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Rosetta: An Analyst's Co-Pilot

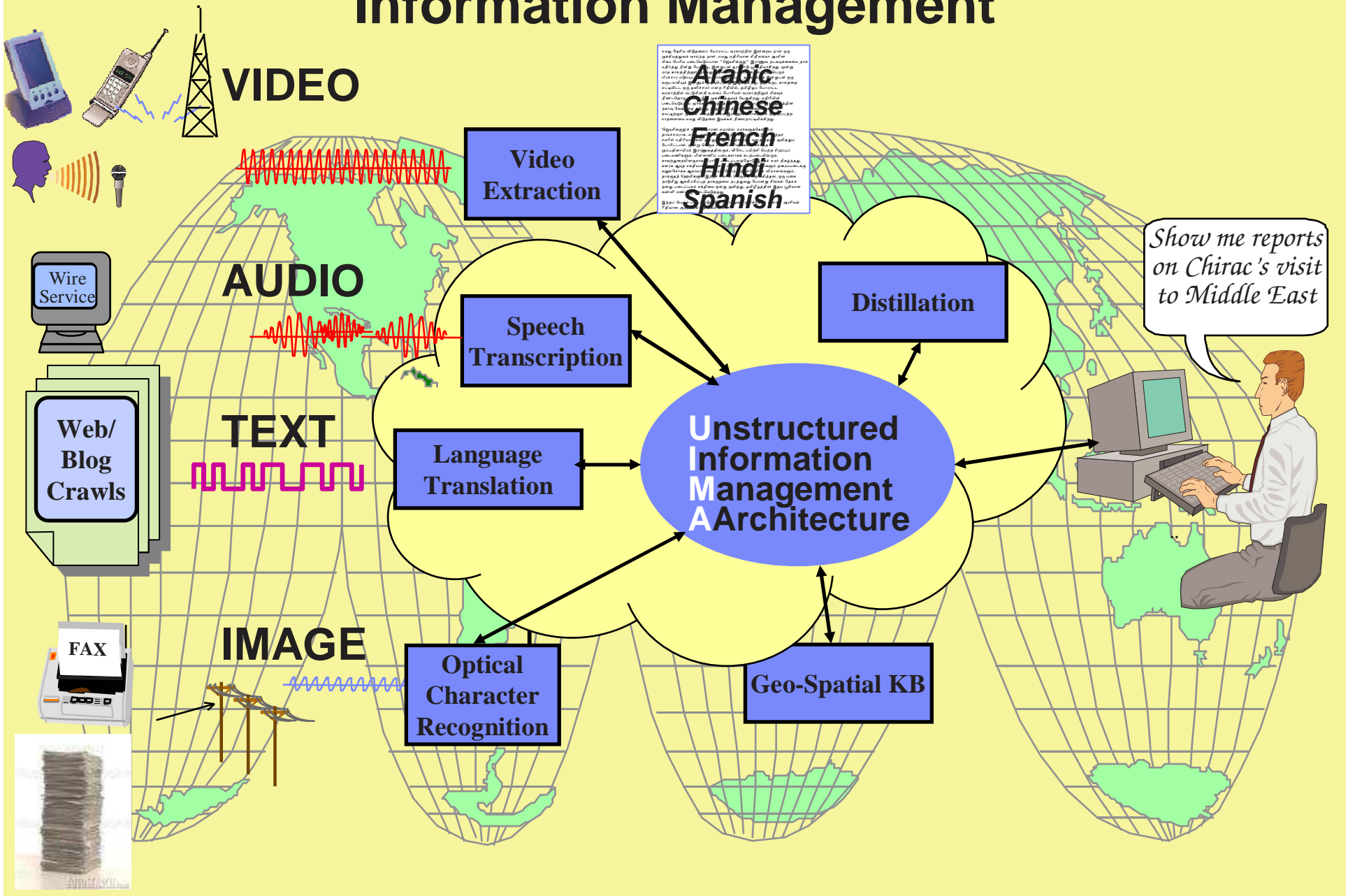
Salim Roukos
IBM TJ Watson Research Center



