Interpreting low template DNA profiles

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ABSTRACT

We discuss the interpretation of DNA profiles obtained from low template DNA samples. The most important challenge to interpretation in this setting arises when either or both of “drop-out” and “drop-in” create discordances between the crime scene DNA profile and the DNA profile expected under the prosecution allegation. Stutter and unbalanced peak heights are also problematic, in addition to the effects of masking from the profile of a known contributor. We outline a framework for assessing such evidence, based on likelihood ratios that involve drop-out and drop-in probabilities, and apply it to two casework examples. Our framework extends previous work, including new approaches to modelling homozygote drop-out and uncertainty in allele calls for stutter, masking and near-threshold peaks. We show that some current approaches to interpretation, such as ignoring a discrepant locus or reporting a “Random Man Not Excluded” (RMNE) probability, can be systematically unfair to defendants, sometimes extremely so. We also show that the LR can depend strongly on the assumed value for the drop-out probability, and there is typically no approximation that is useful for all values. We illustrate that ignoring the possibility of drop-in is usually unfair to defendants, and argue that under circumstances in which the prosecution relies on drop-out, it may be unsatisfactory to ignore any possibility of drop-in.

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1. Introduction

In 2003 James Garside and Richard Bates were convicted in London of the murder of Marilyn Garside, the estranged wife of James Garside. It was the prosecution’s case that Garside had hired Bates to murder his wife. The initial conviction was successfully appealed in 2004, leading to a retrial in 2005 which was followed by an unsuccessful appeal in 2006 [1].

Mixed DNA profiles from the crime scene formed a key part of the evidence. The major component of these crime scene profiles (CSP) corresponded to the victim and was assumed to be from her (her full profile was available). In addition, up to eight minor-component alleles were identified. Bates was subsequently arrested and profiled, revealing that the 17 distinct alleles in his 10-locus profile (Table 1) included all 8 CSP minor-component alleles, as well as a further six that were masked by the victim. However three of Bates’ alleles, one at locus D2 and two at D18, were not reported in any of the large number of electropherograms (epg) generated in the investigation, from different crime scene samples and different profiling runs under varying laboratory conditions.

The prosecution allegation that Bates was the origin of the CSP minor component implies that allelic drop-out must have occurred at D2 and D18. Drop-out arises when an allele that is carried by an individual contributing DNA to a sample is not reported in a DNA profile obtained from that sample. A related phenomenon, drop-in, occurs when trace amounts of DNA, for example from a crime scene environment or laboratory plasticware, generate one or more spurious alleles in the profile. It is rare for drop-out or drop-in to occur with good-quality samples not subject to degradation or inhibition, but they become more likely as the amount of DNA template is reduced or environmental exposure increases. In Bates, alleles not attributable to the defendant were evident in some crime scene profiles, but not in the most incriminating CSP profile obtained from that sample. A related phenomenon, drop-in, occurs when trace amounts of DNA, for example from a crime scene environment or laboratory plasticware, generate one or more spurious alleles in the profile. It is rare for drop-out or drop-in to occur with good-quality samples not subject to degradation or inhibition, but they become more likely as the amount of DNA template is reduced or environmental exposure increases. In Bates.

At D2, one allele corresponding to Bates was reported, but not his other allele. The RMNE calculation uses the “2p rule”, under

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which the allele frequency estimate is doubled for a single reported allele when drop-out is possible. This rule has previously been shown to be non-conservative [3] particularly when the drop-out probability is low. Here we show that the problem is more severe in the presence of masking, and that it can extend to the high-drop-out setting.

The three alleles of Bates that were discordant with the prosecution hypothesis were referred to in the judgment as “voids”. The judge in the second trial described the choice facing the court in the following terms:

“What are the consequences of the impossibility of assigning a statistical weight to the voids? The alternatives are to exclude the evidence entirely or to admit it subject to an appropriate warning to the jury of the limitations of the evidence, and particularly highlighting the fact that although what was found was consistent with Bates’ DNA profile, the voids at D2 and D18 in particular may have contained an allele or alleles, the presence of which would have been wholly explicable. In arriving at the correct conclusion it is important to remember that scientific evidence frequently only provides a partial answer to a case, or to an issue in a case. However, the test of admissibility is not whether the answer is complete, but whether science can properly and fairly contribute to the matter in question...”.

At the final appeal the court ruled:

“We can see no reason why partial profile DNA evidence should not be admissible provided that the jury are made aware of its inherent limitations and are given a sufficient explanation to enable them to evaluate it.”

It is welcome to hear from the judiciary that evidence interpretation does not have to be perfect to be admissible. However, the quoted statements are worrying in several respects. Talk of “voids” that may have contained an allele suggests an assumption that drop-out has occurred, whereas there may be no “missing” allele at these loci: the alleles of the true culprit could have been masked by the victim’s alleles. In the presence of masking we cannot know if any alleles have dropped out, and it is not clear to which one(s). It is therefore dangerous to speak of “voids” without some form of qualification, such as “according to the prosecution case”. Since such qualification is awkward to use frequently, we propose the term “discordant alleles” to indicate alleles that are not consistent with the prosecution case.

