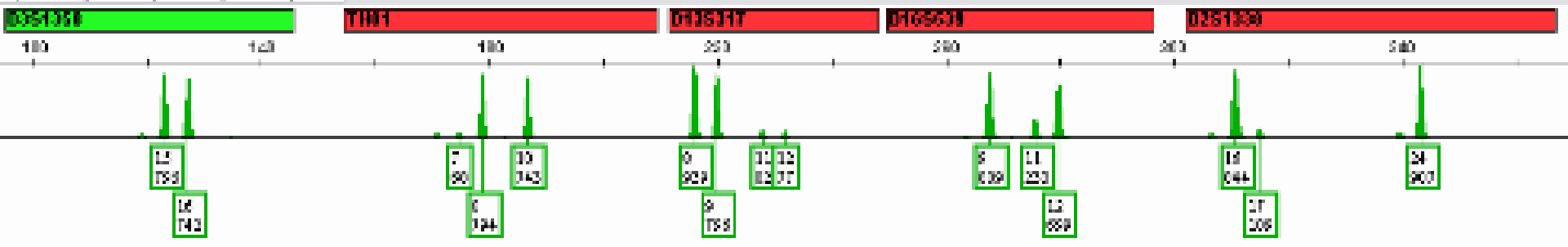
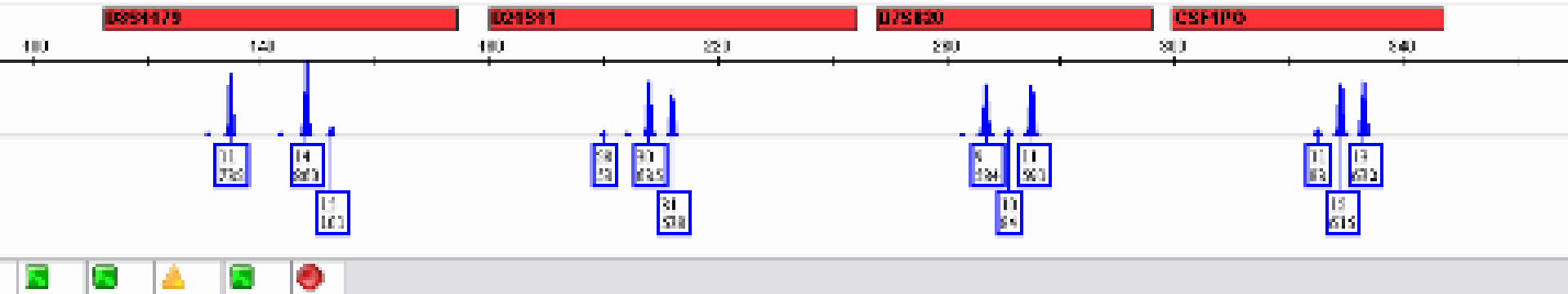
“Challenging” for  
Deconvolution