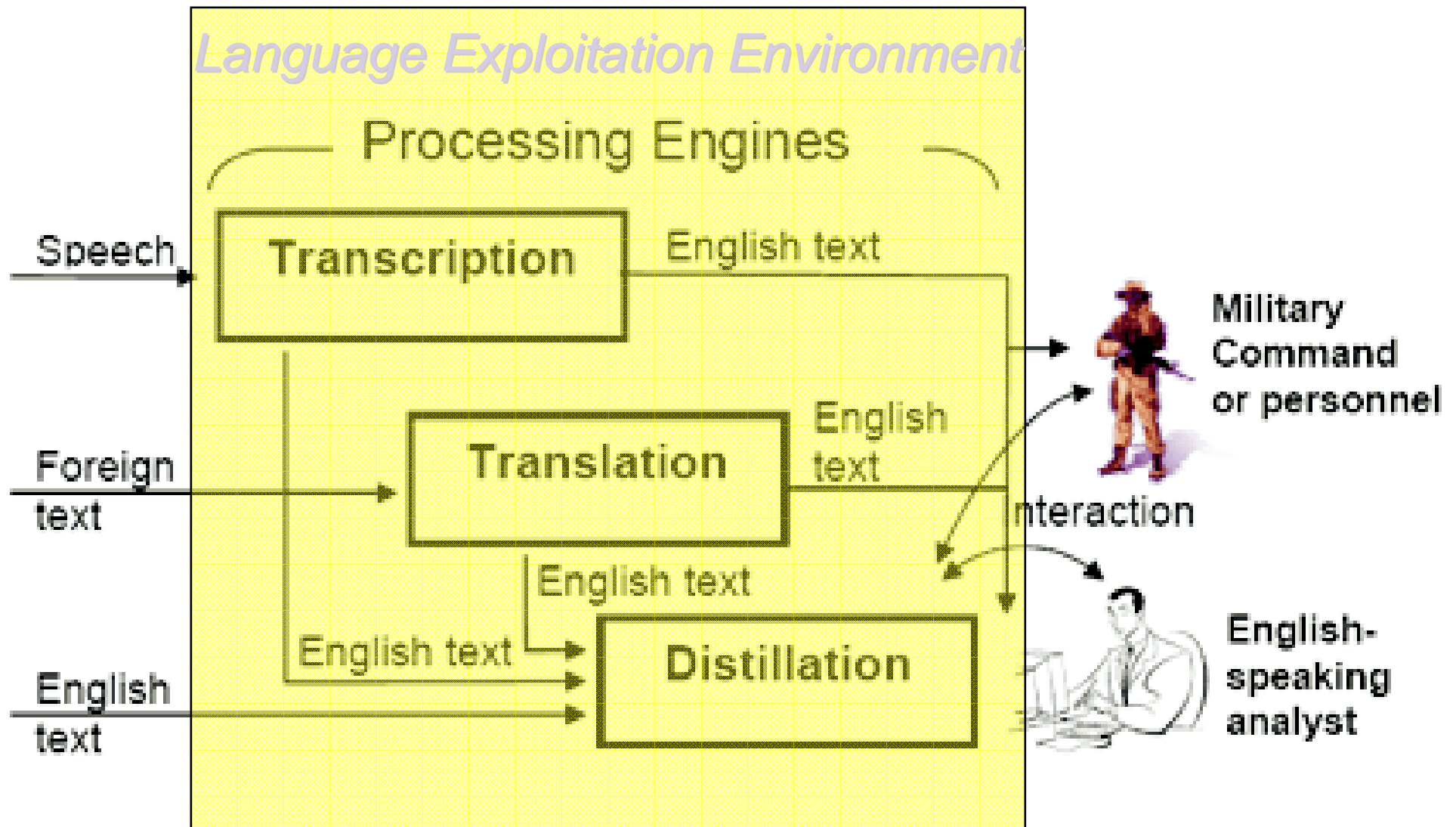
OUTLINE

- **Overview of GALE tasks**
- **Analysis of HTER GALE results**
- **Speech-To-Text overview**
- **Direct Translation Model II**
- **UIMA: Interoperability**
- **TALES demo**