The Garside and Bates case is far from unique in grappling with difficulties associated with interpreting low template DNA (LTDNA) profiles with discordant alleles. The profiles shown in Table 3 arose in an ongoing US-based case (hereafter US case). The observation of more than two alleles at two loci in the CSP indicates the presence of DNA from at least two individuals, and the absence of reportable alleles at five loci indicates that substantial drop-out affects the profiles of all contributors. A known individual v is expected to be a contributor of DNA to the mixture and the issue at stake is whether or not s is another contributor. Although we cannot be sure that there are only two contributors, the low number of observed alleles suggests that this is the case [4] and we proceed under this assumption.

The prosecution’s RMNE calculation ignored the eight loci at which no CSP alleles could be attributed to s, while the remaining seven loci were given equal weight, even though at four of these loci the alleles attributable to s could also have come from v. Neither of the two homozygote alleles in the profile of s appeared in the CSP, yet the two homozygous alleles of v did appear. We argue that failure to observe the homozygotes tends to exclude s as a potential contributor, whereas the RMNE calculation ignored both loci. Indeed, in an RMNE calculation evidence can never favour a defendant. We present below alternative calculations that lead to very different conclusions from those made by the prosecution in this case.

We agree with [5] that there does not yet exist a fully satisfactory analysis of DNA profiles obtained from LTDNA samples, and so approximations and simplifying assumptions are required. However we are concerned that some approximations and assumptions being used in courts have not been adequately justified. Relevant theory for the interpretation of LTDNA profiles based on the likelihood ratio (LR) has been published [5–7]. Software for computing LRs in complex scenarios has been described [8], but is not widely available for courtroom use. We believe that the LR forms the basis of the most
rational approach to evidence evaluation [9,10], and in particular the LR approach has very significant advantages for complex evidence interpretation problems, such as those involving discordant alleles [11–16]. In our experience (in several English-speaking countries), appropriate LR-based methods do not always underpin courtroom practice in such complex cases, when the approach is at its most advantageous.

In this paper we seek to synthesize previous work and advance the LR-based methodology for LTDNA profiles, to highlight its advantages and to illustrate its impact for the two cases outlined above. We hope thereby to encourage the adoption in courts of an LR-based approach, but we do not seek here to prescribe to forensic practitioners exactly how they should present evidence, nor which approximations should be used in court. The important issue of a null minor component in the presence of masking has not been explicitly addressed in the literature, even though methods for dealing with it are straightforward extensions of existing methods. Other novel aspects developed here involve drop-out at homozygote alleles, and uncertainty about whether a peak is allelic or stutter or masking. We also provide freely-available software that underpins the analyses of this paper and allows researchers to implement and to extend or alter our formulae.

2. Background to LTDNA interpretation

DNA profiling from small amounts of DNA sometimes uses additional PCR cycles to increase amplification and hence compensate for the low starting template, and this approach is sometimes labelled low copy number (LCN) profiling. Because of ambiguity as to whether or not LCN refers to a specific profiling technique, the broader term LTDNA is used when there is limited DNA template, whether or not additional PCR cycles or other enhancement techniques are employed. In addition to drop-out and drop-in, LTDNA profiles are sometimes characterised by low epg peak heights, exaggerated peak imbalance and large stutter artefacts (Fig. 1). Typically labs employ a “reporting threshold”, often 25–50 RFU, below which any apparent peaks are not reported [12] and an “LTDNA threshold”, often 150–300 RFU, above which drop-out is excluded, so that a single allele peak is assumed to correspond to a homozygote [5,16]. In what follows, we assume that all single peaks from an unknown source lie between these two thresholds. If replication of an LTDNA profiling run is possible, it provides an opportunity to reduce uncertainty arising from stochastic effects. We will return to this issue below, but for now we will take the viewpoint of a court that is presented with a single or consensus profile, and is not given details of individual replicate runs if they exist.

Computing an LR requires a probability model to describe the real-world processes underlying an LTDNA profile. However, LTDNA profiling is complex, and there are many plausible probability models, and hence many different LRs can be proposed, each with some justification. Inevitably, fully realistic models of all sources of variation are unachievable, and approximations must be made and assessed [15,16].

