DEVELOPMENT UNDER EXTREME CONDITIONS:
FORENSIC BIOINFORMATICS IN THE WAKE OF THE
WORLD TRADE CENTER DISASTER

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The terrorist attacks of September 11, 2001 resulted in death and devastation in three
locations, and extraordinary efforts have been exerted to identify the remains of all victims.
As mass fatalities go, this one has been unusual at a policy level because the goal has been
not merely to identify remains for every decedent, but to identify every bit of remains found
so that even small pieces of tissue can be returned to families for burial. While the human
impact at the Pentagon and Shanksville, PA was horrific, the World Trade Center site
presented a particularly complex challenge for forensic DNA matching and data handling. A
complete and definitive list of all those killed is still elusive, and human remains were
crushed and co-mingled by the falling towers. Software tools had never been considered for
a problem of this scale and scope. New data handling systems had to be created under
extreme software development conditions characterized by incomplete requirements
specifications, chaotically changing priorities, truly impossible deadlines and rapidly rolling
production releases. Partly because of the company’s experience with mtDNA tools built for
the Armed Forces DNA Identification Lab starting in 1997, the New York City Office of
Chief Medical Examiner [OCME] contacted Gene Codes Corporation in late September as
existing data-handling tools began to fail. We began work on the project in mid-October,
2001. Our approach to the problem included:

• Extreme Programming [XP] methodology for functional software development,
• On-site time and motion analysis at the OCME for user interface design,
• Evidentiary references between STR, SNP and mtDNA analysis results, and
• Separate data Quality Control [QC] and software Quality Assurance [QA] initiatives.

A substantial software suite was developed called M-FISys, an acronym for Mass-Fatality
Identification System.

1 Background

The New York City Office of Chief Medical Examiner [OCME] faced an
unprecedented problem after September 11, 2001. Early estimates suggested that
over 10,000 people had been killed in the attacks at the World Trade Center (this
number would be revised down to 2,801 by the first anniversary of the disaster).
Forensic identification of remains would be complicated by the substantial
fragmenting of remains (with individuals recovered in as many as 200 pieces) and
the heat and moisture associated with the fighting of jet fuel fires that burned for
months. Certainly some fatalities would be identified by classical methods such as
viewed remains, fingerprints, dental records and personal property. Because of the
crushing and co-mingling of remains, personal belongings found at “Ground Zero”
have been considered highly suspect as identifying evidence.\(^a\) The vast majority of individual remains would have to be identified by DNA matching, and the associated data handling is the subject of this paper. Because of the public interest that we all share in this case, attention will be given to the circumstantial pressures that impacted the software development process.

Functionality had to be added to the M-FISys program [pronounced like *emphasis*] on a very fast schedule without sacrificing software quality and testability. In fact, a new release of M-FISys has been delivered to the OCME almost every week since mid-December, 2001 (the 38th iteration is being released at the time of this writing). The need for stringent software quality control is self-evident and the moral magnitude of the task plus the knowledge of the catastrophic impact of any false identification was a constant weight on the shoulders of the developers. Extreme Programming [XP] \([14,15,16,17,18]\) has been an invaluable methodology under the existent pressures. Since almost all studies of software correctness indicate that code reviews are one of the most valuable QA tools, XP programmers work in pairs, with one engineer constantly reviewing the work of the other. Both unit tests and acceptance tests must be written before functionality can be added. In order for new work to be checked back into the source code control system, not only must it pass these tests, but it must pass all other tests that have been written against all functions since the project began. An obvious benefit of this discipline is that it makes it difficult to unknowingly “break” a part of the program while fixing a seemingly unrelated bug. It can also add a cost any time there is a major architectural refactoring because many of the legacy tests will have to be re-written to accommodate the new architecture.