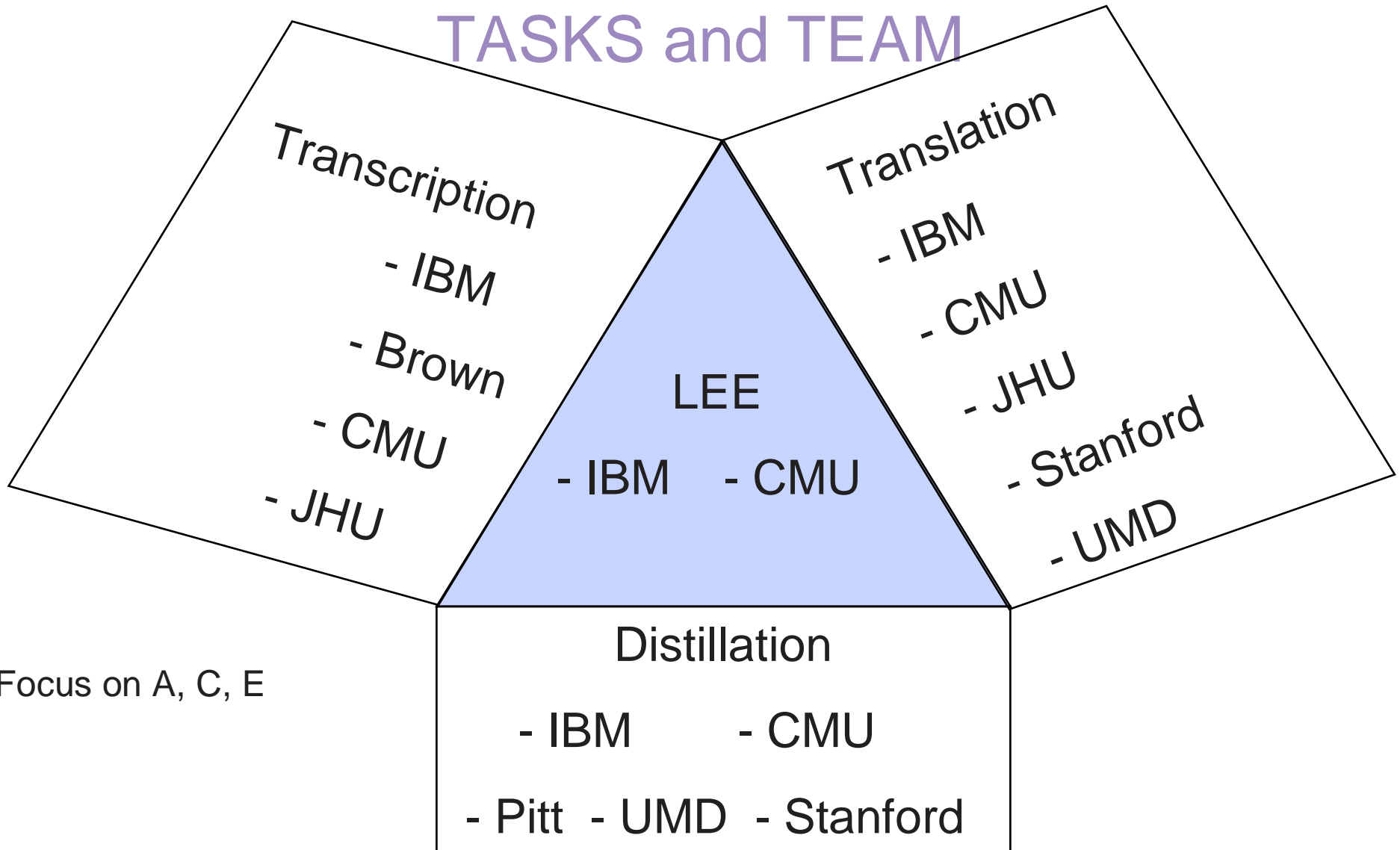
Multi-Lingual & Multi-Modal Information Management