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Table 3

<table>
<thead>
<tr>
<th>Locus</th>
<th>Known contributor v</th>
<th>Suspected contributor s</th>
<th>CSP</th>
<th>LR: $D = 0.5$</th>
<th>Multi-source RMNE</th>
<th>Single-source RMNE</th>
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<tbody>
<tr>
<td>D8</td>
<td>13,16</td>
<td>12,13</td>
<td>12,13,16</td>
<td>4.5</td>
<td>4.0</td>
<td>3.4</td>
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<td>29,29</td>
<td>28</td>
<td>0.38</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>D7</td>
<td>8,10</td>
<td>9,10</td>
<td>–</td>
<td>1.1</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>CSF</td>
<td>8,10</td>
<td>10,11</td>
<td>–</td>
<td>1.1</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
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<td>14,16</td>
<td>16,17</td>
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<td>1.4</td>
<td>4.3</td>
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<td>1</td>
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<tr>
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<td>4.0</td>
<td>1</td>
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<td>12,13</td>
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<td>18,20</td>
<td>19,20</td>
<td>18,20</td>
<td>1.4</td>
<td>18</td>
<td>1</td>
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<td>TPO</td>
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<td>11,12</td>
<td>9.11</td>
<td>1.3</td>
<td>7.0</td>
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<td>–</td>
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<td>1</td>
<td>1</td>
</tr>
<tr>
<td>D5</td>
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<td></td>
<td></td>
<td>2.0</td>
<td>96K</td>
<td>8.3</td>
</tr>
</tbody>
</table>

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Fig. 1. An electropherogram (epg) showing several peaks close to the reporting threshold of 50 RFU (lower grey line). The highest peak, labelled 13, has no partner but since it is above the LTDNA threshold of 300 RFU (upper grey line) it is assumed to represent a 13,13 homozygote. In contrast, the genotypic designation 11,0 for a low peak allows the possibility of any other allele, including another 11, in the genotype of the contributor.
2.1. Drop-out and drop-in probabilities

We will write \( D \) for the probability of drop-out of one allele of a heterozygote, and we assume that drop-out is independent across the two alleles of a heterozygote and also across loci. The value of \( D \) at a locus could depend on the peak height if a single peak is observed, and could increase with allele length, because degradation affects longer alleles more than shorter alleles. Here we follow previous authors [8,16] and assume that \( D \) is constant, but what follows is readily extended to allow \( D \) to vary over alleles.

Some theory and data are emerging to assist in the task of choosing a suitable estimate for \( D \) [8,27], but we agree with [5] that “judgment calls are still required by the reporting scientist”. Assessments of \( D \), whether quantitative or informal, can be based on the stochastic phenomena displayed in the epg over all loci, and informed by the results of laboratory trials that employ LTDNA profiling on samples of known origin and mixture proportions. It will usually be impossible to specify a precise value, and instead a range of plausible values should be considered. Fortunately, as noted by others [8,17], the \( LR \) often does not vary greatly over the range of plausible values for \( D \). If a point estimate is required, because the whole-profile \( LR \) can involve high powers of \( D \), and because the average of a power is larger than the power of the average, appropriate point estimates for \( D \) should typically be towards the high end of the plausible range.

We write \( D_2 \) for the homzygote drop-out probability. It is natural to seek to express \( D_2 \) in terms of \( D \). One possibility is to assume that the allele on each homologous chromosome drops out independently with known probability \( D \), so that \( D_2 = D^2 \) [2,8]. This is likely to overstate \( D_2 \), if both alleles can generate partial signals that combine to reach the reporting standard, whereas each individual signal would fail to reach this standard. For example, suppose that under the prevailing conditions a heterozygous allele would generate a peak with mean height 40 RFU and that \( D = 0.7 \). If the signal from a homzygote is a superposition of two signals, with expected height close to 80 RFU, the corresponding drop-out probability is likely to be much less than \( D^2 = 0.49 \). In 40 laboratory trials of 10 pg DNA samples, 9 instances of homzygote drop-out were observed, from which \( D_2 \approx 0.225 \). The value of \( D \) estimated from 160 heterozygote alleles at the same three loci was 0.66, so \( D^2 \approx 0.44 \) and hence \( D_2 \approx 1/2D^2 \) (Simon Cowen, personal communication). On the other hand, at very low template levels the epg peak for each allele may be closer to “all or nothing”, in which case the independence assumption may be reasonable. Here, we assume that \( D = \alpha D^2 \). The appropriate value of \( \alpha \) should be chosen on the basis of laboratory trials. Typically we expect that \( 0 < \alpha < 1 \), and we use \( \alpha = 1/2 \) and \( 1 \) in the numerical examples below.

2.2. Drop-in

Here we assume that at most one drop-in occurs per locus, with probability \( C \), but this restriction is easily relaxed. We also exclude the possibility of a drop-out followed by drop-in of the same allele. Following [8], we treat drop-ins at different loci as mutually independent, and independent of any drop-outs. Multiple drop-ins in a profile may, depending on the values of \( C \) and \( D \), be better interpreted as an additional unknown contributor. Typically we use \( C \leq 0.05 \) below, reflecting that drop-in is rare [8], and such low values of \( C \) automatically penalise a prosecution hypothesis requiring multiple drop-in events.