Under the XP methodology, functionality is added each week through observation and direct negotiation with users in the field. One or two staffers have been on-site daily at the OCME since the project began, and the chief designer has traveled back and forth between the user location in Manhattan and the developer site in Ann Arbor each week. Weekly reviews identify opportunities for process improvement. Productivity of the engineering is carefully monitored to the point that one can literally choose any week since the project began, recover the source code as it was during that period, identify what new functionality was scheduled to be added, which engineers were available, initial time estimates for tasks and review the actual programming velocity of the entire team for that iteration.

An important principal of our methodology has been to rigorously resist the urge to propose functionality to the OCME. This is very different from taking the luxury of adding our own “helpful” tools as we would in a commercial design and

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\(^a\) To illustrate the complexity of the investigation, it is worthwhile to note that a number of remains would be identified by tracing serial number on implants such as cardiac pacemakers and artificial joints, contacting the manufacturer (sometimes overseas), tracing the hospital to which the item had been sold, and then cross-referencing all implantation surgeries against the list of missing persons from the Trade Center.
development project. Our process analysts work through proposals based on the immediate needs and priorities of the forensic biologists, and while we take responsibility for architecture and initial interface design, we do not allow the engineering team to pursue its own vision for the program’s functionality. It takes effort for skilled and creative engineers to stick to this discipline, but it is critical to keeping up our development velocity. Outside contractors and advisors have become frustrated and occasionally openly hostile when pet proposals have not made it onto the development list because OCME staff have evaluated those proposals as adding insufficient value. Certainly the emotional content of this development project is a major component of the managerial complexity.

2 Methods of Identification

The most widely used forms of DNA-based human identification involve Short Tandem Repeat [STR] analysis at 13-15 nuclear loci. A second procedure involves sequencing the hypervariable regions of the mitochondrial genome. A relatively new forensic procedure, pioneered by Orchid Biosciences, is based on Single Nucleotide Polymorphisms [SNPs] with well-characterized inheritance and frequency patterns. All of these techniques had to be combined in the M-FISys program.

The theoretical basis of STR analysis in human identification can be traced back to the work of Alec Jeffreys [24]. In modern applications, 13-15 unlinked STR loci are sized, resulting in a “profile.”

Table 1: Sample STR data

<table>
<thead>
<tr>
<th>STR Locus</th>
<th>Value 1</th>
<th>Value 2</th>
</tr>
</thead>
<tbody>
<tr>
<td>D3S1358</td>
<td>15/16</td>
<td>16/17</td>
</tr>
<tr>
<td>vWA</td>
<td>26/28</td>
<td>14</td>
</tr>
<tr>
<td>FGA</td>
<td>14</td>
<td>30/322</td>
</tr>
<tr>
<td>D8S1179</td>
<td>15/16</td>
<td>12/13</td>
</tr>
<tr>
<td>D21S11</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>D18S51</td>
<td>9/12</td>
<td>9.3</td>
</tr>
<tr>
<td>D5S818</td>
<td>6/9</td>
<td>10/11</td>
</tr>
<tr>
<td>D13S317</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>D7S820</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>D16S539</td>
<td>9/12</td>
<td>9.3</td>
</tr>
<tr>
<td>THO1</td>
<td>6/9</td>
<td>10/11</td>
</tr>
<tr>
<td>TPOX</td>
<td>6/9</td>
<td>10/11</td>
</tr>
<tr>
<td>CSF1PO</td>
<td>6/9</td>
<td>10/11</td>
</tr>
</tbody>
</table>

To orient the reader, the above profile can be read as follows: at the FGA locus, this person inherited 26 repeat-elements from one parent and 28 from the other. The D8S1179 locus is homozygous for 14 repeats. Based on experimentally observed and published frequencies of each repeat value in various ethnic populations, this profile would be expected to be seen no more than one in $10^{17}$ individuals.

The likelihood value for a given STR profile is the product of the likelihoods of each of the 13 STR loci. As likelihood values are inverses of frequencies, we can

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b In addition to being chosen for genetic distance and independence, the core loci have been selected to be medically uninformative, thereby protecting the medical privacy of individuals in a forensic or criminal investigation.
justify a simple multiplication since the value at each locus is an independent event, unrelated to the other loci.