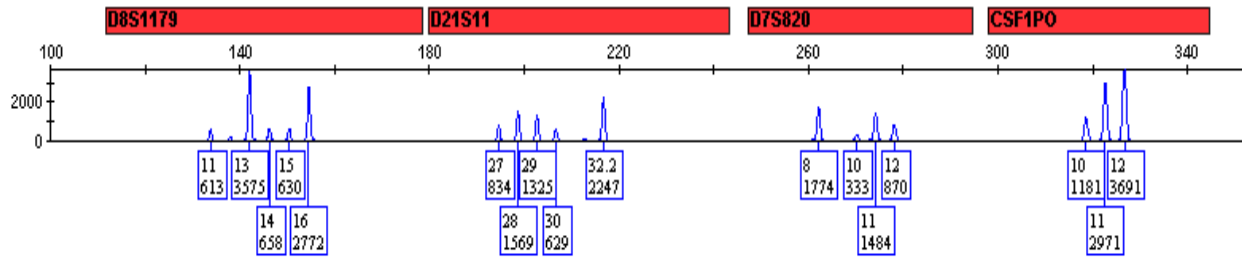
# Match Score in Duplicate Runs



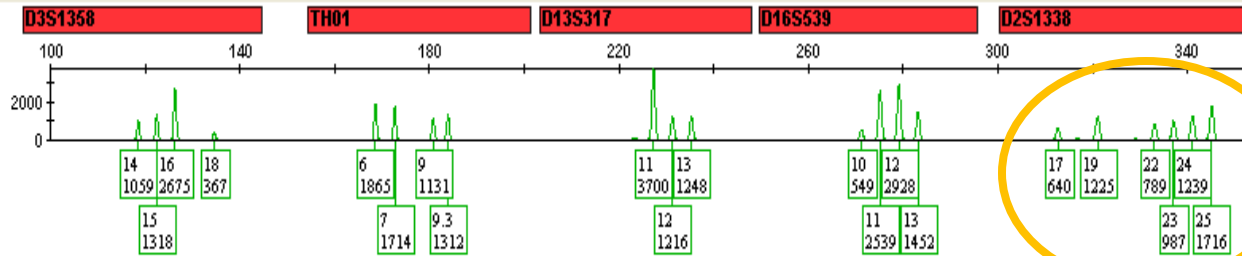
“Difficult” for  
Deconvolution



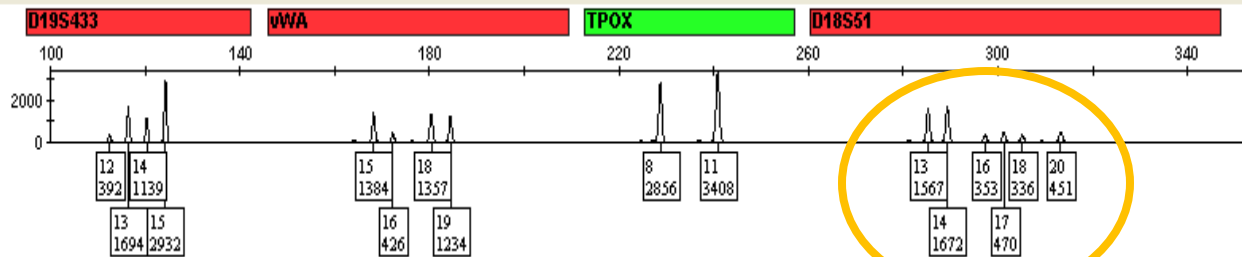
# Complex Mixture



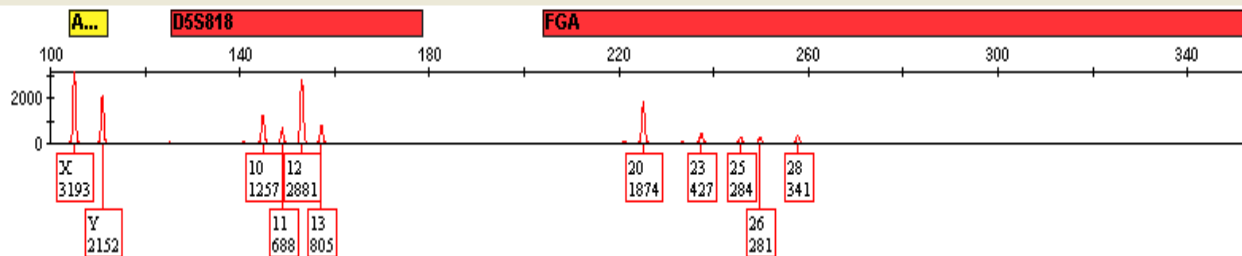