2.3. Allele frequencies and shared ancestry

In the examples below, to make some allowance for sampling uncertainty we use \( Pr(X = (n_X + 4)/(n + 4)) \) if \( s \) is a \( XX \) homozygote, \( Pr(X = (n_X + 2)/(n + 4)) \) for each allele of a heterozygous \( s \) and \( Pr = n_s/(n + 4) \) if allele \( X \) is not included in the profile of \( s \), where \( n_s \) is the count of \( X \) in a database of size \( n \); see [18] for a discussion and [19] for alternative approaches.

We do not here consider alternative sources of the CSP that are direct relatives of \( s \), even though we regard this as an important possibility in practice. To allow approximately for the effects of shared ancestry between \( s \) and other possible sources of the CSP, we replace the allele proportion \( P \), adjusted for sampling error as described above, with \( (2F_{ST} + (1 - F_{ST})p)/(1 + F_{ST}) \) for each allele when \( s \) is heterozygote, and with \( (2F_{ST} + (1 - F_{ST})p)/(1 + F_{ST}) \) in the homozygote case [18,20]. For alleles not in the profile of \( s \), the allele proportion \( P \) is replaced with \( (1 - F_{ST})p/(1 + F_{ST}) \). See [20] for a discussion of the appropriate value of \( F_{ST} \). We recommend using a value of at least 0.01 in all cases, and use 0.02 in numerical examples below. In the numerical examples below, genotype proportions are obtained from allele proportions assuming Hardy-Weinberg Equilibrium.

3. \( LR \) for a single contributor

Explicit formulae can be complicated, but logically the computation of the \( LR \) when drop-out and drop-in are possible is straightforward [7]: the denominator involves a summation over every genotype for the unknown contributor, while the numerator consists only of the term in the denominator corresponding to the profile of the alleged contributor, \( s \). To illustrate notation and concepts, consider first a single CSP that shows only an \( A \) allele, whereas the genotype of the alleged contributor \( s \) obtained from a good-quality reference sample that is assumed to be profiled without error, is \( AB \). This would normally constitute an exclusion of \( s \), but if \( D > 0 \) then the prosecution hypothesis, \( H_s \), that \( s \) is the source of the CSP, can be sustained. Writing “\( \approx \)” for “has profile”, and assuming that \( Pr(s \approx AB) \) does not depend on whether \( H_s \) or \( H_i \) is true, the \( LR \) can be written

\[
LR = \frac{Pr(CSP = A|s = AB, H_s)}{Pr(CSP = A|s = AB, H_i)}
\]

in which \( H_i \) is the alternative hypothesis that \( s \), unrelated to \( s \), is the source of the CSP. Large values of the \( LR \) imply strong evidence against \( s \), provided that all his relatives can be excluded as alternative possible contributors to the CSP; otherwise the \( LR \) for various relatives of \( s \) should also be provided.

Under \( H_i \) we have observed a non-drop-out allele (\( A \)), a drop-out (\( B \)), and no drop-in, so that the numerator is \( DDC \), where we introduce the notation \( X \) for \( 1 - X \). In the denominator, if no drop-in has occurred then \( i \) can have genotype \( AA \) or \( AQ \), where \( Q \) denotes any allele other than \( A \), whereas if the \( A \) allele has dropped in then \( i \) can be \( QQ \) or \( Q \), where \( Q \) is neither \( A \) nor \( Q \). The \( LR \) is then:

\[
LR = \frac{DDC}{(D_2P_{AA} + DDP_{AQ}C + (D_2P_{QQ} + D^2P_{QQ})CP_A)}
\]

where \( P_{XY} \) and \( P_X \) denote the probabilities of genotype \( XY \) and allele \( X \), given the alleles observed in the CSP. Note that the denominator of (1) does not depend on the profile of \( s \) except through the \( P_{XY} \), for which dependence can arise through possible shared ancestry [18,20]. Eq. (1) is similar to the formula set out set out in Table 1 of [8], except that we distinguish homzygote drop-out as a special case, and also we have corrected an error in the first term of the denominator.

In (1), \( LR \rightarrow 0 \) as \( D \rightarrow 0 \), corresponding to strong evidence for exclusion of \( s \) if drop-out is very unlikely. This is intuitive since, in this scenario, a case against \( s \) cannot be sustained unless drop-out has occurred. Several authors [3,11] have noted that the “2p rule” overstates the correct \( LR \) substantially for low values of \( D \), but Fig. 2 shows that the effect can be more substantial than previously noted, especially in the presence of masking, discussed further below. It has
not previously been noted that \( LR \to 0 \) as \( D \to 1 \), which occurs in (1) if \( \alpha < 1 \). This arises because, if drop-out is very likely, observation of a single \( A \) allele suggests that the contributor to the CSP has genotype \( AA \), so once again the evidence points away from \( s \) who is \( AB \).

