

# Continuous models for mixtures

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# A comparison of statistical models for the analysis of complex forensic DNA profiles

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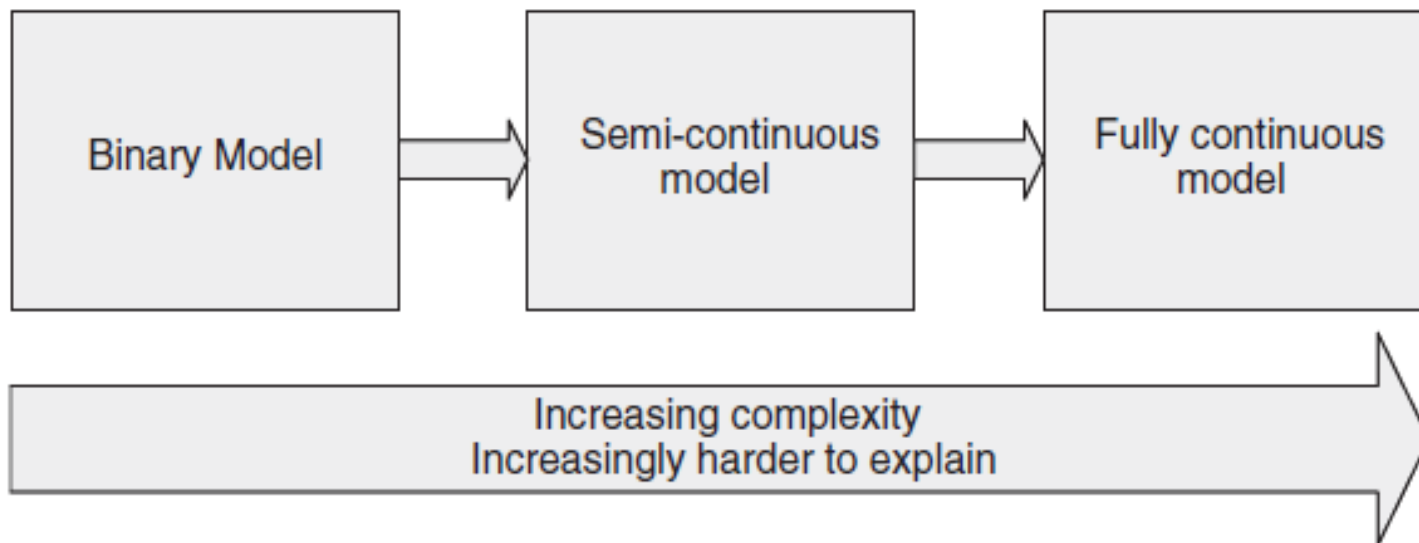
Forensic DNA interpretation

## ABSTRACT

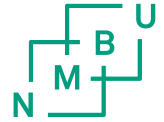
Complex mixtures and L<sub>1</sub>DNA profiles are difficult to interpret. As yet there is no consensus within the forensic biology community as to how these profiles should be interpreted. This paper is a review of some of the current interpretation models, highlighting their weaknesses and strengths. It also discusses what a forensic biologist requires in an interpretation model and if this can be realistically executed under current justice systems.

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<http://www.cstl.nist.gov/strbase/training/ISFG2013workshops.htm>



**Fig. 3.** Summary of the relationship of the different models for forensic DNA interpretation.



# Binary model

1. Qualitative binary model (aka unrestricted...)
  - Treats alleles as present or absent and does not take into account peak height information
2. Semi-quantitative binary model (aka restricted ...)
  - Declares some of the combinations as possible or impossible

# Example

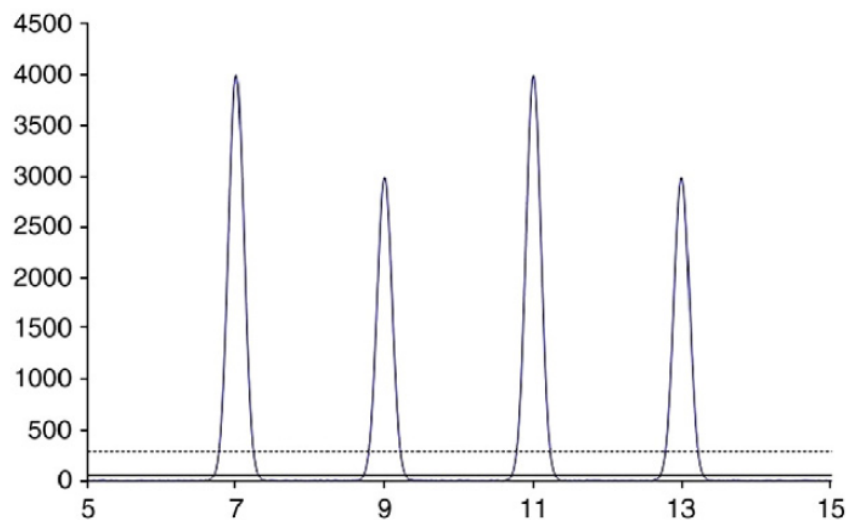


Fig. 2. Artificial epg of four-peak locus for a two-person mixture.