Mark Sample for Deleti



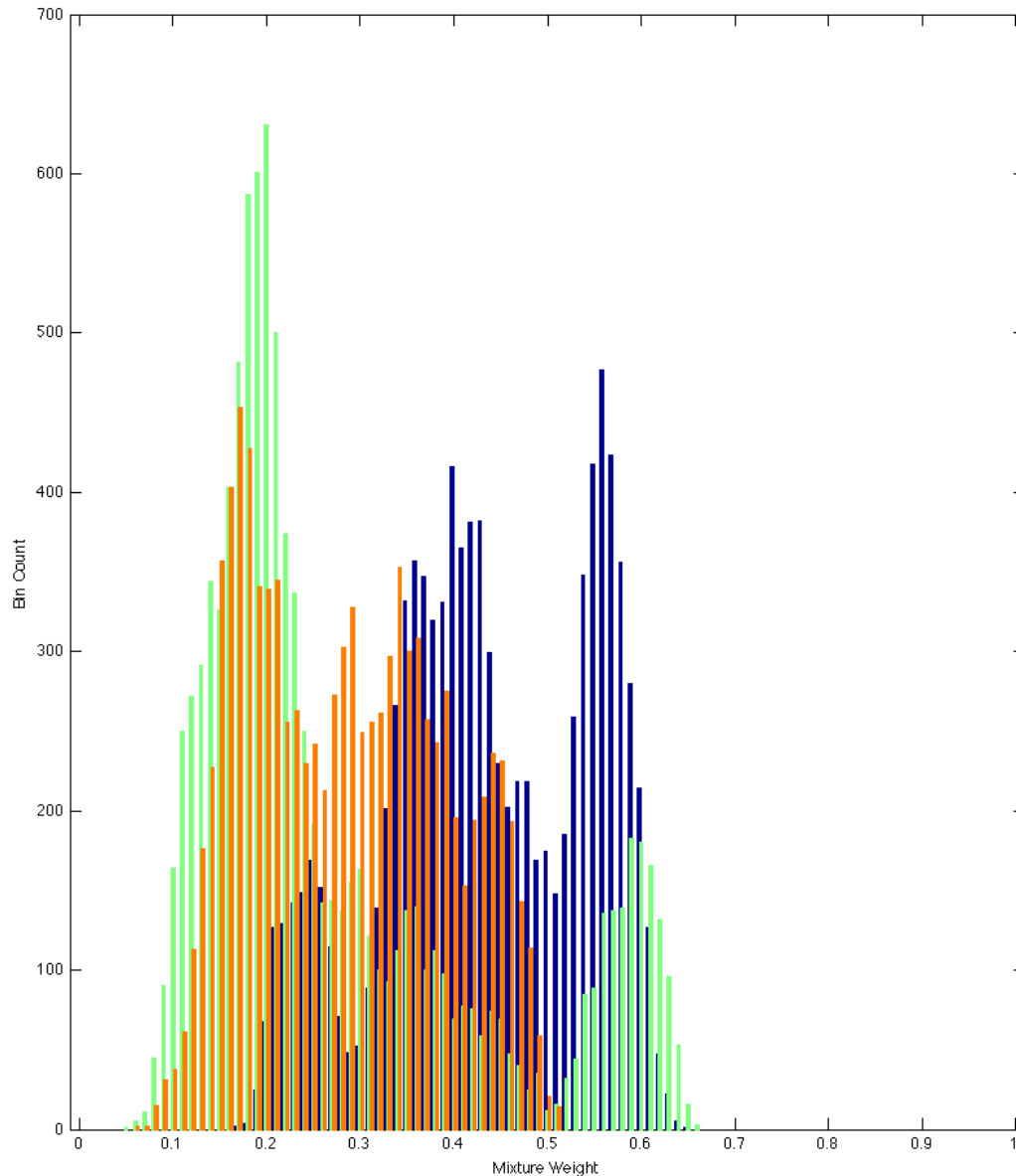
Mark Sample for Deleti



Mark Sample for Deleti



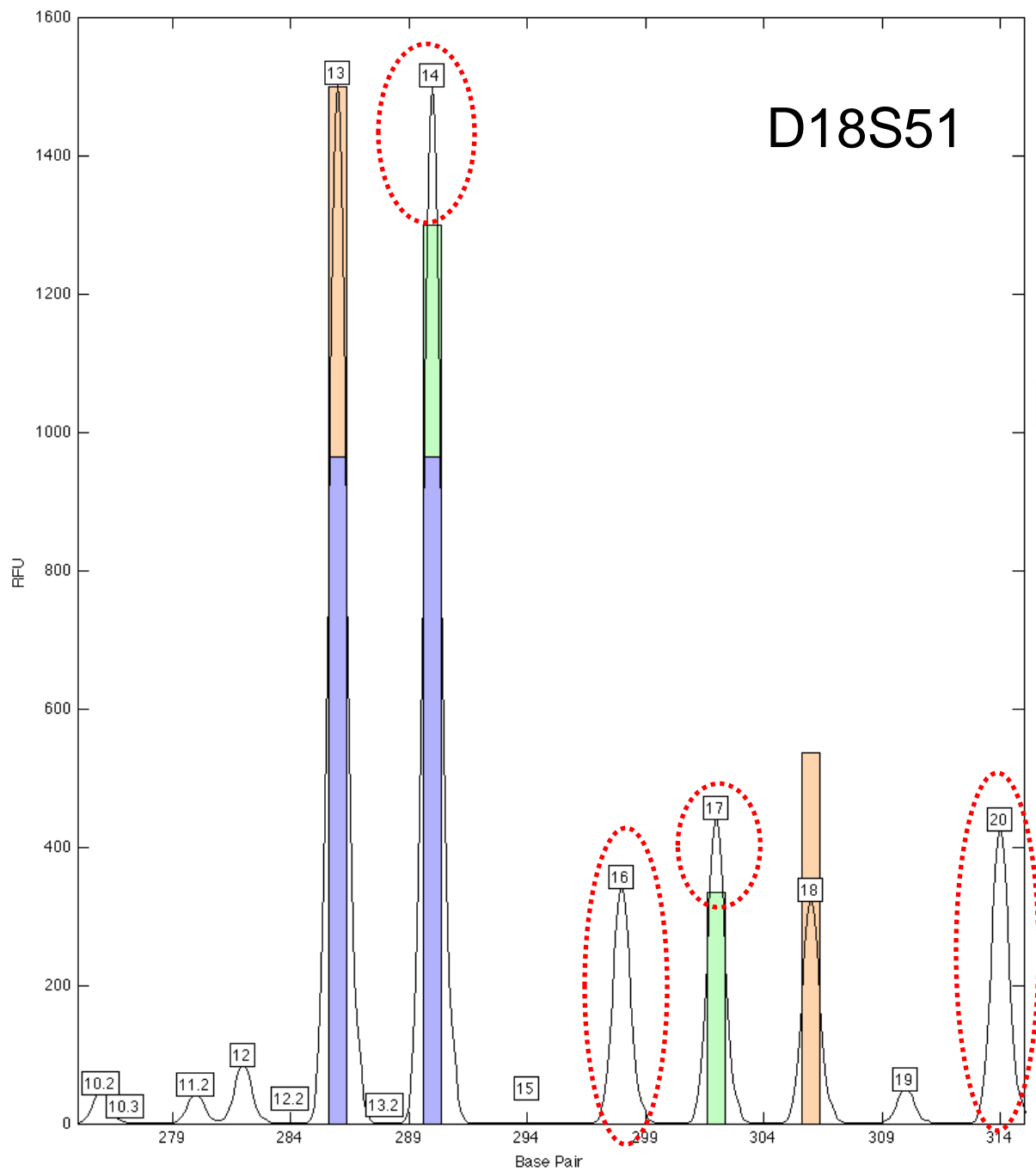
# True Allele Results – 3 person mixture



100K examinations  
3 unknowns  
(no conditioning)