One important conclusion, here and in what follows, is that if drop-out is invoked to sustain the prosecution case then estimation of drop-out probabilities cannot be avoided. DNA-based prosecutions that rely on drop-out and do not explicitly estimate plausible ranges for the drop-out rate parameter are, in our view, defective.

4. LR for multiple contributors, one unknown

The two cases introduced above both involve mixed profiles with a known contributor, \( v \), accepted by both prosecution and defence. Alleles from \( v \) can mask alleles from the unknown contributor in each case, either directly or via stutter peaks. We initially ignore the latter, but return to them below.

4.1. One minor-component allele

Let \( M \) denote the set of major component alleles from known contributors, while \( s \equiv AB \), the CSP minor component is \( A \), and neither \( A \) nor \( B \) lies in \( M \). The numerator is the same as in (1), but under \( H_i \) we must consider the possibility that alleles of \( i \) lie in \( M \) and hence are masked. Again using a derivation similar to that of (7), the \( LR \) is

\[
LR = \frac{Pr(CSP = A | s = AB, M, H_i)}{Pr(CSP = A | s = AB, M, H_i)} \times \frac{DDC}{(D_2P_{AA} + DD_P_{AQ} + D_P_{AM})C + (D_P_{QQ} + D^2P_{QQ} + P_{MM} + D_P_{MQ})CP_A} \quad (2)
\]

where \( Q \) and \( Q' \) denote distinct alleles that are both neither \( A \) nor in \( M \). As expected, (2) reduces to (1) if \( M \) is null. Once again, \( LR \to 0 \) in (2) both as \( D \to 0 \) and, if \( \alpha < 1 \), as \( D \to 1 \).

If the genotype of \( s \) is now \( AA \) while the CSP minor component is still \( A \), the numerator of (2) becomes \( D_2C \), while the denominator remains unchanged (the values of the \( F_{ST} \) change because of the changed effect of shared ancestry of \( i \) with \( s \)). A possible approximation is to assume that \( D = C = 0 \), so that

\[
LR = \frac{1}{P_{AA} + P_{AM}}.
\]

This would be acceptable for a good-quality profile, but in the LTDNA setting it is usually unfair to defendants. The RMNE approximation \( LR = 1/P_{AA} \) is usually conservative in this setting (Fig. 3, top).

4.2. Two minor-component alleles

If \( s \equiv AB \) and the minor component CSP \( \equiv AB \), it may seem natural to use the RMNE value \( LR = 1/P_{AA} \). However, masking and the possibility of drop-in can make this value unfair to defendants. In fact,

\[
LR = \frac{D^2C}{P_{MM} + D_P_{MQ} + D_2P_{QQ} + D^2P_{QQ}} \quad (3)
\]

where \( Q \) denotes an allele that is not in \( \{A,B\} \cup M \). Only the first term of the denominator of (3) is included in an RMNE calculation. Fig. 3 (middle) shows a large effect of the additional terms when \( D \) is not small, but masking has little effect.

4.3. Null minor component

The effect of masking can be very important when the minor component CSP is null, especially if \( D \) is small. For \( s \equiv AB \) the \( LR \) is

\[
LR = \frac{D^2}{P_{MM} + D_P_{MQ} + D_2P_{QQ} + D^2P_{QQ}} \quad (4)
\]

where \( Q \) and \( Q' \) are any alleles not in \( M \). If \( s \equiv AA \), then \( D_2 \) replaces \( D^2 \) in the numerator. In either case, (4) favours the defence hypothesis when there are masking alleles, strongly so unless \( D \) is large (Fig. 3, bottom). Further, even if \( M \) is also null so that the first two terms in the denominator of (4) are zero, \( LR \neq 1 \) unless \( \alpha = 1 \). If \( \alpha < 1 \), the observation of a null CSP supports the conclusion that \( i \) is heterozygous and so \( LR > 1 \). Conversely, if \( s \equiv AA \) then \( D_2 \) replaces \( D^2 \) in the numerator of (4) and \( LR < 1 \) when \( \alpha < 1 \). In contrast, the RMNE value is always 1, irrespective of masking and the genotype of \( i \).

5. Multiple unknown contributors

The principles that we have used to construct \( LR \)s, based on those set out in [7] and implemented in [8], extend readily to multiple unknown contributors. The denominator now involves summations over the possible genotypes for each unknown contributor. The result can be approximated by averaging over repeated simulations of their genotypes, which can reduce the computational burden at the cost of some approximation. However, for one or two unknown contributors, the exact value is fast to compute and so simulation-based approximation is not necessary. Our two example cases are assumed to involve only one unknown contributor, and to keep the exposition as simple as possible we limit attention to this case.
6. Uncertain allele calls

6.1. Stutter

Stutter peaks are usually located one repeat unit shorter than the allelic peak, but stutters at minus two or plus one repeat units are sometimes observed. Interpretation guidelines often specify a threshold height, as a proportion of the allelic peak height, above which a peak in a stutter position is called as an allele, and otherwise is regarded as stutter \([12]\). Inevitably, such thresholds do not always correctly distinguish allelic from stutter peaks, as noted by \([11]\) at Recommendation 6. Calling a peak as stutter tends to favour \(s\) if he has the allele, but if not the apparent stutter could mask a minor-component allele that, if reported, would tend to exonerate \(s\). Conversely, if a possible stutter is called as allelic, this favours \(s\) if he has no allele at that position, but otherwise is unfair to \(s\) because the allele is counted as evidence against him with no allowance for the possibility that it could be just a stutter artefact.