The benefit of this is that we can have comparative likelihoods across partial profiles. If a policy decision is made to set a threshold likelihood in order to report an identification, for example a likelihood of no less than $10^{10}$, an incomplete STR profile's likelihood may still surpass this value if the available alleles are rare enough.

The likelihood value for a given profile will differ depending on ethnic population; however, access to that information is not always available or reliable where badly damaged remains are involved. We therefore assume the "worst case scenario" by examining the likelihood values across four races (Asian, Black, Caucasian & Hispanic) and take the lowest value as the final likelihood. By using this lowest value, we err on the side of preventing false positives.

To determine the likelihood of a particular locus, we take the inverse of its frequency. This frequency can be first approximated by ignoring population structure and using the Hardy-Weinberg proportions:

- $p^2$ for homozygous alleles, where $p =$ probability of the allele
- $2pq$ for heterozygous alleles, where $p,q =$ probabilities of each allele

However, since all humans are related, variance of allele proportions is factored in with the inbreeding coefficient $\theta$ (theta), with:

- $p^2 + p(1-p)\theta$ for homozygous alleles
- $2pq(1-\theta)$ for heterozygous alleles

As stated in *Interpreting DNA Evidence* by West & Weir, “$\theta = 0.03$ is a conservative upper bound on the values appropriate for human populations (National Research Council 1996).”[1] For the WTC project, we use the value of 0.03 for $\theta$.

The probabilities $p,q$ are determined empirically themselves, based upon population. (The values we used are in the STR13.xls spreadsheet provided by Dr. Howard Baum, Deputy Director of Forensic Biology at the New York City OCME. [12])

Let $S$ be any STR profile containing $A_k$ alleles, $k<=13$. Furthermore let $p_k$ be the probability frequency of the high allele of $A_k$, and (if heterozygous), let $q_k$ be the probability frequency of the low allele of $A_k$. Then the Likelihood of $S$, $L(S)$, is defined as:

$$L(S) = \prod_{k \in \text{Alleles}} \frac{1}{P(A_k)}$$

$\text{c} \text{ The Medical Examiner's Office in New York chose to use the equation involving } \theta \text{ only for the homozygous alleles, so that the Likelihood values were always more conservative.}$
where

\[ P(A) = \begin{cases} 
  p^2 + p(1-p)\theta & \text{homzygous} \\
  2pq & \text{heterozygous} \\
  1 & \text{non-existent}
\end{cases} \]

for each k and \( \theta = 0.03 \).

3 Implementation and Design Decisions

We will first address the issue of matching profiles that are complete (all 13 loci plus a gender locus) or nearly so. Limit this discussion to those cases where profiles have a likelihood of no less than \(10^{-10}\).\(^d\) Even at this level of match stringency, the first task was to address the information glut of having 25, 50 or 100+ individual samples that had identical profiles because they were fragments of the same person.

For the first three weeks after the disaster, attempts to match remains used a software package called CoDIS (Combined DNA Index System). CoDIS was designed as a criminal investigation tool to compare a single DNA profile to a database of profiles from crime scenes and from convicted felons.

CoDIS, the result of a federally funded software development project, is difficult to use under ideal circumstances. Early attempts to apply it to the World Trade Center disaster were described by the forensic scientists as a disaster in itself. The program generated up to 4,000 printed pages of internally consistent matches (matches between samples and between samples and exemplars). Teams of forensic scientists would literally spend days pouring over CoDIS output with highlighter and ballpoint pens in an attempt to sift through the information. The usability issue of reducing the size of this haystack was clearly a first priority.

The first working version of M-FISys, delivered on December 13, had the immediate advantage of collapsing multiple samples with the same profile into an expandable “aggregate” sample. This alone was such an improvement over the CoDIS-based methodology that 55 new ID’s were recognized the first day the program was installed. In Figure 1, the line labeled RM #34 (6) indicates that there are six internally consistent profiles in the collapsed set.\(^e\)

\(^d\) As suggested earlier, the threshold of \(10^{-10}\) was a policy decision. The basis of the selection was that in a population of 5,000 victims (the estimate of fatalities at the time of the discussion), this would reduce the likelihood of even a single misidentification based on a fortuitously shared STR profile to less than 1 in 1,000,000.