GALE



ROSETTA TASKS and TEAM



Goals for ROSETTA System

- **Ingest traditional and informal media:**
 - broadcast news, talk shows, ...
 - Newswire, news web sites, blogs, ...
- **Scale to large volumes of multimodal/multilingual inputs**
 - Accurate, robust, quickly deployable engines, near real-time (up to 3x), 24x7, ...
- **Start w/Arabic, Chinese, English; scalable to 10's of languages**
- **Adaptive to user needs -- Personalized digests**
 - Robust, explainable, and controllable models of user and task
 - Automatic generation of focused reports & graphics, ...
- **End2End system as living laboratory**
 - Continuous testing

ROSETTA TASKS: LEE

- **Accelerate research & speedup insertion**

UIMA

- Common Annotation Structure (CAS) as input/output of multimodal processing engines/annotators/components
- Plug&Play: composition/integration of UIMAfied components
- Local/remote components with different OS's
- Open source

Rosetta will create:

- Common Type System
- Common Repository for componentry
- **MEMT: combine multiple MT engines**

ROSETTA TASKS (continued):

- Transcription
 - **Tightly integrated translation: small marginal error rate by combining speech-to-text and translation**
 - **3xRT or less runtime: fast, reliable, deployable system using common structure across languages and genres**
- Translation
 - **Preserving meaning: who did what to whom**
 - **Confidence measures: reducing human correction/editing**
- Distillation
 - **End2End system: task based eval. of improved components**
 - **Entity/relations networks, adaptive tracking, focused summarization, user modeling**

GNG (To Go or Not To Go:-) Evaluation

■ Transcription and Translation (HTER)

- Human post edits system output
 - Editor makes “minimum edits” of system output to reproduce correct meaning
 - HTER: Human Translation Error Rate
 - Control for human instruction across conditions/years – re-use fixed set of error full translations
- YEAR1: GNG edit distance
 - Transcription: 65% accuracy
 - Translation: 75% accuracy
- YEAR5: Both at 95%

DISTILLATION Evaluations

- **GO/NOGO**

- Compare automatic system output to human
- YEAR1: machine 50% of human using chosen metric

- **UTILITY**

- Compare human output in a task using either baseline or GALE system
- Open spec -- showcase technology

DISTILLATION GNG: Sample NL Question Schemata I

Two types of questions: **OPEN** and **SPECIFIC**

OPEN:

- **LIST FACTS ABOUT EVENTS DESCRIBED AS FOLLOWS: z**
- **WHAT [people/org/countries] ARE RELATED TO y:event AND HOW?**
- **PRODUCE A BIOGRAPHY OF [person]**
- **PROVIDE INFORMATION ON [organization]**
- **FIND STATEMENTS MADE BY OR ATTRIBUTED TO [person] ON [topic(s)]**
- **DESCRIBE THE RELATIONSHIP OF [person/org] TO [person/org]**
- **DESCRIBE [topic(s)] AND INVOLVEMENT OF [country]**
- **DESCRIBE THE PROSECUTION OF [person] FOR [crime]**
- **HOW DID x:country REACT TO y:event?**
- **WHAT CONNECTIONS ARE THERE BETWEEN [event 1/topic 1] and [event 2/topic 2]?**

DISTILLATION GNG: Sample NL Question Schemata II

SPECIFIC:

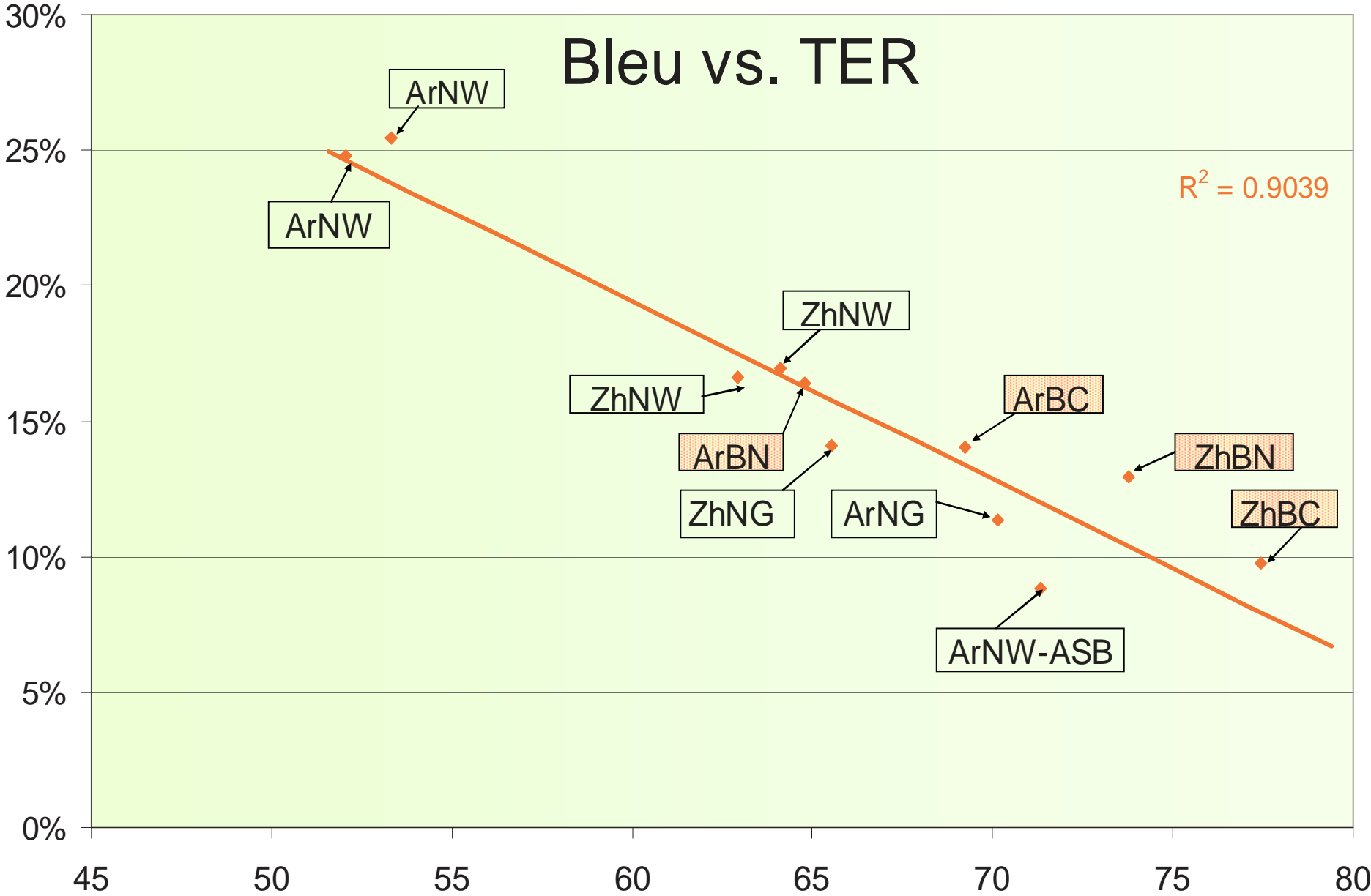
- **FIND MUTUAL ACQUAINTANCES OF [person] AND [person]**
- **TELL ME ABOUT [person's] MEETINGS ON [topic]**
- **FIND PASSAGES ABOUT [attacks] BY/OR ATTRIBUTED TO [group]**
- **FIND PASSAGES ABOUT [attacks] {IN [location] DURING [time interval]}**
- **DESCRIBE OUTBREAKS OF [disease] (IN [region] IN [time period])**
- **IDENTIFY PERSONS ASSOCIATED WITH [organization] WHO HAVE BEEN INDICTED ALONG WITH HOW THEY'RE RELATED**
- **IDENTIFY PERSONS ARRESTED FROM [organization] AND GIVE THEIR NAME AND ROLE IN ORGANIZATION AND TIME AND LOCATION OF ARREST**
- **DESCRIBE ATTACKS in [location] DURING THE PAST [duration] GIVING LOCATION (AS SPECIFIC AS POSSIBLE), DATE, AND NUMBER OF DEAD AND INJURED**
- **WHERE HAS [person] BEEN AND WHEN?**