Stutter is most problematic when there are multiple contributors to the CSP, but it can also cause problems for single-contributor LTDNA profiles. For example, a low peak at allele \(X - 1\) and a higher peak at \(X\) could be consistent with an \(XX\) contributor and substantial stutter, or an \((X - 1)X\) contributor with unbalanced peaks, or an \(XY\) contributor with both stutter and drop-out of the \(Y\). Gill et al. \([5]\) proposed upper and lower bounds on the LR obtained by considering the possible assignments of a peak in a stutter position. Below we propose weighted averages of LR numerator and denominator, similar to \([21]\).

6.2. LTDNA masking

In the US case the known contributor’s DNA is also at low template and subject to drop-out, and so masking is uncertain: the appearance in the epg of a peak in a masking position could partly reflect an allele of the unknown contributor. The simplest way to deal with uncertain masking, usually erring in favour of \(s\), is to include in the masking set \(M\) every above-threshold peak that either corresponds to an allele of a known contributor or is in a stutter position to such an allele. Below we develop a more satisfactory quantitative analysis of uncertain masking, by assigning probabilities that an allele is masking.

6.3. Near-threshold peaks

An allele of \(s\) that is not reported in the CSP tends to exonerate, and this effect can be large when drop-out is unlikely. Intuitively, the effect is tempered if there is an apparent peak in the epg at the appropriate position, but which fails to achieve the reporting threshold. An ideal analysis would take into account all available information in a continuous way, without reference to thresholds. In practice it is extremely difficult to take into account all the information contained in a complex epg, and no algorithm for this
Currently seems to be available although there have been promising developments [22,23].

In principle, a full LR analysis of a sub-threshold peak requires probabilities of observing the peak if the unknown contributor does, and does not, have an allele at this position. Intuitively, if there is a peak at A in the CSP epg that is well above the reporting threshold, while no sub-threshold peaks are obvious, then the locus should be exculpatory if s is AB. If, however, the A peak is only just above threshold, while a peak at B is clearly visible just below the threshold, then the locus might informally be regarded as inculpatory, even though one allele is not reportable. Our analysis at (1) does not discriminate between these two possibilities and makes no direct use of the heights of reported peaks or any apparent sub-threshold peaks.

6.4. Probabilistic allele designations

Towards a more satisfactory analysis of uncertain stutter, masking and near-threshold allele calls, we propose that, rather than just make a single judgment “call” of an epg to designate peaks as stutter, masking or allelic, an expert assesses probabilities for alternative possible designations. In computing the LR, both numerator and denominator must be averaged over the possible calls, weighted according to their respective probabilities (NB this is not equivalent to a weighted average of LRs). Most laboratories would have information available to help assess such probabilities. Typically the baseline noise has been quantified in general, in order to establish the drop-out threshold, and further information could come from assessing the baseline for the epg at hand.

For example, if s = AB and the CSP peak at allele A is above the reporting threshold, while that at B is slightly sub-threshold, an expert may assess probabilities of 1 and 0.6 that the peaks at A and B are allelic. The numerator of the LR is then 0.4 times the numerator of (2) plus 0.6 times the numerator of (3). Similarly, the denominator is 0.4 times the denominator of (2) plus 0.6 times that of (3). Fig. 4 illustrates the effect of averaging over two equi-probable allele calls.

7. Replication

If replicate epgs are available for the same crime sample, a common approach is to decide on a consensus profile and then to compute an LR for it, without explicitly modelling the details of the replication. Another possibility is to use the results of replicate epgs to inform the assessment of allele calling probabilities, described above.

If the replicates can be assumed to be independent, it is straightforward to extend our LR formulae to simultaneously analyse all the replicates [6,7]. The denominator is still a sum over possible genotypes for the unknown contributor, i. Every term in this sum includes a probability for that genotype, times the product of probabilities for the drop-outs and drop-ins required in all replicates in order to reconcile this genotype with the observed CSP alleles. The numerator consists of the term in the denominator for which i has the same genotype as s.