\(^e\) RM stands for Reported Missing. Each presumed victim has an RM number. These numbers have been changed and names obscured for purposes of this paper.
Data reaches the DNA ID group at the OCME in three categories: 1) DNA profiles from victim remains, 2) ante-mortem DNA profiles of personal references from missing persons (e.g., a pre-existing pathology sample, a toothbrush or a lipstick) and 3) profiles from buccal swabs taken from first-order relatives. The first attempt is to match a victim sample to a personal reference exemplar.

Unlike a criminal investigation application where a single sample might be compared to the database, the WTC recovery required repeated all-against-all comparisons as additional samples were recovered and analyzed in the lab. Furthermore, as compromised samples were reanalyzed with hopes of extracting additional information, “Virtual Profiles” needed to be created, showing all accumulated values while keeping track of exactly which laboratory attempt had generated which data points. As data is added and operator decisions are made about assignment of data, a history (necessarily including free-text annotations) is kept for each sample.

The possibility of fortuitous (coincidental) matches increases as match stringencies are lowered. For samples with low partial profiles (likelihoods in the $10^3$ – $10^4$ range) one might expect several possible direct matches. This is particularly true if one further allows for the possibility of allelic drop out (loss of one of the two alleles in a locus because of the damaged condition of the sample – allowing an experimentally homozygous 12 to match a 12/16 heterozygous reference). M-FISys helps a forensic scientist to resolve these ambiguous matches by a process we call \textit{iterative pruning}. The operator can confirm or exclude possible matches for any ambiguous sample, annotating reasons along the way. For instance, a right hand with degraded DNA can be excluded from a potential ID on the grounds that a full profile for a right hand has already been reported. As more information is accumulated, more matches can be excluded or confirmed. mtDNA or SNP data may
help to confirm or refute a potential match. Finally, experimental errors can also be corrected and annotated as part of an exquisitely meticulous Administrative Review and Quality Control process that is beyond the scope of this paper.

When direct matches to ante-mortem exemplars are not possible, or if the chain of custody for the exemplar cannot be reliably traced\(^1\), kinship analysis is the next source of evidence. Many ethical, legal and social issues are brought into sharp relief at this stage of the process. For instance, the State of New York temporarily suspended Informed Consent rules for the case of collecting buccal swabs from family members. This was not, as some mistakenly inferred, motivated by a need to collect family references more quickly. Rather it was a compassionate ruling allowing for samples to be taken (in many cases from young children) without forcing the contributor to engage in a discussion about the need for DNA samples. There are those who might suggest that the legal suspension of Informed Consent regulations gives tacit permission to participating scientists to perform population genetics research using the database of collected victim and kin samples. This would be a gross and egregious violation of the dignity, privacy and civil rights (under other laws in the State of New York) of the victims and their families. A less extreme issue is the decision on what to do if *false paternity* is ever detected. It seems only reasonable and compassionate to withhold any such findings permanently: What could be more devastating for a father than to learn that not only had a child been lost to a senseless act of terrorism, but that the child had never been his?

As part of the WTC recovery and in collaboration with Dr. Charles Brenner and Dr. Benoit LeClaire, kinship analysis can be tactically approached in two different ways. Using Brenner’s approach, family pedigrees are constructed based on the reported relationships to the victim and the set of victim profiles is scanned for promising candidates to fill in the missing (victim’s) position in the pedigree. The data can then be revised to allow for incorrectly reported relationships and matches can be further refined. Dr. LeClair takes the conservative approach of assuming that any reported relationship could be wrong (this turns out to be true with unfortunate frequency due to errors in the original collection of data). Each kinship sample is compared to victim samples and tested for likelihood at various relationships such as parent-child, sibling or half-sibling. These are sorted by most-likely match. In an accurate match for a particular victim sample, the correct familial samples will float to the top of the list.