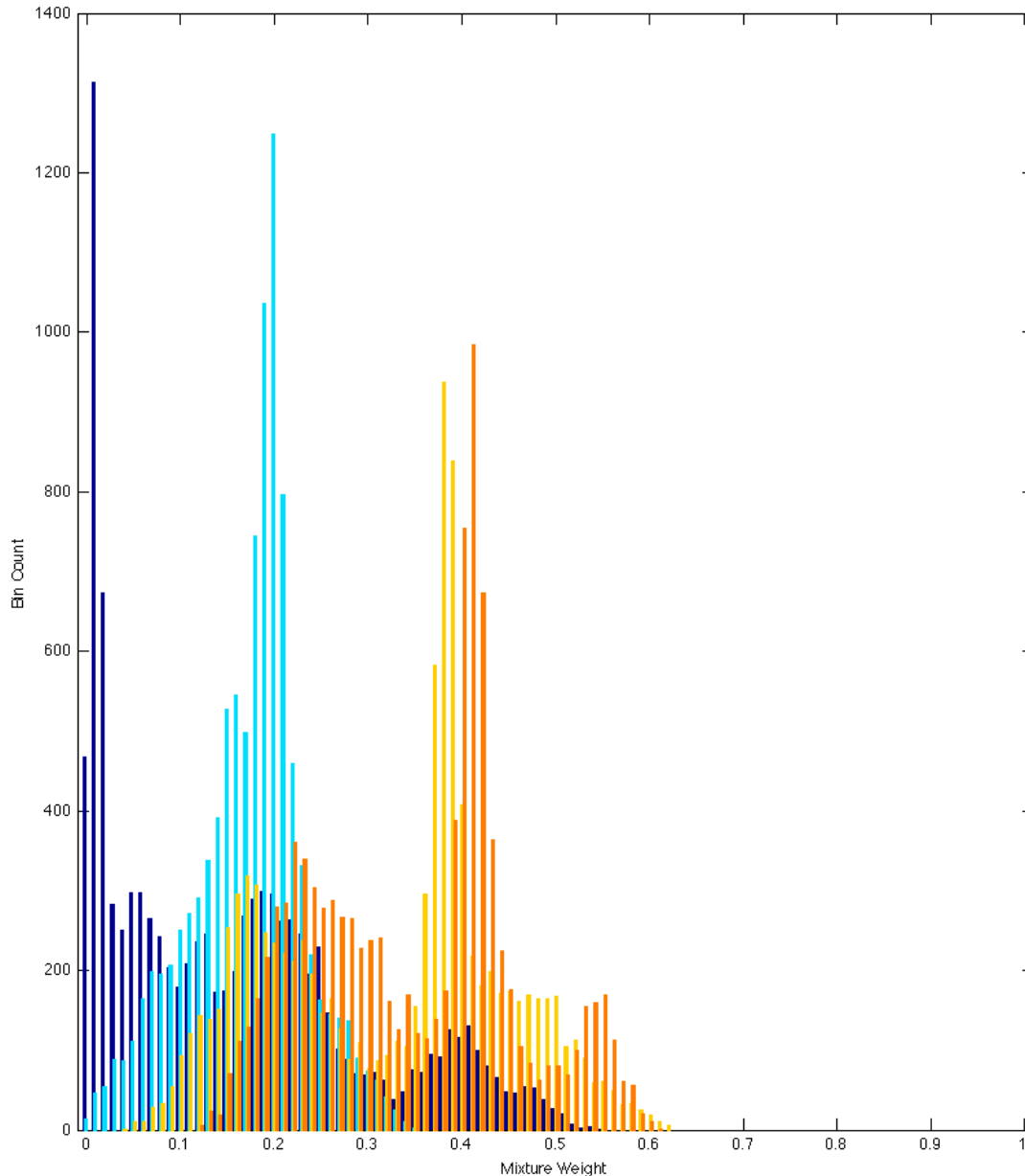
No clear separation

Mix ratio (green)  
10-60%



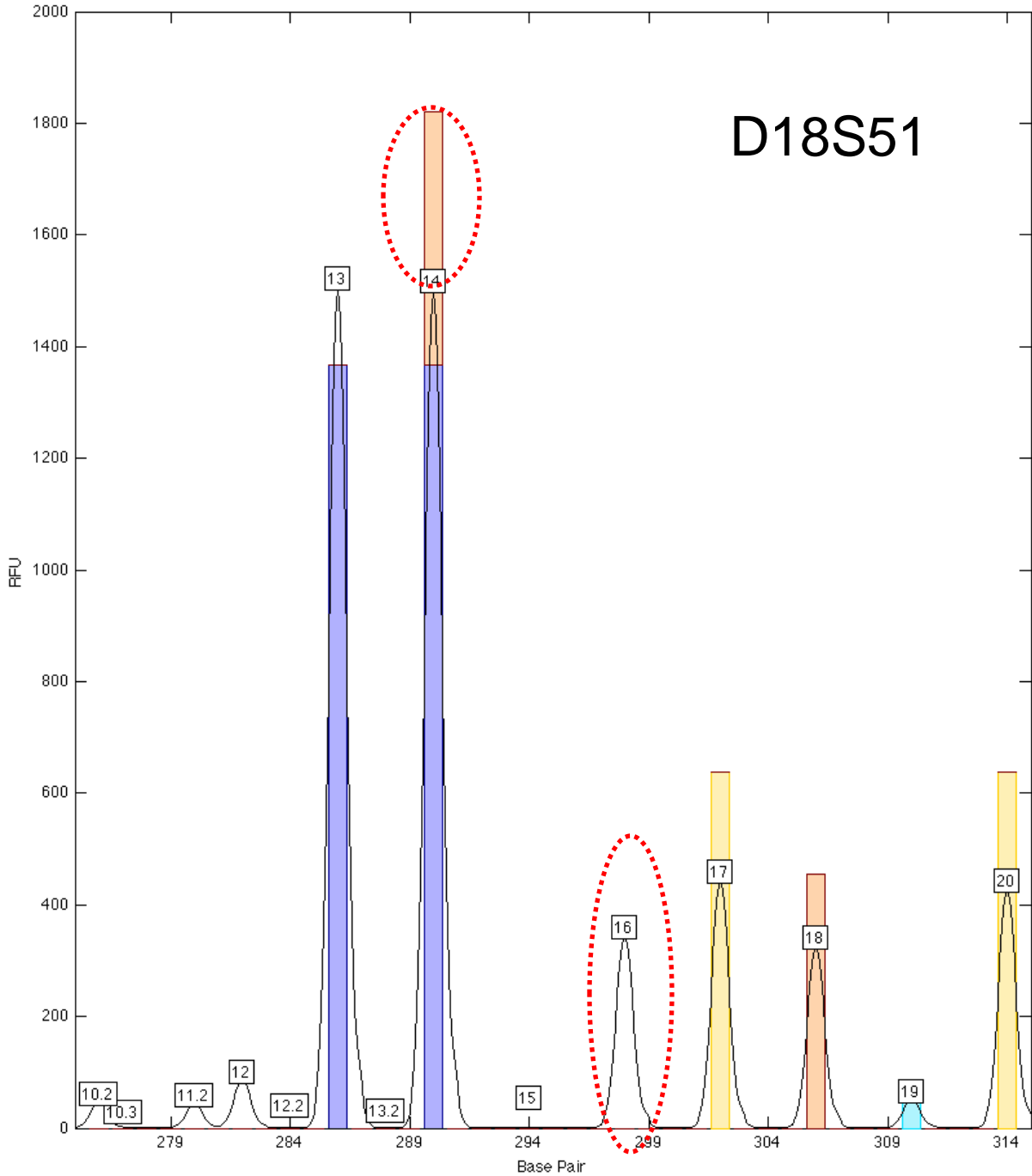
VERY  
Poor fit of the data  
to the model