GALE Transcription & Translation GNG Evaluation

- **Arabic and Chinese**
 - Speech
 - Broadcast News (BN) 10kw
 - Broadcast Conversation (BC) 10kw
 - Text
 - Newswire (NW) 10kw
 - NewsGroup/WebLog (WL) 10kw
- **1 Gold Reference with some word/phrase alternations**
- **3 Consortia participated in GALE06 Eval**
 - Agile (BBN)
 - Nightingale (SRI)
 - Rosetta (IBM)

Bleu vs. TER

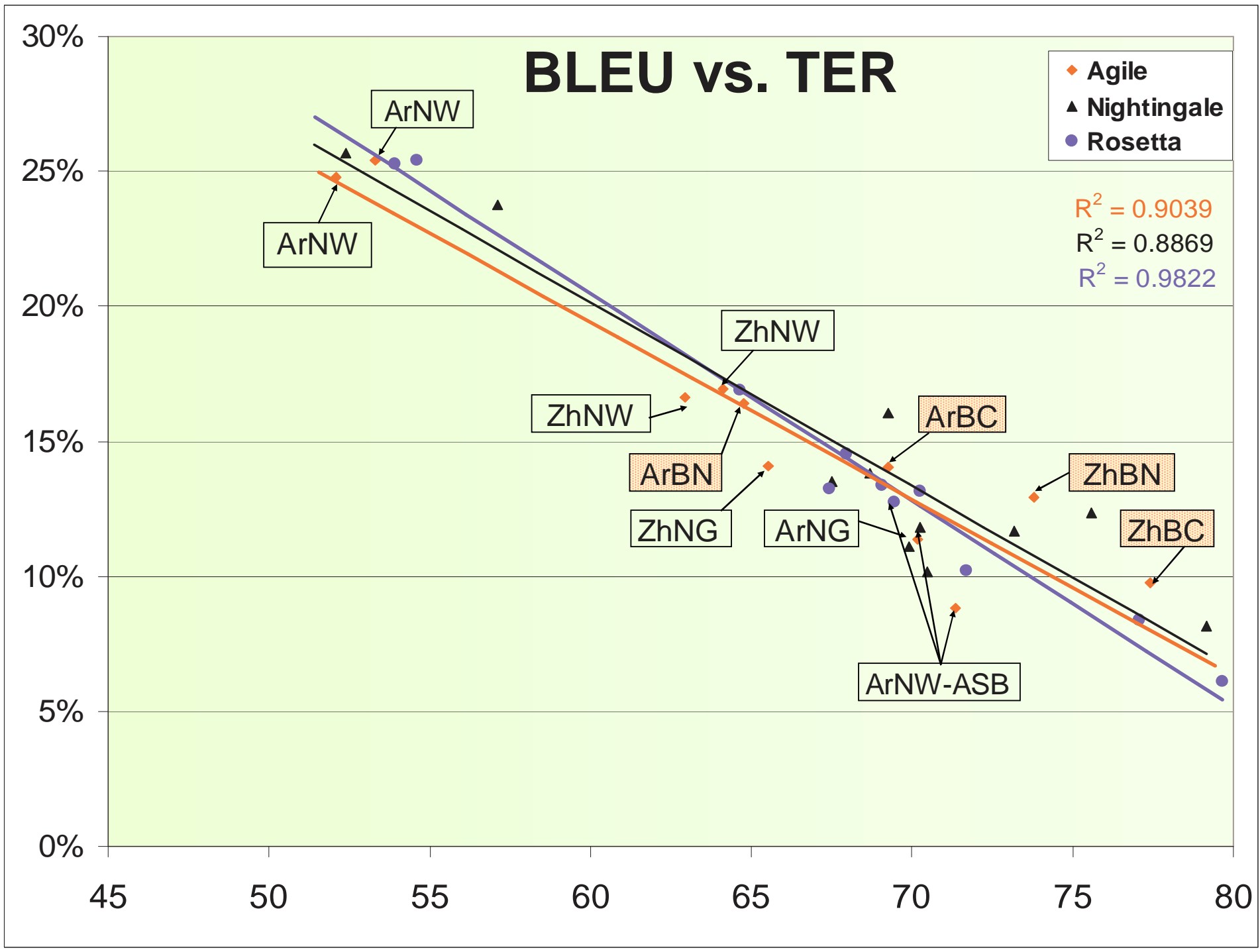
$R^2 = 0.9039$



BLEU vs. TER

- ◆ Agile
- ▲ Nightingale
- Rosetta

$R^2 = 0.9039$
 $R^2 = 0.8869$
 $R^2 = 0.9822$



HTER

- Human editors post-edit MT output to get same meaning as reference translation
- HTER (Human Translation Error Rate)
 - Count all the edit operations

$$HTER = \frac{I + D + S + M}{|R|}$$

- M is number of word or phrase shift movements

Begin

Ref

This is segment 1.
This is segment 2.
This is segment 3.
This is segment 4.
This is segment 5.
This is segment 6.
This is segment 7.
This is segment 8.
This is segment 9.
This is segment 10.

MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

First Pass Editing

editor #1

Ref

This is segment 1.
This is segment 2.
This is segment 3.
This is segment 4.
This is segment 5.
This is segment 6.
This is segment 7.
This is segment 8.
This is segment 9.
This is segment 10.

MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

E-MT

This is MT segment 3.
This is MT segment 7.

HTER

3.5
14.9
77.1
16.0
22.2
15.9
47.9
11.9
12.2
20.8

editor #2

Ref

This is segment 1.
This is segment 2.
This is segment 3.
This is segment 4.
This is segment 5.
This is segment 6.
This is segment 7.
This is segment 8.
This is segment 9.
This is segment 10.

MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