The mathematics of kinship analysis is well established.\(^2\)\(^,\)\(^2\)\(^,\)\(^4\) Scores are reported as a hypothesis-based “likelihood ratio,” calculated as the likelihood that the

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\(^1\) Verifying the origin of a reference sample has proven to be an enormous challenge in this process. Because of the chaos and emotional stress of the first weeks following 9-11, many contributed samples were accepted with incomplete or even incorrect paperwork. As a result, all identifications are based on no less than two match modalities.
relationship between the victim sample and the kin sample is as hypothesized (e.g., full sibling in the Hispanic population) divided by the likelihood of seeing the same number of matching alleles in two unrelated persons in the same population.

Similar to the Likelihood in direct STR matches, Kinship Likelihood is the product of Kinship Locus Likelihood across all existing alleles (partial profiles are possible here as well). What is different is that two profiles are being compared instead of just one. In the case of partial profiles, only those loci extant in both profiles can be compared. As in the case of direct STR Likelihood, Kinship Likelihoods were calculated across the four aforementioned races, the lowest value taken as "a worst case scenario".

The Kinship Locus Likelihood takes as input a relationship between the two profiles, which is one of: Parent/Child, Full Sibling, Half Sibling. Formulae exist for First Cousin relationships as well, but it was felt to be less useful for the World Trade Center project, so was never used (swabs from cousins were not taken); however, because it was coded into M-FISys, the formula will be shown here as well. Let \( p \) be the frequency probability of the high allele value, \( q \) the frequency probability of the low allele value. Then the Kinship Locus Likelihood is defined the following way [13]:

\[
\frac{(a_2p_2 + a_1p_1 + a_0p_0)}{p_2}
\]

where \( a_i \)'s are proportions based on relationship:

- Full Sibling: \( a_2 = 1/4, a_1 = 1/2, a_0 = 1/4 \)
- Half Sibling: \( a_2 = 1/2, a_1 = 1/2, a_0 = 0 \)
- Parent-Child: \( a_2 = 0, a_1 = 1, a_0 = 0 \)
- First Cousin: \( a_2 = 3/4, a_1 = 1/4, a_0 = 0 \)

and the \( p_i \)'s are defined as follows:

\[
p_2 = p^2 \quad \text{if the victim is homozygous and matches one of the relative's alleles}
\]

\[= 2pq \quad \text{otherwise} \]

\[
p_1 = 0 \quad \text{if relative and victim alleles share no common value}
\]

\[= p \quad \text{if the relative is homozygous and the victim shares its (homozygous or heterozygous) low value} \]

\[= q \quad \text{if the relative is homozygous and the victim shares its (heterozygous) high value} \]

\[= p/2 \quad \text{if the relative is heterozygous and shares exactly one value with the victim's (homozygous or heterozygous) low allele} \]

\[= q/2 \quad \text{if the relative is heterozygous and shares exactly one value with the victim's heterozygous high allele} \]

\[= (p+q)/2 \quad \text{if relative and victim are identical and heterozygous} \]

\[
p_0 = 1 \quad \text{if the relative and victim alleles are identical}
\]

\[= 0 \quad \text{otherwise} \]
The unfortunate truth is that some samples are so severely burnt that full STR profiles will not be available for either direct or kinship analysis. When samples are badly compromised through harsh environmental degradation, many of the loci may be blank and the likelihood of observation can drop to levels where one would expect many instances of shared “partial profiles” in a population of 2,801 victims. At this point, the options for making a positive ID are a) to re-extract and retest DNA with hopes of collecting information at additional loci, or b) to use other techniques that might be more effective on highly degraded samples, even if they are not as discriminating. One of these alternative techniques is mitochondrial DNA analysis.