6 genotype combinations

$\{7/9, 11/13\}, \{7/11, 9/13\}, \{7/13, 9/11\}$

$\{11/13, 7/9\}, \{9/13, 7/11\}, \{9/11, 7/13\}$

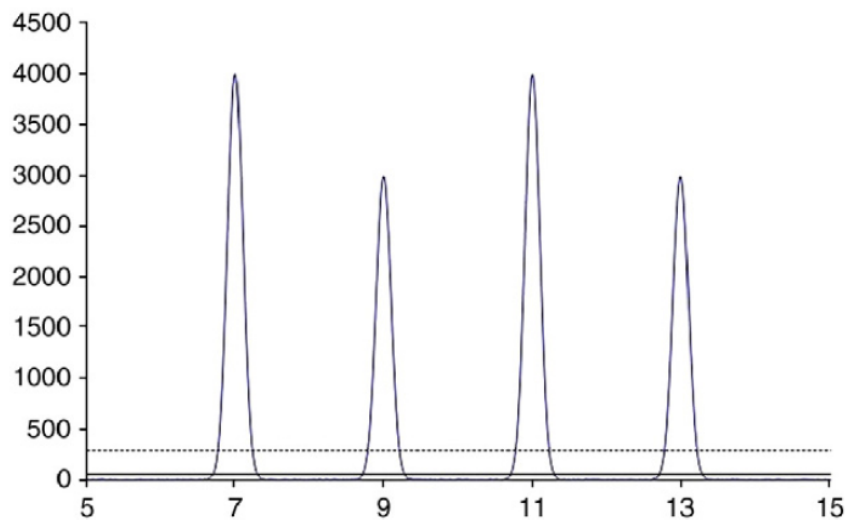


Fig. 2. Artificial epg of four-peak locus for a two-person mixture.

Suspect 7/11

$H_P : S+NN1, H_D : NN1+NN2$

Assume only combinations are 7/11, 9/13

$$LR_2 = \frac{2p_9p_{13}}{2p_7p_9 + 2p_{11}p_{13} + 2p_{11}p_{13} + 2p_7p_9} = \frac{1}{4p_7p_{11}}$$

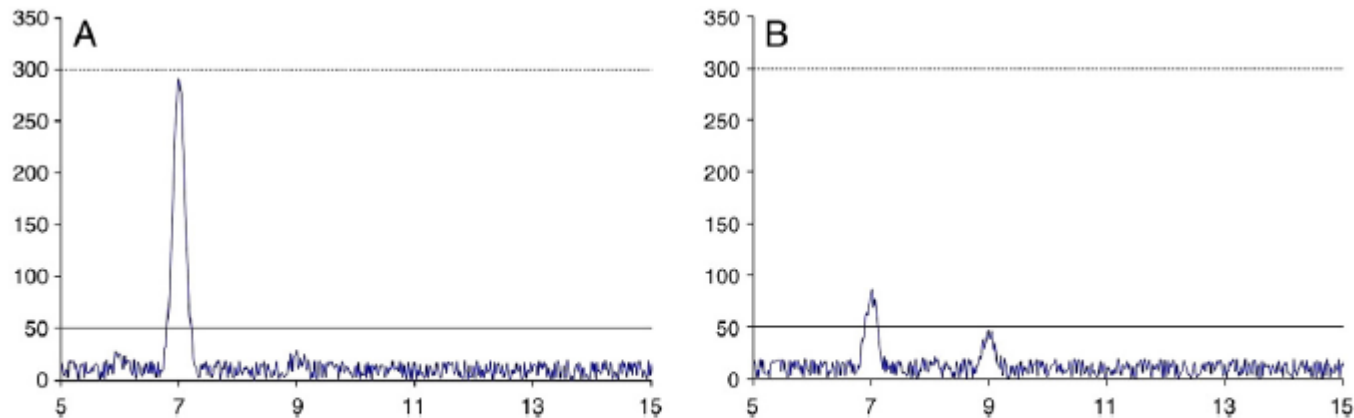
Ignoring peak information

$$LR_1 = \frac{2p_9p_{13}}{24p_7p_9p_{11}p_{13}} = \frac{1}{12p_7p_{11}}$$

$$LR_2 = 3LR_1$$

# Semi continuous models

- Artefacts:
  - Drop-out
  - Drop-in
  - Stutters



**Fig. 1.** Two examples of non-concordance where  $POI = 7,9$ . A large concordant 7 allele with no 9 peak observed (non-tolerable non-concordance) and B small concordant 7 allele with a non-concordant 9 peak visible sub-threshold (tolerable non-concordance). Stochastic threshold = 300 RFU, limit of detection 50 RFU.