# True Allele Results – 4 person mixture



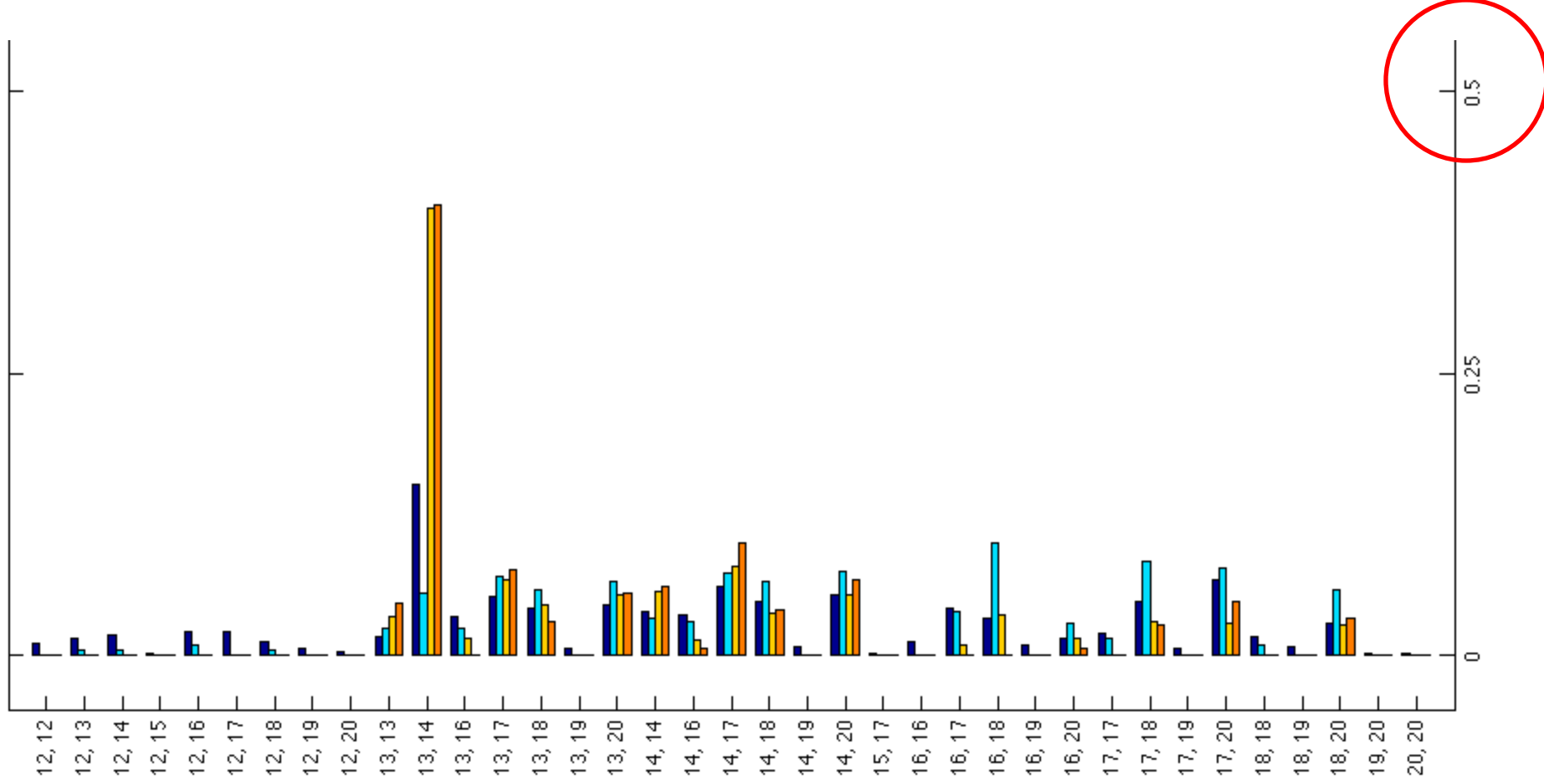
100K examinations  
4 unknowns  
(no conditioning)

Better separation,  
Still uncertainty.



D18S51

Still a poor fit of the data to the model





# Potential Suspects

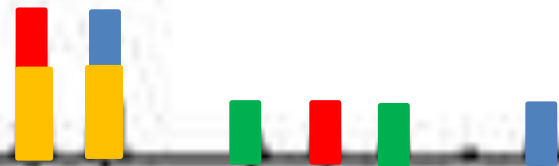
- A, B, C and D are the four individuals in the mixture.
- John Butler is also a suspect (The Butler did it).
- “Omni man” is also a possible suspect.