Mitochondrial typing involves direct sequencing of the highly variable regions of the genome adjacent to the origin of replication. The standard in forensic communities is to report results not in the form of the sequence itself, but in a compact format that only shows the difference between the experimental results and an established and internationally recognized standard known as The Anderson Sequence. If the sequence being typed is identical to the Anderson reference, the difference report will be null. Point mutations are described as a base position, plus the base that differs from the reference sequence. Deletions are represented as a “D” character (not to be confused with the IUB code for “A, G or T, but not C”). We call this difference list the “delta representation.” A typical mitotype might look like this:

```
16093:  C
16224:  D
16311:  C
395:    C
263:    G
315.1:  C
```

Note that the base positions cross the origin of the 16569-base genome. The first lines indicate that this sample has a C at position 16093 where a T is usually found, and a deletion of the base at position 16224. To maintain the integrity of numbering, an insert is indicated as a decimal point position on the base that the insert follows. In this case, “315.1: C” indicated that there is a C insert after position 315 in the reference sequence.

Although standards have been promulgated, there is still the occasional disagreement over nomenclature, particularly as it applies to inserts. For instance, if the reference sequence includes the sequence “TTT” starting at position 16091, and the sample under study has four T’s rather than three, it can logically be represented as any of the following:

```
16093:  C
16224:  D
16311:  C
395:    C
263:    G
315:    C
```

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8 This reference is actually a consensus made up of several sequences and is not necessarily the most common. An alternative reference sequence called The Modified Cambridge Reference Sequence is gaining in popularity. The Anderson Reference is used in the WTC analysis.
The sequence used in mitotyping is not in a coding region and, since the inserted T has no biological significance, it can be referenced as an insert in any of several places. Two of the collaborating labs were not in precise agreement on the standard notation for this insert, even though both were producing accurate mtDNA profiles. The overtly political positioning (almost exclusively from outside advisors not working at either lab or with the software development team) over the “correctness” of one naming representation or another was an example of one of the greatest non-technical challenges of this project. We defined this to be a non-issue for the current project, since our solution was to reverse-translate the difference list to generate a non-ambiguous original sequence and compare those sequences themselves.

mtDNA is very hardy material that can survive intact under conditions where nuclear DNA degrades. However, the variation in the hypervariable d-loop of the mitochondrial genome is not nearly as discriminating as a 13 locus STR profile. In fact, over 5% of the Caucasian population share the same, common mitotype. Mitotyping is most often used to exclude a match but by itself would not normally contain enough information to confirm an identification with a high degree of confidence. Fortunately, we can take advantage of the fact that mitotypes are independent of the STR profiles and use them as additional information to supplement a partial STR match.

Mitochondrial DNA is inherited only through the maternal line. Therefore, many of the samples available to the OCME are useless for mitotype matching. Fathers will not contribute mtDNA to their offspring, and children of male victims (roughly 2/3 of the victims were male) do not have relevant mtDNA for identification. However, as an Operations Research issue, it was found to be more efficient and less expensive to mitotype all kin swabs than to pull out and re-array only those samples that were part of a matrilinear line to a victim.

A proposal was made that mitotype frequency statistics be generated from roughly 5,200 personal effect and kinship samples available. With just a quick review of the database as of September 16, 2002, it was found that at least 24% of the 2,801 victims had no maternally-related kin samples available to the process. This, plus the enormous amount of effort that would be required to confirm the validity of all kin swabs and personal effects, argues that a frequency database from

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b There was a concern that the differences in nomenclature could obscure matches between reported mitotypes. Still, it was sociologically fascinating that a complete and provable alternative that avoided that problem did not allow us to avoid a significant expenditure of resources in the debate over who was more “correct.” Several people have observed that this has been an issue primarily for outside advisors who, like all of us, wanted desperately to be of material help in the face of such a tragedy.
the general population might be preferable to one that could be reasonably and reliably created specifically for WTC victim population.

