Human Assisted Speaker Recognition

CRAIG S. GREENBERG, ALVIN F. MARTIN, MARK A. PRZYBOCKI

#### NATIONAL INSTITUTE OF STANDARDS AND TECHNOLOGY, INFORMATION TECHNOLOGY LABORATORY, INFORMATION ACCESS DIVISION

#### NIST Speaker Recognition Evaluations (SRE)

NIST SRE measures speaker detection performance of stateof-the-art research systems on common test data

- Since 1996: sponsored by DoD, managed by NIST
- Open to participants worldwide
- Machine only: no listening or other human interaction allowed
- Recorded samples compared may differ in channel and style, as in forensic/biometric apps:

• Interviews and telephone conversations, many microphones

#### The Speaker Detection Task

Given pairs of speech recordings:

- A "training" recording of 10sec, 5min, 8 min...
- A "test" recording of any such length
  Telephone or microphone, conversation or interview
- Prior probability, and cost of miss and false alarm System response, for each pair:
- Same voice: Y/N?
- How likely? (log likelihood)

SRE 10 Evaluation Test Conditions											
		<b>Test Conditions</b>	est Conditions								
		10sec	5min (tel/mic)	summed channels							
IS	10sec	optional	-	-							
ndition	5min (tel/mic)	optional	required	optional							
g Co	8conv	optional	optional	optional							
Trainin	8conv summed channels	-	optional	optional							

Number of trials: 31,387 - 610,748 per test condition Number of speakers: 596 Data from the Linguistic Data Consortium (LDC)

#### **Performance Metrics**

Detection (not identification)

- False reject (miss): incorrectly reject a speaker
- False accept (false alarm): incorrectly accept a speaker
- Tradeoff made by decision threshold

• Measures:

Equal-error-rate (EER)

× DCF

- o DET Curve w/ all tradeoff points
- Example Figures of Merit:
  - × %EER (easy to explain)
  - × %FR @ .01%FA (forensic, military)
  - × %FA @ 10%FR (access control)



### Why evaluate? SRE Performance History on Similar Tasks

#### 0.08 个个个 New Metric Ahumada (Spanish) Landline 0.07 Interview train: (40 target Multimodal (FBI) + Cell/Land 2 speaker detection land/cell test speaker Landline (summed channel) paradigm) different mic test 0.06 Landline 2 2-speaker same mic test min train 30 Cellular 2 min (summed Actual DCF sec test train 30 sec test channel) 0.05 Cell/Land Eng. 5min train Cross-mic 0.04 and test 1conv train(tel) 1 conv test (mic) 0.03 Cross Language 0.02 Cell/Land 8conv train English Only 0.01 1 conv test 0 1996 1997 1999 2000 2001 2002 2003 2004 2005 2006 2008 2010 1998 Year

**History of Performance** 

Original Chart provided by Douglas Reynolds of MIT-Lincoln Laboratory

# Wow, that's great! Do humans even matter any more?

# <u>**ALL</u> Speaker Recognition Applications Involve Humans!**</u>

- Forensic
- o Biometric
- Watchlist

**)** ...



How can human experts effectively utilize speaker recognition technology?

 HASR (*Human Assisted Speaker Recognition*) began addressing this question – a 2010 pilot test

## The HASR Task:

Given two different speech segments, determine whether they are both spoken by the same speaker

• HASR included two tests:

HASR1HASR215 trials150 trials

• HASR systems may use human listeners, machines, or both

• Participation open to all who might be interested



#### Sample Trials





#### HASR1 Results Summary

													_						
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	Misses	FAs	Total	
System 1	t	f	f	f	f	f	t	f	f	t	f	f	f	t	f	2	-	2	Correct Accept
System 2	t	t	f	f	t	f	t	t	f	t	f	f	t	f	t	1	3	4	concernecept
System 3	t	t	f	f	t	t	f	f	f	t	t	f	f	t	f	2	3	5	
System 4	t	t	f	f	t	t	f	f	f	t	t	f	f	t	t	1	3	4	Correct Reject
System 5	t	t	f	f	t	f	t	t	f	t	f	f	t	f	t	1	3	4	5
System 6	t	f	t	t	f	t	f	f	t	f	t	f	f	t	f	4	5	9	
System 7	f	t	f	t	f	f	f	t	f	f	f	f	f	t	f	5	3	8	Misses
System 8	f	t	t	t	f	t	f	t	t	t	t	f	f	t	f	4	7	11	
System 9	t	t	f	t	t	f	f	f	t	t	t	t	t	t	f	2	6	8	
System 10	t	t	f	t	t	f	f	f	t	t	t	t	t	t	f	2	6	8	False Alarms
System 11	t	t	t	t	t	t	t	t	t	t	t	t	t	t	t	-	9	9	
System 12	f	f	t	f	t	t	t	t	t	t	t	t	f	t	t	1	6	7	
System 13	f	t	t	f	t	t	t	f	t	t	t	t	t	t	f	2	7	9	
System 14	f	t	t	f	t	t	t	f	t	t	t	t	t	t	f	2	7	9	
System 15	t	f	f	f	f	f	t	f	f	t	t	f	f	t	f	2	1	3	
System 16	f	t	f	f	f	f	t	f	f	t	t	f	f	t	f	3	2	5	
System 17	t	t	t	t	f	t	f	f	f	t	t	f	f	t	f	3	5	8	
System 18	t	t	t	t	t	t	f	f	t	t	t	t	t	f	t	2	8	10	
System 19	f	f	f	f	t	f	f	t	f	t	t	f	f	t	t	2	2	4	
System 20	f	f	f	f	f	t	f	f	f	t	f	f	f	f	f	5	1	6	
KEY	Т	F	F	F	Т	F	Т	F	F	Т	F	F	F	Т	Т	-	-	-	
Number of Errors	8	14	8	8	8	11	11	7	9	2	15	7	8	4	<b>13</b>	<b>46</b>	87	133	







#### HASR2 and <u>Leading</u> SRE10 Automatic Systems



135 HASR2 trials

Six HASR systems (thin lines)

Six Automatic systems (thick lines)

11/30/2012

#### HASR2 and <u>Corresponding</u> SRE10 Automatic Systems



#### 135 HASR2 trials

Five HASR systems (thin lines)

Five Corresponding Automatic systems (thick lines)

11/30/2012

#### Conclusions

- Humans are part of all speaker recognition applications
  Understanding their capabilities and limitations is important
- Strong machine performance does not imply ready for deployment in any particular application
- The assumption that humans are superior to machines at speaker id needs to be qualified
- Spun off a whole line of research within the community
- More experiments planned