D18S51

60

300

340



13  
1567

16  
353

18  
336

20  
451

14  
1672

17  
470

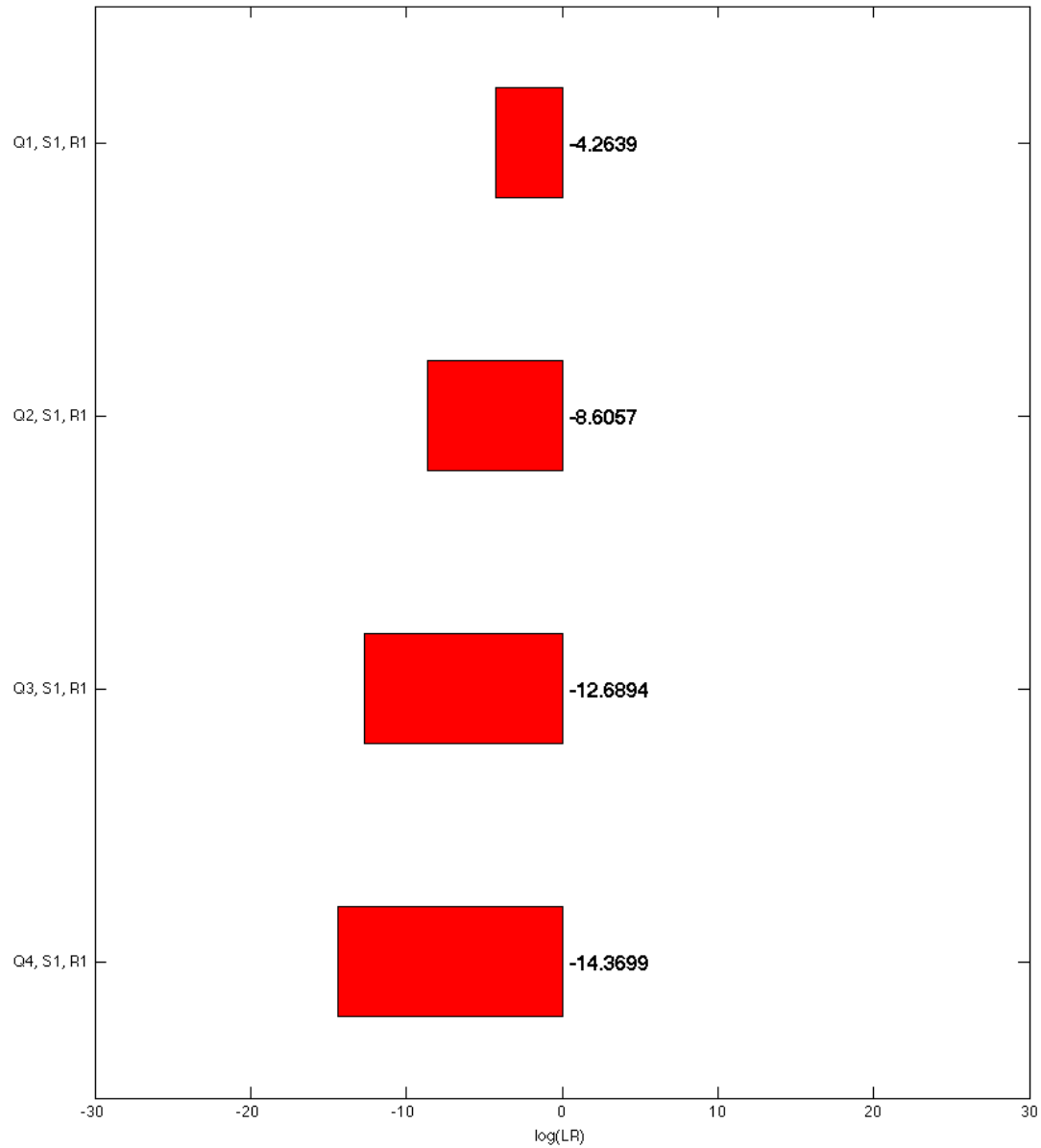
ABCD

- 14,20
- 16,18
- 13,17
- 13,14

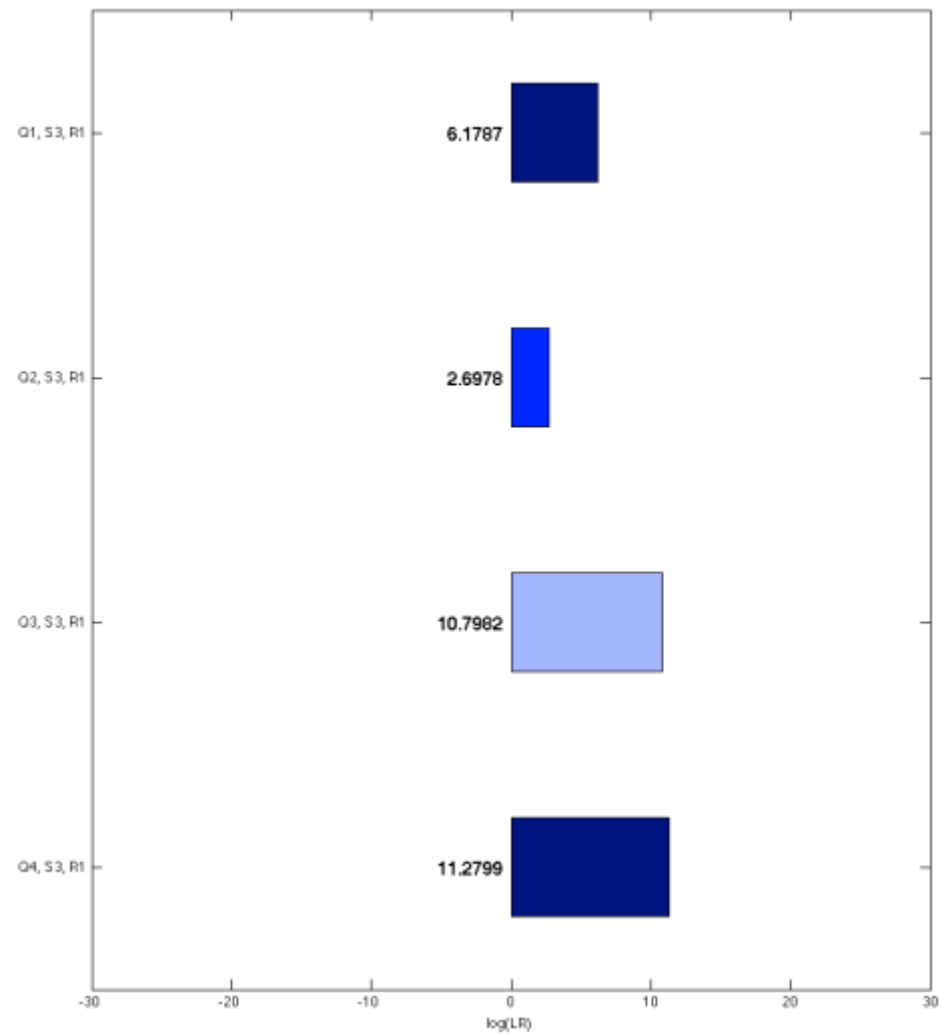
Omni

14,17

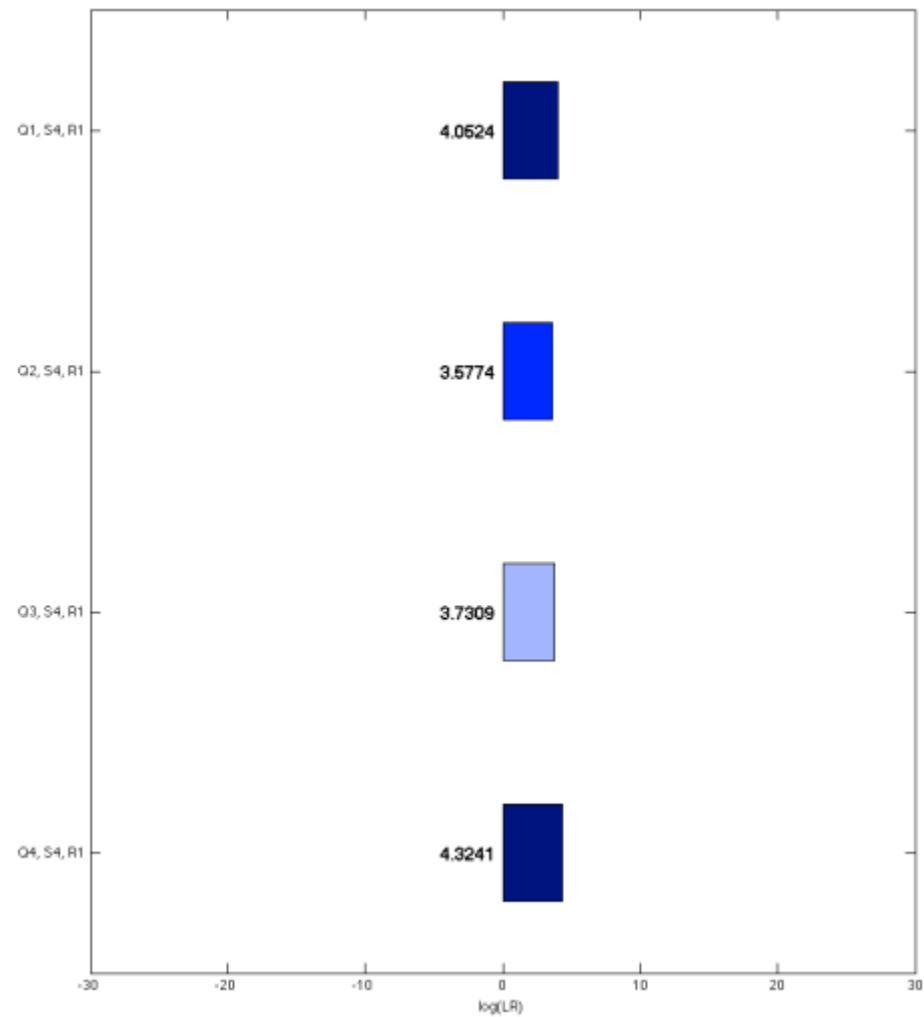
# “The Butler”