At the time of this writing SNP data has not been applied to the identification process. The ability to collect data from very short sequences makes this an exciting technology which offers great hope for collecting identifiable information from badly degraded samples. SNPs occur within the human genome on average every 100 to 300 bases and are stable from an evolutionary standpoint, making them easier to follow in populations as they do not change much from generation to generation. A panel of 70 SNPs has been characterized by the GeneScreen division of Orchid Biosciences, all from nuclear DNA. Roughly 2 out of every 3 SNPs involves replacing a C with a T, and it is among these biallelic loci that the 70 forensic SNPs are chosen, specifically ones in which C and T are equally likely.

For kinship analysis, it is important to ensure that these loci are genetically independent of each other and not linked to nuclear STR loci used in identifying the same sample. In an ideal world, all loci would be no less than 50MB or 50cM apart. With 70 SNPs plus up to 15 STR loci in a genome that is only 3.5 billion bases long, this is clearly not possible.

In a study by Dr. Ranajit Chakraborty of the Center for Genome Information, the estimated allele frequencies for the current panel ranged between 15.5% to 81.3%, although the great majority were nearly equal (an average heterozygosity of 46% across the three population groups Caucasian, Black and Hispanic). Fortunately, his study showed the allelic dependence being relatively small, 5.71%, comparable to the 5% expected by chance alone. He further concluded that despite the lack of theoretically independent loci, an analysis of 713 genotypes from 3 different populations support the efficacy of using this 70 SNP panel for identification purposes. With a complete profile of 70 SNPs at probability 1/2, the likelihood of match would be $2^{70}$, or approximately $10^{21}$. Even with a revised probability average of 46%, the likelihood is $10^{18}$.

The authors would reject the proposal that this panel of SNPs should be replaced with one that is more evenly spaced. Because of the limitation of the genome size, linkage in a 70 locus panel is a problem intractable in the laboratory but addressable in software. M-FISys can take an arbitrarily large panel of SNP and STR loci and, given a threshold likelihood value, select those loci that create an unlinked set of STRs and SNPs to reach that threshold in a valid match (assuming the data is available). The remaining loci, of course, would still be reviewed and any directly conflicting data highlighted to indicate experimental error or other anomalies that should be resolved by a forensic biologist before a positive ID is reported.

Combining STR, mtDNA and SNP data is as much a human-computer interaction issue as it is a computational one. Keep in mind that the goal is not for the software to make identifications; as a matter of law, neither the developers nor the software have such authority. Rather it is to present data to a qualified and
authorized forensic biologist to make it easier for that person to certify an identification. Instead of developing an interface that combined all STR, mtDNA and SNP data in a single view, M-FISys is broken up into STR-\textit{centric}, mito-\textit{centric} and SNP-\textit{centric} views of the data, with indications that other data supports or contradicts a proposed identification.

Fig. 2. STR-centric view of the Master List displays details of STR results as well as indicators of supporting or conflicting experimental evidence using mitotyping or SNP analysis.

Figure 2 shows the STR-centric view of the master comparison list. The first column shows the names of samples including personal effects (starting with “SP” because they were typed at the State Police lab in Albany) and disaster samples that, following an optional prefix, are coded as DM01nnnnn, where the final 5 digits are the order in which the remains were discovered and checked in at the morgue.\textsuperscript{1} The fifth column contains three values. The number at the top is the number of loci (including the gender locus) that yielded results. There is a character on the lower left and the lower right of that column that indicates consistency between the STR match and mito data, and SNP data, respectively. The possible values for mito and SNP data are a green check mark (√: the data is consistent with the STR results), a red X (X: the data is inconsistent or contradictory), a question mark (?: the data is partially or ambiguously consistent) or a dash (—: this analysis has not yet been performed). Following this is the complete STR profile of the sample.

Similarly, in the mito-centric view of the same data (Figure 3), the operator can see the complete mitotype, plus an indication of whether any STR or SNP data supports or contradicts a mito-based match. This approach allows us to add an arbitrarily large number of different modalities for identification. Individual forensic biologists may be charged with reviewing a particular type of data within their specialty, but are immediately alerted if contradictory data needs to be reviewed.