We applied this approach to the example in Table 2 of [8], which has two replicate epgs for a CSP that we assume has two known and one unknown contributor, all subject to drop-out. Following [8], we assumed a common D for all contributors. We also modified our formulae to allow up to two drop-ins per locus, as this was required under H, at locus D18 (we allowed for the possibility at every locus). Our result has features in common with that shown in Fig. 1 of [8]: the LR is maximised at very low values of D (around D = 0.005) and declines rapidly with increasing D. However, our computed LR is several orders of magnitude smaller: our maximum value is around 10^2 versus 10^12 in [8]. The source of that discrepancy is currently under investigation. In contrast, the Locomation software described in [8] gave similar results to those reported below for the Bates case (James Curran, personal communication).

8. Results

8.1. Garside and Bates

The most incriminating of the many CSPs presented in evidence is summarised in Table 1. The prosecution in effect presented an RMNE computation (see Table 2) applied to the CSP alleles not masked by the known contributor. In general, this approach is favourable to defendants if drop-out is not required under the prosecution case, but it does not use any value for D and we have noted above that it can be unfair to defendants in the presence of discordant alleles. The reported value was 0.61 million (M). We perform a similar calculation (Table 1), but because we have used different FSR and sampling adjustments, our corresponding value is 0.33M.

Fig. 5 shows single-locus and overall LRs as functions of D, using allele frequencies reported in [24]. Because of the three discordant alleles, the overall LR is zero (exclusion of Bates) if D = 0. However, from Fig. 5 we see that the LR rapidly increases with D. Using our LR calculations, the RMNE value is in fact conservative for 0.02 < D < 0.98. Given the small number of alleles in the minor component CSP and the stochastic nature of the results of repeat profiling runs, it would not have been difficult to argue that D lies in this range. But in the absence of any such argument, the claim that the LR was conservative was not adequately justified.

The RMNE value of 1 for locus D18 is not conservative for any value of D, and Table 1 illustrates that the non-conservative error can be small or large depending on the value of D. This conclusion conflicts with [12], who asserted in the context of the Bates case that drop-out of both alleles “must be regarded as neutral”, but did not consider the effect of masking alleles that were present in this case. The prosecution in Bates defended its assumption by arguing that drop-out is more likely at locus D18 than at other loci because of its high molecular weight, but provided no quantitative backing for this argument. We have here assumed the same drop-out and drop-in probabilities at each locus.

At locus D2 the “2p rule” gives 11 (Table 1), which is non-conservative compared with an LR of 2.8 (D = 0.05) or 9.3 (D = 0.5). In fact, a peak was observed at the 19 allele, but did not reach a height such that it could be confidently distinguished from stutter artefacts generated by Mrs Garside’s two 20 alleles. Treating the 19
null, result (i.e. the CSP includes only the 14 allele illustration let us assume that there were two replicates both replicates because of varying conditions and samples, but for various conditions and in none of these replicates were the 12 or 15 allele reported at D18. Repeat epgs were usually not strict allowance for drop-out when two or more alleles are observed, nor for drop-in. In particular, although \( \nu \) is an acknowledged contributor to the CSP and \( s \) is an alleged contributor, because the calculation makes no allowance for drop-out, neither \( \nu \) nor even \( s \) is included in the 1 in 96K proportion of the population evaluated by the multi-source RMNE. Thus the RMNE value bears little relevance for evidential weight, and it could be highly misleading to a court.

In fact, at four of the loci included in the RMNE computation, all the alleles in the CSP can be attributed to the known contributor \( v \), and so these loci provide little if any evidence for \( s \) being a contributor to the CSP. Removing these four loci brings the RMNE value from 1 in 96K to 1 in 47. More appropriate is to recognise that \( \nu \) is a contributor to the CSP and use the single-source RMNE calculation that is based only on the (three) alleles in the CSP that could have come from \( s \) but not from \( \nu \), giving an RMNE probability of about 1 in 8 (Table 3, final column). However, an LR calculation shows that even this much reduced value still overstates evidential strength, because of the five loci at which the evidence favours the proposition that \( s \) is not a contributor to the CSP; no RMNE can accommodate this possibility. In fact (Table 3 and Fig. 6), under our assumptions the LTDNA evidence only very weakly supports the prosecution case, with \( LR < 3 \) when \( \alpha = 1/2. \) We used the California Caucasian allele frequencies reported by [25], with a sampling adjustment and \( f_{CT} = 0.02. \) We also assumed the same value of \( D \) for both sources of the CSP.

The five loci at which the CSP showed no alleles are all slightly discriminating under our analysis (\( LR \approx 1.1 \) when \( \alpha = 1/2 \)), because \( s \) is heterozygous at each of these loci, whereas an alternative contributor might be homozygous and hence less susceptible to drop-out. Our novel treatment of homozygote drop-out is potentially important here because if \( \nu \) is an acknowledged contributor, because the calculation makes no allowance for drop-out, neither \( \nu \) nor even \( s \) is included in the 1 in 96K proportion of the population evaluated by the multi-source RMNE. Thus the RMNE value bears little relevance for evidential weight, and it could be highly misleading to a court.

