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## “Real” Caller ID Project Update and Demonstration

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## Project Review

- We would like a means of identifying a caller passively (through normal use).
- Don't want to force users to authenticate in order to place or accept calls
  - Repeated auth is too cumbersome
  - How frequently is auth done?
- Alternative: use what we can observe of calls passively to determine caller identity

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## Passive Caller ID

- Leverage biometrics (voice) and other context to passively establish caller identity
- Each party still authenticates the other by
  - sound of voice
  - common knowledge
- Passive caller ID is not a substitute for user authentication! (MITM threats exist)

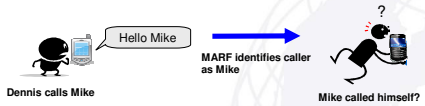
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## Jan 12 2011 Meeting

- Showed speaker recognition using MARF
- Mixed results using speech alone
- Showed need for exploiting caller context



- Focus since has been exploiting context

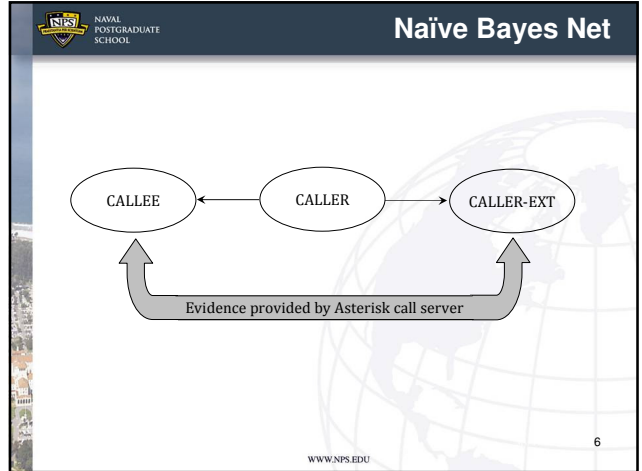
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## Approach

- Build a Bayesian network for caller ID
- Treat MARF output as one piece of evidence
- Consider more evidence:
  - CALLEE (which number was dialed?)
  - CALLER-EXT (which ext placed the call?)
  - CALL-TIME (when was call placed?)
  - EXT-LOC (where was CALLER-EXT located?)

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## Example

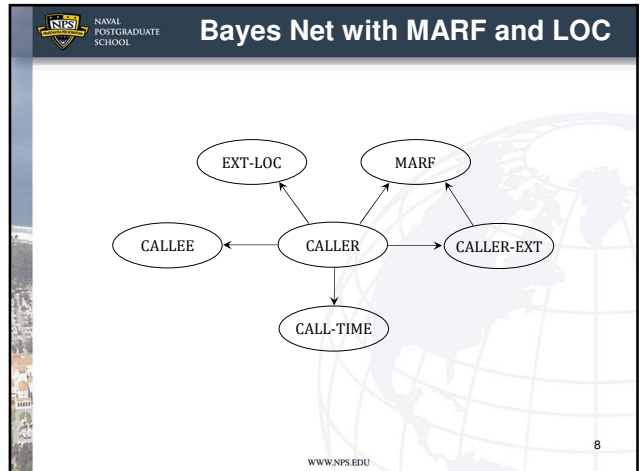
$CALLER = CALLEE = \{2002, 2003, 2004\}$   
 $CALLER-EXT = \{1002, 1003, 1004\}$

$$P(CALLER = v \mid CALLEE = c, CALLER-EXT = e) = \frac{P(CALLER = v, CALLEE = c, CALLER-EXT = e)}{P(CALLER = c, CALLER-EXT = e)}$$

$$= \frac{P(CALLER = c \mid CALLER = v)P(CALLER-EXT = e \mid CALLER = v)P(CALLER = v)}{\sum_{CALLER=2002, \dots, 2004} P(CALLER = c, CALLER, CALLER-EXT = e)}$$

$$= \frac{P(CALLER = c \mid CALLER = v)P(CALLER-EXT = e \mid CALLER = v)P(CALLER = v)}{\sum_{k=2002, \dots, 2004} P(CALLER = c \mid CALLER = k)P(CALLER-EXT = e \mid CALLER = k)P(CALLER = k)}$$