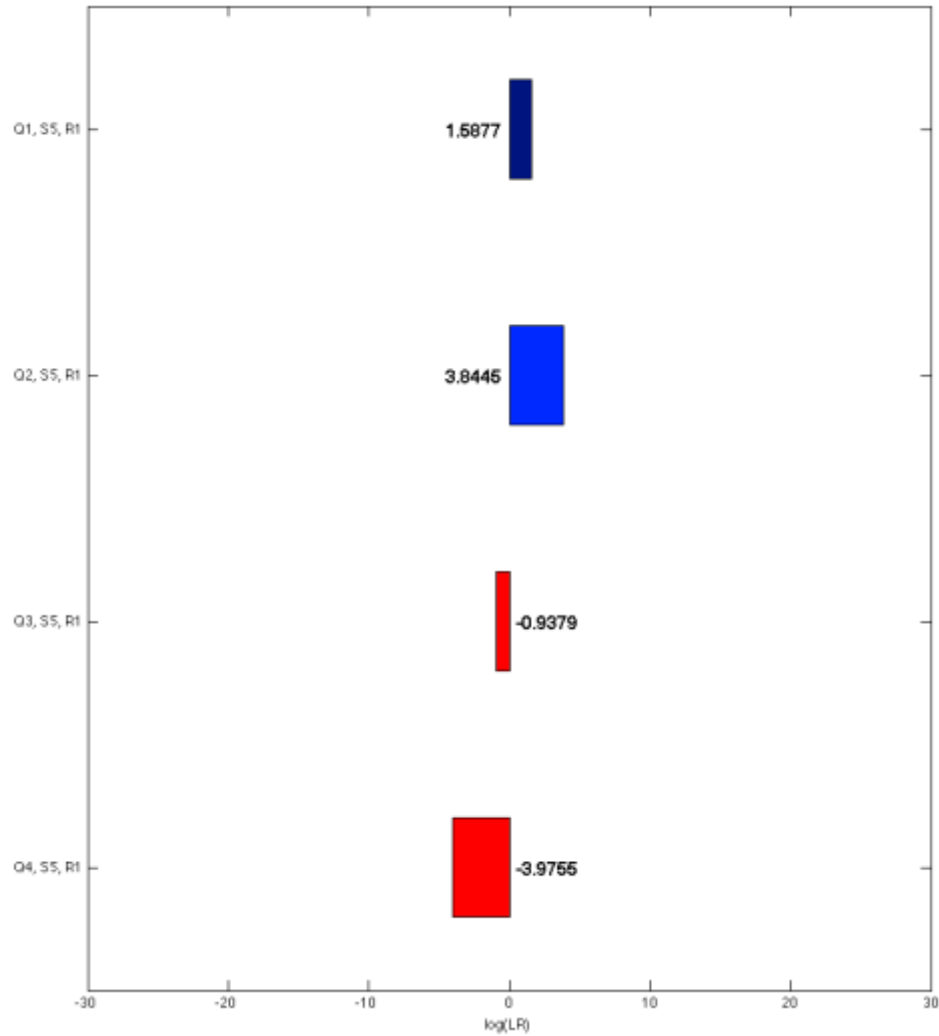
# Suspect A



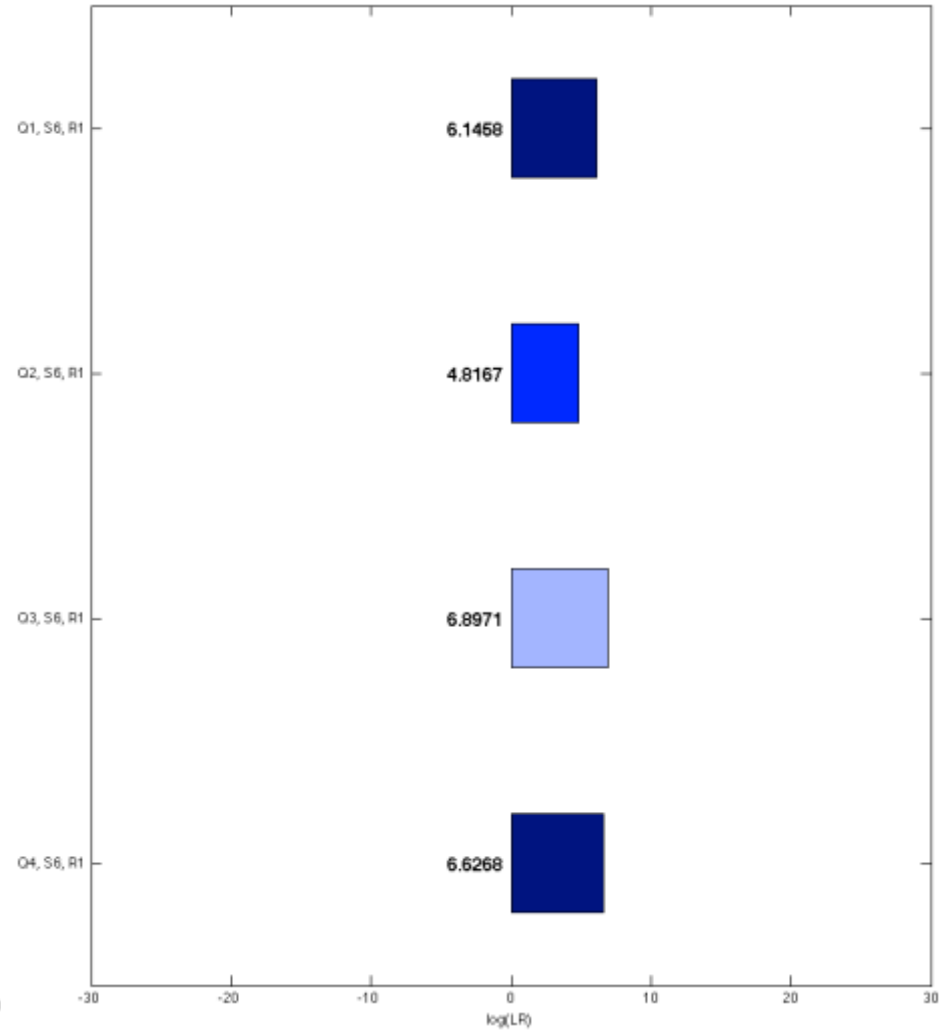
# Suspect B



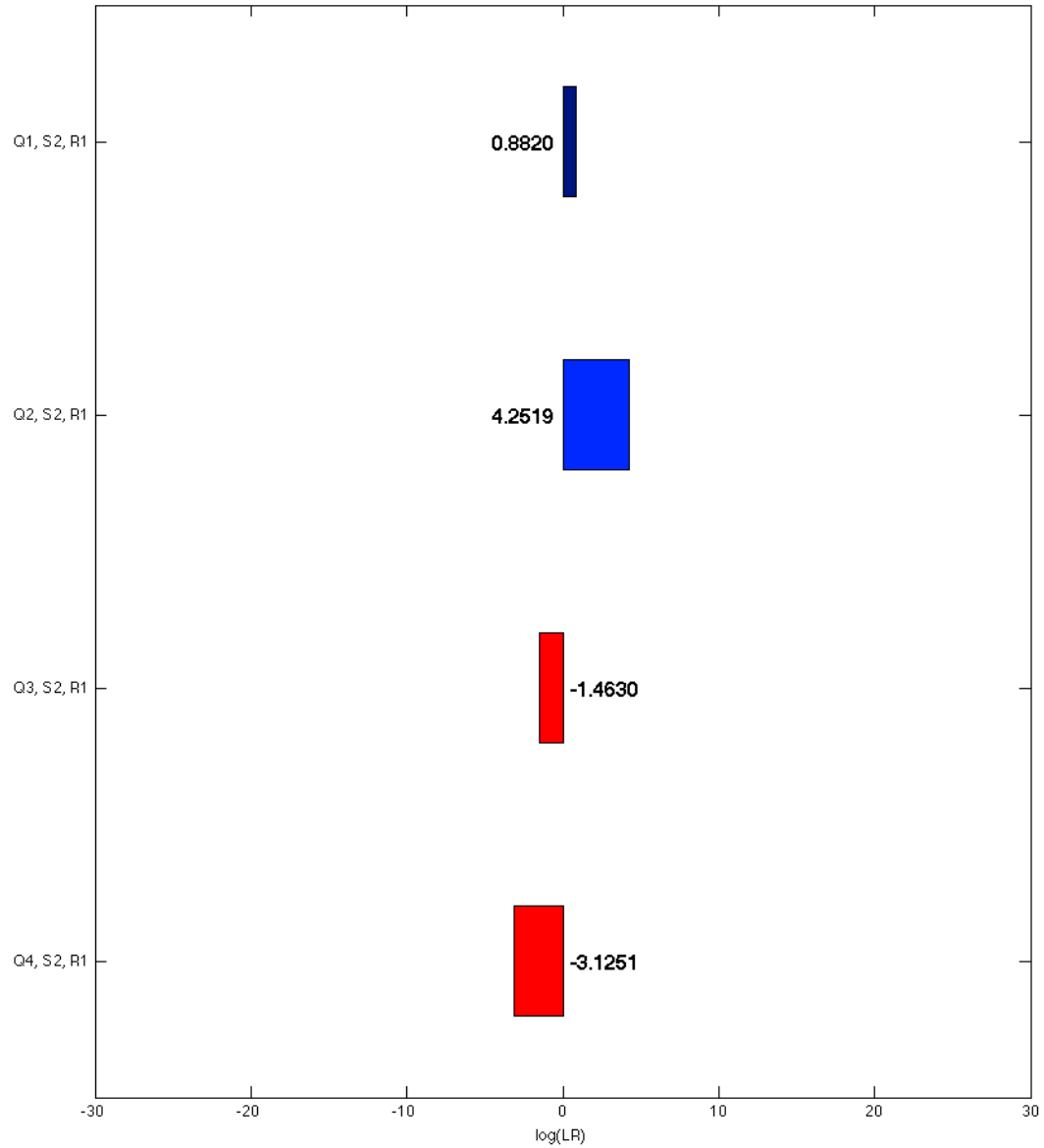
# Suspect C



# Suspect D



# Omni Man



# Strategies

- Conditioning will help...
- This may not be possible.
- Multiple replicates will be necessary.
- There is a need to determine an appropriate method for an inclusion log(LR).

# Summary of the Issues

- We need to move away from the interpretation of mixtures from an “allele-centric” point of view.
- Methods to incorporate probability will be necessary as we make this transition and confront the issues of low-level profiles with drop-out.
- “Just as logic is reasoning applied to truth and falsity, probability is reasoning with uncertainty”  
-Dennis Lindley



# Summary of the Issues

- The LR is the preferred method to evaluate low-level, complex mixture evidence with drop-out. ISFG recommendations are in press.
- This will require (obviously) software solutions... however, we need to better understand and be able to explain the statistics as a community.
- “But, for my own part, it was Greek to me”  
— William Shakespeare, *Julius Caesar*
- “We know what we are, but know not what we may be.” — William Shakespeare, *Hamlet*

# Summary of the Issues

- Extensive training will be necessary – and a single 8 hour workshop will once a year will not suffice.

- “Do, or do not. There is no try.”  
— Yoda



# Thank You

- “I can no other answer make but thanks, and thanks.” - William Shakespeare, *Twelfth Night*



[http://es.wikipedia.org/wiki/William\\_Shakespeare](http://es.wikipedia.org/wiki/William_Shakespeare)

California Association of Criminalists  
Dr. Steve Lee (SJSU) and Students!  
CyberGenetics

Dr. Robin Cotton  
Dr. Catherine Grgicak  
Dr. Charlotte Word  
Dr. John Butler

NIJ

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