

DNA Identification as an Information Science

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Cybergenetics

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The DNA Crime Solution



Evidence

- solve a crime
- match DNA evidence to suspect

Investigation

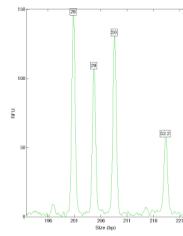
- DNA database of crimes and criminals
- match evidence to convicted offenders

Prevention

- all evidence: property crime, sexual assault, ...
- all criminals: felonies, misdemeanors, ...
- DNA databases interrupt criminal careers

The DNA Evidence Problem

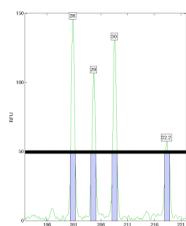
Most DNA evidence is uncertain,
producing uncertain genotypes



- mixtures: two or more contributors
- degraded or damaged DNA
- low template DNA amounts

An Interim Approach

Apply "thresholds" to the quantitative data



- peak height: variation vs. absolute
- variance scaling: linear vs. constant
- data: distribution vs. false negative
- probability: concentrate vs. disperse
- information: preserve vs. discard
- identify: succeed vs. fail
- protect: better vs. worse

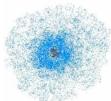
Probabilistic Science

Classical Physics



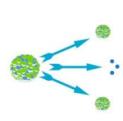
interim solution (1913):
electron in an orbital

Quantum Mechanics



true solution (1925):
electron with a probability

Validation: The Atom Bomb



Splitting the atom



War in Pacific is over

Probabilistic Genotypes

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Information Theory

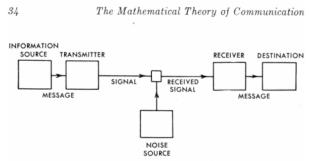
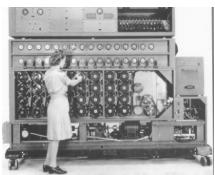


Fig. 1.—Schematic diagram of a general communication system.

Quantify the information in a system

Validation: The Bombe Computer



Breaking the code



War in Europe is over

DNA Likelihood Ratio

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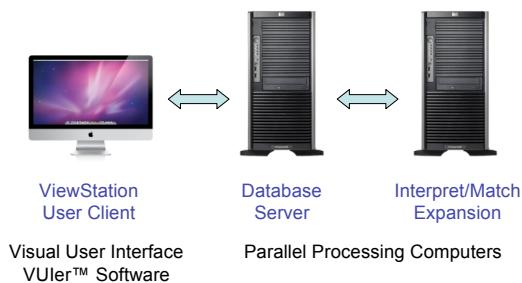
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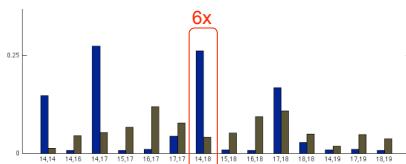
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TrueAllele® Casework System



DNA Identification

- 1 infer genotype
(up to probability)
- 2 match genotype
(compute LR)



Validate Reliability

- log(LR) is a standard measure of information
- translates genotype match into a single number