\textsuperscript{1} The header “DM01” stands for “Disaster: Manhattan, 2001.” There are also “DQ01” samples, referring to the “Disaster: Queens” when flight 587 lost part of its tail and crashed in November of the same year.
The last area of functionality we will review is quality analysis of incoming data. An enormous amount of energy at the OCME is expended to avoid false identifications. While every effort is made to make the process run as quickly and efficiently as possible, the idea of having to tell a family that they have buried the wrong remains is horrifying. Because remains were crushed, fragmented and co-mingled, it is important not only to make sure that the correct remains have been typed, but that no additional remains are being released when a funeral director accepts custody of human tissue from Memorial Park.

For that reason, the first person to review remains when they arrive at the morgue is not a Medical Examiner, but a Forensic Anthropologist. This person’s job is to be as certain as possible that a) the recovered remains are human, and b) that only one person is included in a collection. The policy is that if remains are not physically connected by tissue or sinew, that they should be separated into individual pieces and typed independently.
An example of the kind of QC that M-FISys performs is shown in Figure 4. Sample DM0101234 (not the real number) has been typed twice. In one case, the tissue was typed by Myriad Genetics. But bone from the same sample was sent to Bode Technology Group, a well-regarded forensics lab in Virginia with a highly developed capacity for extracting DNA from bone. If you look at the values for each locus, several of them (in yellow) are in conflict. To a forensic biologist, these profiles clearly represent two different people. In fact, they are different genders.

There are several possible resolutions to this conflict. It could be that, upon examining the shipping manifests, it is found that a sample has been mislabeled. This is almost never a consideration, but it still must be an option for resolving the problem within the program. Smaller discrepancies can be the result of experimental error or allelic drop out. In this case, a review of the remains showed that bone from one person was embedded in the muscle tissue of another. Certainly a bizarre circumstance in most medical examiner’s offices, but a sadly familiar possibility at the World Trade Center. The program will prevent these samples from being identified and released until this conflict of data for a single “victim sample” has been resolved.

Another of the several data-integrity tests in M-FISys compares the profiles of personal effects to the swabs taken from family members. Families of missing persons are able to call the “DNA Hotline” at the OCME and find out if enough DNA has been collected for an identification. For instance, a mother whose husband did not come home after the attacks might bring in the presumed victim’s toothbrush as well as the oldest child of the victim to give a buccal swab. However, she might ask that the OCME determine if there is enough DNA available from the personal effect to make an identification without additional kin samples because she does not want to further traumatize her younger children by asking them to provide swabs to help identify their father. A call to the DNA hotline might indicate that the toothbrush gave a full profile and that further samples are not required. But what if the wrong toothbrush was brought in? If the toothbrush matches the DNA for the wife or the oldest child, the full profile is not valid and a swab from additional children might be key to making a final identification.

Bone is difficult for several reasons. One issue is that the dust created when a sample is extracted must not contaminate the next sample from the same bench. More significantly, Calcium is a PCR inhibitor. Bode Technology Group has a core competency in overcoming these obstacles that has been invaluable in the WTC project.
4 Summary

When we were brought onto this project several weeks after the September 11 attacks, we developed five goals and two major deliverables. The goals were:

1. Identify individual remains,
2. Reunify partial remains so that they can be returned to families,
3. Collect and warehouse meta-data for administrative review of reference samples,
4. Track samples among collaborating labs, and
5. Create an information management system to report metrics and make problem resolution proposals to supervisors at the OCME.

Not all aspects of the engineering effort have been discussed in this paper, but we hope a broad overview of both the process and the challenges has been conveyed.

The first deliverable has been to create a mass-fatality identification and recovery system for the OCME, creating installed value to address specific data-handling problem on the fastest possible schedule without running the risk of false identifications. This has been a seven-day-a-week job for over eleven months as of this writing, and all of the participants in the project feel that it is the most important thing they will do in their professional lives. The second deliverable will be to create a generalized and portable version of this tool that can be deployed anywhere in the world where there is a massive human tragedy, be it natural or man made.

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