Semi-continuous model includes a drop-out probability  $d$   
 Software: LRmix (Haned), ....

# Continuous model

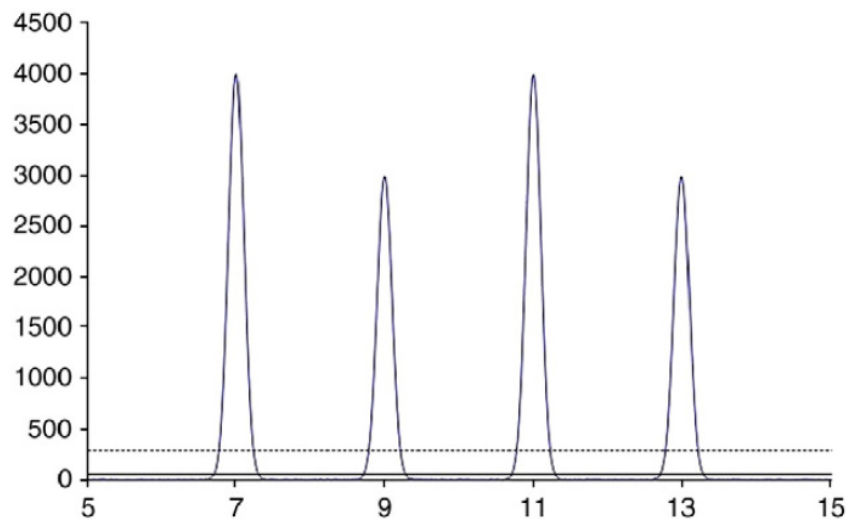


Fig. 2. Artificial epg of four-peak locus for a two-person mixture.

6 genotype combinations

$\{7/9, 11/13\}, \{7/11, 9/13\}, \{7/13, 9/11\}$

$\{11/13, 7/9\}, \{9/13, 7/11\}, \{9/11, 7/13\}$

weighed according to a  
probability distribution

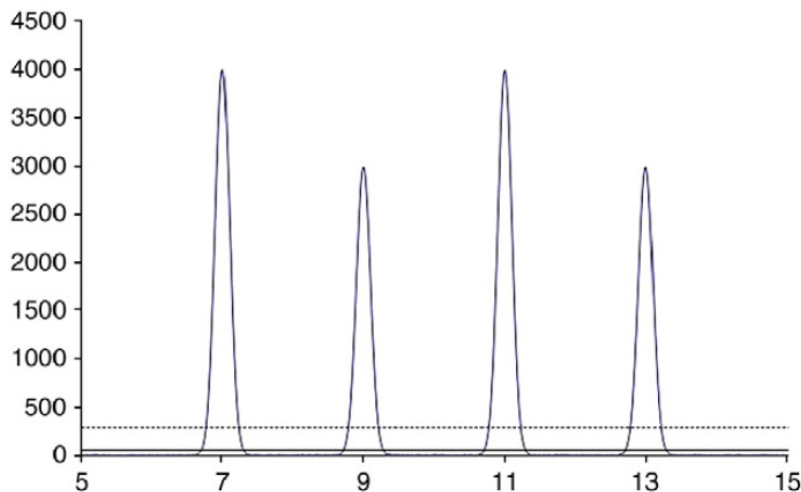


Fig. 2. Artificial epg of four-peak locus for a two-person mixture.

Suspect 7/11

$H_P : S+NN1, H_D : NN1+NN2$

$$LR = \frac{w_2 2p_9 p_{13}}{w_1 2p_7 p_9 2p_{11} p_{13} + w_2 2p_7 p_{11} 2p_9 p_{13} + \dots + w_6 2p_9 p_{11} 2p_7 p_{13}}$$