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## MARF Conditional Probabilities

MARF

| Caller Ext | 1003 |      |      |
|------------|------|------|------|
| Caller     | 2003 | 2004 | 2002 |
| 2002       | 0.18 | 0.14 | 0.87 |
| 2003       | 0.6  | 0.17 | 0.08 |
| 2004       | 0.22 | 0.69 | 0.05 |

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9

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## Demo Conditional Probabilities

Call\_Time

| Caller | 2003 | 2004 | 2002 |
|--------|------|------|------|
| AM     | 0.5  | 0.4  | 0.2  |
| PM     | 0.5  | 0.6  | 0.8  |

Callee

| Caller | 2003  | 2004 | 2002 |
|--------|-------|------|------|
| 2003   | 0.01  | 0.89 | 0.89 |
| 2004   | 0.495 | 0.01 | 0.1  |
| 2002   | 0.495 | 0.1  | 0.01 |

Caller\_Ext

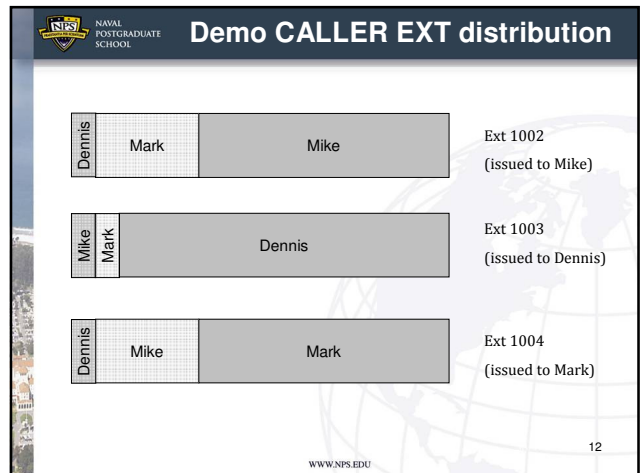
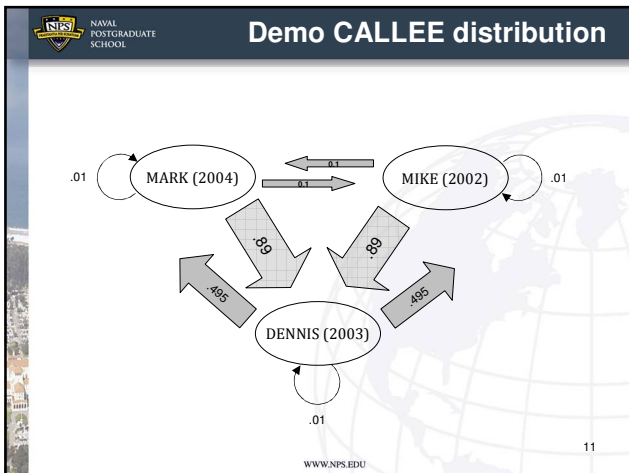
| Caller | 2003 | 2004 | 2002 |
|--------|------|------|------|
| 1003   | 0.9  | 0.05 | 0.05 |
| 1004   | 0.05 | 0.8  | 0.15 |
| 1002   | 0.05 | 0.15 | 0.8  |

Caller

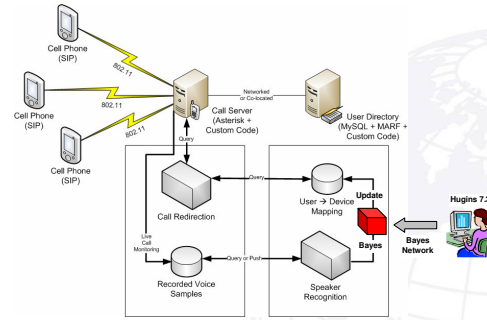
|      |          |
|------|----------|
| 2003 | 0.333333 |
| 2004 | 0.333333 |
| 2002 | 0.333333 |

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10



- Updated call server with Bayes net
- Bayes net built using Hugin 7.3
- Hugin produces executable Bayes net
- Bayes net and MARF run after every call
- Before only MARF ran
- MARF output is evidence into Bayes net



Contact Information

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