#### Scientific Approaches to Statistical Analysis and Collection of Handwriting Databases

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# Individuality

**Biometric Individuality** (of a <u>population</u> with respect to a <u>comparison methodology</u>): The probability that two (different) randomly selected writers from the population have indistinguishable writing profiles with respect to the comparison methodology being used.



- Fixed RNMP so we can study the RMP as a function of document size.
  - We will control the RNMP at 1% and model the RMP as a function of the size of writing samples selected from each writer's body of handwriting.

#### Match Probability

• Random Match Probability (RMP) is the chance of randomly selecting two subjects from the population and then randomly selecting a writing sample (of a given size) from each subject that is declared a match by the biometric.

 Random Non-Match Probability (RNMP) is the chance of randomly selecting a single subject and then sampling two documents from the selected subject's body of handwriting that are declared a non-match with respect to the biometric

### Pilot Study

- ~434 different writers
  - Approximately 10 samples (5 in print and 5 in cursive) of a modified "London Letter" paragraphs per writer
  - Collected from volunteers at the FBI, training classes, and at various conferences over a 2-year period.
  - Two of the five script paragraphs from each of 100 writers.
    - "FBI 100" data set

#### Data Processing

- Automated process represents each segment by a graphical isomorphism
  - Referred to as an isocode.
- Each document is reduced to the frequency of isocodes used to write each letter.
- Writing samples then consists of a set of isocode/letter pairs.
  - Each writing sample is represented as a cross-classified table of isocode by letter.

#### **Data Processing**



### Sub-sampling Algorithm: RNMP

RNMP sub-sampling algorithm :

- 1. Randomly select one writer.
- 2. For the selected writer, construct two "random" writing samples by selecting, without replacement, a pre-specified number of characters from that individual's collection of documents: *n*1 being the number of characters making up the writing sample from the first writing sample, and *n*2 being the number of characters making up the second writing sample.
- 3. Compare the two "random" writing samples using the chi-squared similarity score.

Application of the re-sampling algorithm many, many times over a variety of writing sample sizes results in a set of "data" of the form:

(n1, n2, x) where x = chi-squared similarity score.

#### Sub-Sampled Within-Writer Similarity Scores

#### (K=1000 for each document size)



Score



Score

### Sub-sampling Algorithm: RMP

RMP sub-sampling algorithm :

- 1. Randomly select two writers without replacement.
- 2. For each selected writer, construct a "random" writing sample by selecting, without replacement, a pre-specified number of characters from that individual's collection of documents: *n*1 being the number of characters making up the writing sample from the first selected writer, and *n*2 being the number of characters making up the writing sample from the second selected writer.
- 3. Compare the two "random" writing samples using the chi-squared similarity score and record whether or not a match has occurred.

Application of the re-sampling algorithm many, many times over a variety of writing sample sizes results in a set of "data" of the form:

(n1, n2, x) where x = 1 if the two writing samples match; 0 if the two writing samples do not match.

### **RMP** Modeling



	Coefficient	Standard Error
(Intercept)	-2.28675	0.07749
<i>n</i> 1	0.01075	0.00059
<i>n</i> 2	0.00998	0.00059

#### **RMP Modeling: Equal Document Sizes**



# The Modeled Variance



**Document Size** 

#### Properties of the Estimators

- 1. Consistent as the number of Writing Samples and the number of simulated documents tends to infinity.
  - The Writing Sample Size can remain fixed!!
- 2. Asymptotically Normal Estimators.
- 3. Unbiased for the *RMP* and  $E(p_i^2)$ .

### The Design of an Individuality Study

- The sub-sampling models provide guidance on the relationship between the size of a writing sample collected and the RMP.
- Basic probability inequalities can give an idea on the behavior of upper confidence bounds on the RMP.
  - Given combination of writing sample size and number of sampled writers.
- The ideal setting is when we have a sample of documents from a large number of people and observe no matches when the collected documents are combined.

#### **Confidence Bounds**



95% Upper Confidence Bounds for the RMP when no observed matches are observed with *n* writers.

# Writing-Sample Sizes Needed for Specified Number of Writers

- Based on a one sided version of Chebyshev's inequality.
  - Sometimes called Cantilli's inequality.
  - The probability of observing no matches when comparing writing samples pairwise from each of *n* writers is at least 50%, 80%, and 95% for the following writing sample sizes

	Probability of No Matches		
Number of Writers (N)	50%	80%	95%
50	751	828	916
100	869	945	1032
200	985	1062	1147
500	1137	1213	1298
700	1193	1268	1353
1000	1251	1326	1411
2000	1364	1439	1523

#### The Determination of Writing Sample Size

The chance of observing a match in the n(n-1)/2 pairwise comparisons is a function of the writing sample size.

For example, say the desired upper confidence bound on the RMP is 1 in 100,000.

- 1. Then the smallest number of writers we could use to achieve this bound is 700.
- 2. To have at least an 80% chance of achieving no matches in the 244650 cross- comparisons:
- 3. We would need to have each person submit a writing sample of about 1268 characters

# An Example



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