Qualitative binary model:  $w_1 = \dots = w_6 = 1$

Semi – quantitative binary model:  $w_2 = 1, w_1 = w_3 = \dots = w_6 = 0$

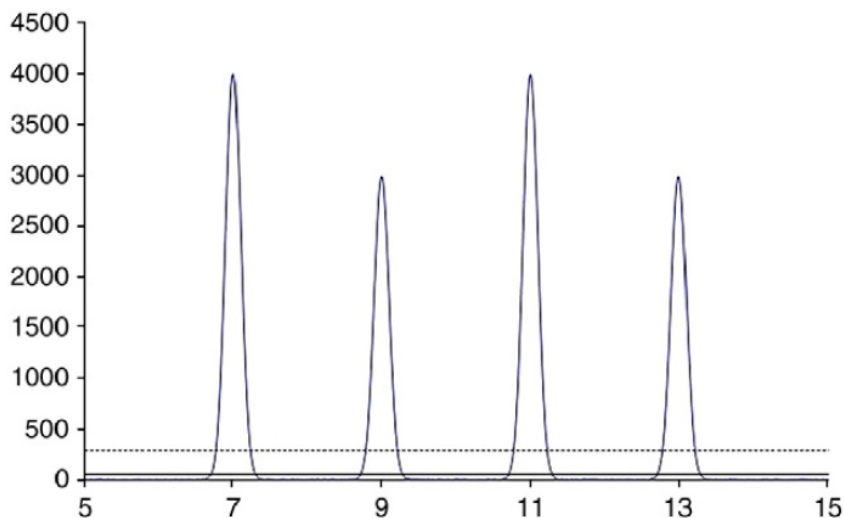


Fig. 2. Artificial epg of four-peak locus for a two-person mixture.

Suspect 7/11

$H_P : S+NN1, H_D : NN1+ NN2$

More terms with drop-out, drop-in (*drop - terms*)

$$LR = \frac{w_2 2p_9 p_{13}}{w_1 2p_7 p_9 2p_{11} p_{13} + \dots + w_6 2p_9 p_{11} 2p_7 p_{13} + \text{drop - terms}}$$

# Computation

$$LR = \frac{w_2 2 p_9 p_{13}}{w_1 2 p_7 p_9 2 p_{11} p_{13} + \dots + w_6 2 p_9 p_{11} 2 p_7 p_{13} + \text{drop - terms}}$$

- Hard part: Modelling and estimating weights
  - $w_j = \Pr(\text{peak heights} | \text{known contributors})$
  - ❖ Bayesian networks (Graversen, Cowell, Lauritzen, ...)
    - [DNAmixtures](#)
  - ❖ MCMC (Taylor, Bright,...)
    - STRmix <http://strmix.esr.cri.nz/>
    - TrueAllele <http://www.cybgen.com/>
  - ❖ Numerical integration



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Forensic population genetics – original research

### The interpretation of single source and mixed DNA profiles

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### Developing allelic and stutter peak height models for a continuous method of DNA interpretation

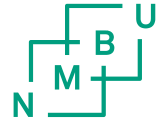
Jo-Anne Bright<sup>a, b</sup>  , Duncan Taylor<sup>c</sup>, James M. Curran<sup>b</sup>, John S. Buckleton<sup>a</sup>

<sup>a</sup> ESR Ltd, Private Bag 92021, Auckland, New Zealand

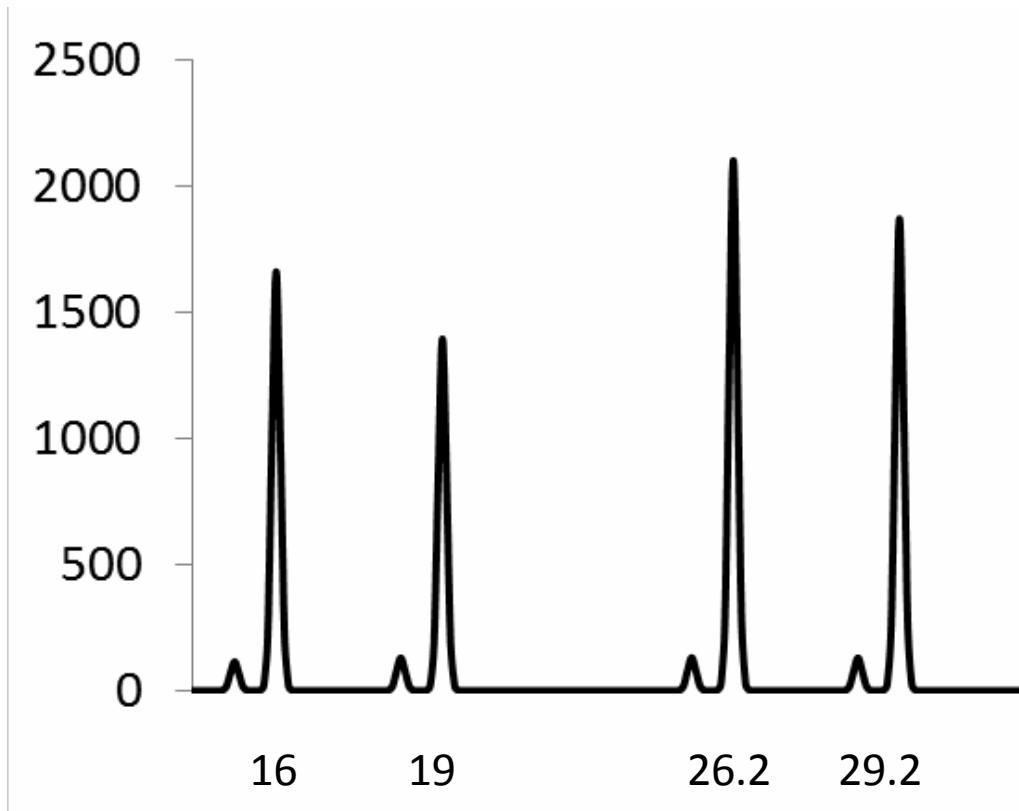
<sup>b</sup> Department of Statistics, University of Auckland, Private Bag 92019, Auckland, New Zealand