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For example, if \( C = 0.02, \) \( D = 0.05, \) and \( \alpha = 1/2 \), and the locus has 10 alleles with genotype proportions equal to 0.01 for each homozygote and 0.02 for each heterozygote, then the LR calculated from the above two formulae are, respectively, 0.118 and 0.121. Thus, for these parameter values, the evidence still points away from \( s \) and towards \( i \), despite one observation of a 12 allele.

### 8.2. US case

The prosecution proposed an RMNE probability of 1 in 96K for Caucasians (Table 3, “multi-source RMNE” column), based on the seven loci (D8, D3, D16, D19, VWA, TPO, D5) at which the CSP included an allele that could be attributed to the suspected contributor, \( s \). Unlike the single-source RMNE used in Bates, the probability computed was for the profile of a random man to be included among all the CSP alleles at each locus, without distinguishing known from unknown contributors. For zero or one observed allele, the computation is the same in each approach. For two observed alleles, \( A \) and \( B \), the multi-source RMNE value of \( P_{PA} + P_{PA} + P_{PB} \) exceeds the value \( P_{PA} \) for the single-source RMNE. This approach may seem generous to defendants but it makes no allowance for drop-out when two or more alleles are observed, nor for drop-in. In particular, although \( \nu \) is an acknowledged contributor to the CSP and \( s \) is an alleged contributor, because the calculation makes no allowance for drop-out, neither \( \nu \) nor even \( s \) is included in the 1 in 96K proportion of the population evaluated by the multi-source RMNE. Thus the RMNE value bears little relevance for evidential weight, and it could be highly misleading to a court.

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alleles). Conversely, however, the nine CSP alleles attributed to \( \nu \) that are not shared by \( s \) may represent a contribution from a different contributor and so could be regarded as providing evidence for the defence. In accord with the principle that CSP allele calls should not refer to the profile of \( s \), we recalculated the LR treating all the alleles of \( \nu \) that appear in the CSP as masking with probability 0.5 if \( \nu \) is heterozygous, and with probability 1 if \( \nu \) is homozygous. This has the effect of weakening the prosecution case: the overall LR of 2.0 in Table 3 is reduced to 1.2. Thus, attributing CSP alleles to \( \nu \) whenever possible, and ignoring the possibility that they could be from the unknown contributor, is here favourable to the prosecution.

Although our analysis has limitations as discussed above, we believe that it suffices to establish that the LTDNA evidence at best only very weakly supports the prosecution case. Further, an uncritical application of RMNE can very significantly misrepresent the evidence.

### 9. Software

One of us (DJB) has written R code (R is freely available with documentation at [www.r-project.org](http://www.r-project.org)) that computes single-locus LRs for a CSP with one unknown contributor, allowing for drop-out of one or both alleles, and drop-in of up to two alleles. The drop-out and drop-in probabilities can vary over loci but must currently be constant within a locus (it would not be difficult to alter the code to allow allele-specific probabilities). One version of the code allows for uncertain calls of stutter, masking, and near-threshold peaks, while another allows for replicate calls.

The software has been used to generate all the figures in this paper. It is available on request and has been checked extensively but comes with no guarantee of accuracy. It should be regarded as research-level software: we do not currently have the resources to test it fully rigorously, nor to make it widely accessible.

### 10. Conclusion

We have built on the contributions of previous authors to develop an approach to analysis of LTDNA profiles. In this approach, drop-out and drop-in probabilities are parameters that need to be assigned or estimated, as in [8] and its precedents [2,6,7]. An alternative approach would be to build into the LR calculation an automatic estimation of some of these parameters. Even better would be to model the entire epg, taking all peak heights into account without reference to thresholds. Although these approaches may have advantages in principle, they would be difficult to implement.

We allow uncertain allele calls for stutter, masking and near-threshold peaks by averaging over the possible calls when computing the numerator and denominator of the LR. Here we have limited attention to cases involving only one unknown contributor, but the principles extend to multiple unknown contributors [8]. We believe that the approach developed here overcomes many of the challenges presented by LTDNA evidence, and can be explained and defended in court.

We deplore uncritically ignoring loci that are discordant with the prosecution case, and recommend that drop-out and drop-in probabilities be assessed, and their implications be considered. We have also shown through argument and example that the Random Man Not Excluded (RMNE) approach to evidence evaluation has serious flaws in the LTDNA setting, and can potentially lead to serious misrepresentation of the evidence; we recommend that it not be used. Recently [26] have described an RMNE calculation allowing for up to a given number of drop-out alleles. They give no numerical comparison with an LR approach, and they acknowledge that their approach does not make full use of the evidence. In fact, for multi-source CSPs when multiple drop-out is allowed, the approach loses a large amount of evidential information. It also remains subject to the many criticisms of RMNE described above: evidence can never favour a defendant, and it is difficult to interpret for multi-contributor samples because no use is made of the suspect’s profile.

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### References