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# SE33

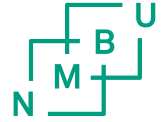


- Two person mixture



Suspect=16/19  
HP: suspect+NN  
HD: NN1+NN2

# Possible genotype combinations



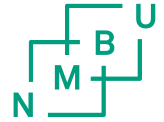
Number	Contributor 1		Contributor 2	
1	19	29.2	16	26.2
2	16	19	26.2	29.2
3	16	29.2	19	26.2
4	19	26.2	16	29.2
5	16	26.2	19	29.2
6	26.2	29.2	16	19



Number	Contributor 1		Contributor 2		Product Pr(O S <sub>j</sub> ,M)	Expected weight
1	19	29.2	16	26.2	26	0.208
2	16	19	26.2	29.2	76	0.608
3	16	29.2	19	26.2	7	0.056
4	19	26.2	16	29.2	12	0.096
5	16	26.2	19	29.2	3	0.024
6	26.2	29.2	16	19	1	0.008

$$LR_C = \frac{\sum_j w_j \Pr(S_j | H_1)}{\sum_u w_u \Pr(S_u | H_2)}$$

# Likelihood ratio



- Assuming person of interest was 16,19
- Assume that according to other (unseen) loci POI must be contributor 1

Number	Contributor 1		Contributor 2		Product $\Pr(O S_j, M)$	Expected weight
1	19	29.2	16	26.2	26	0.208
2	16	19	26.2	29.2	76	0.608
3	16	29.2	19	26.2	7	0.056
4	19	26.2	16	29.2	12	0.096
5	16	26.2	19	29.2	3	0.024
6	26.2	29.2	16	19	1	0.008

# Likelihood ratio

$$\Pr(E | H_1) = 0.608 \times 2 \times f_{26.2} \times f_{29.2}$$

$$\begin{aligned} \Pr(E | H_2) &= 0.208 \times 2 \times f_{19} \times f_{29.2} \times 2 \times f_{16} \times f_{26.2} + \\ & 0.608 \times 2 \times f_{16} \times f_{19} \times f_{26.2} \times f_{29.2} + \\ & 0.056 \times 2 \times f_{16} \times f_{29.2} \times 2 \times f_{19} \times f_{26.2} + \\ & 0.096 \times 2 \times f_{19} \times f_{26.2} \times 2 \times f_{16} \times f_{29.2} + \\ & 0.024 \times 2 \times f_{16} \times f_{26.2} \times 2 \times f_{19} \times f_{29.2} + \\ & 0.008 \times 2 \times f_{26.2} \times f_{29.2} \times 2 \times f_{16} \times f_{19} + \\ & = 4 f_{16} f_{19} f_{26.2} f_{29.2} \end{aligned}$$



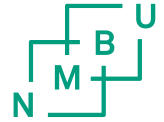
## Likelihood ratio, product rule

Allele	Frequency
16	0.0456
19	0.0659

$$\begin{aligned} LR &= \frac{0.608 \times 2 f_{26.2} f_{29.2}}{4 f_{16} f_{19} f_{26.2} f_{29.2}} \\ &= \frac{0.608}{2 f_{16} f_{19}} \\ &= 101.2 \end{aligned}$$

Several combinations are unlikely and so the binary model gives a lower value:

$$LR_{Binary} = \frac{1}{12 f_{16} f_{19}} \approx 50$$



# Exercise

- Describe the three methods presented and discuss pros and cons!