E-MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 4.
This is MT segment 8.

HTER

16.3
21.5
5.3
82.1
52.0
13.2
9.8
38.5
26.7
31.8

editor #3

Ref

This is segment 1.
This is segment 2.
This is segment 3.
This is segment 4.
This is segment 5.
This is segment 6.
This is segment 7.
This is segment 8.
This is segment 9.
This is segment 10.

MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

E-MT

This is MT segment 5.
This is MT segment 6.
This is MT segment 9.
This is MT segment 10.

HTER

8.7
12.4
51.0
39.6
56.7
18.3
27.3
15.0
99.0
33.9

Second Pass Editing

reviewer #1 (low)

Ref

This is segment 1.
This is segment 2.
This is segment 3.
This is segment 4.
This is segment 5.
This is segment 6.
This is segment 7.
This is segment 8.
This is segment 9.
This is segment 10.

E-MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

HTER

5.2
9.5
10.4
29.4
16.3
5.9
30.4
14.2
39.1
6.3

re-E-MT

This is MT segment 1.
This is MT segment 4.
This is MT segment 9.

reviewer #2 (mid)

Ref

This is segment 1.
This is segment 2.
This is segment 3.
This is segment 4.
This is segment 5.
This is segment 6.
This is segment 7.
This is segment 8.
This is segment 9.
This is segment 10.

E-MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

HTER

4.3
11.8
18.4
9.9
25.7
33.9
51.0
16.4
12.7
8.3

re-E-MT

This is MT segment 2.
This is MT segment 3.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.

Final HTER

Ref

This is segment 1.
This is segment 2.
This is segment 3.
This is segment 4.
This is segment 5.
This is segment 6.
This is segment 7.
This is segment 8.
This is segment 9.
This is segment 10.

MT

This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

HTER

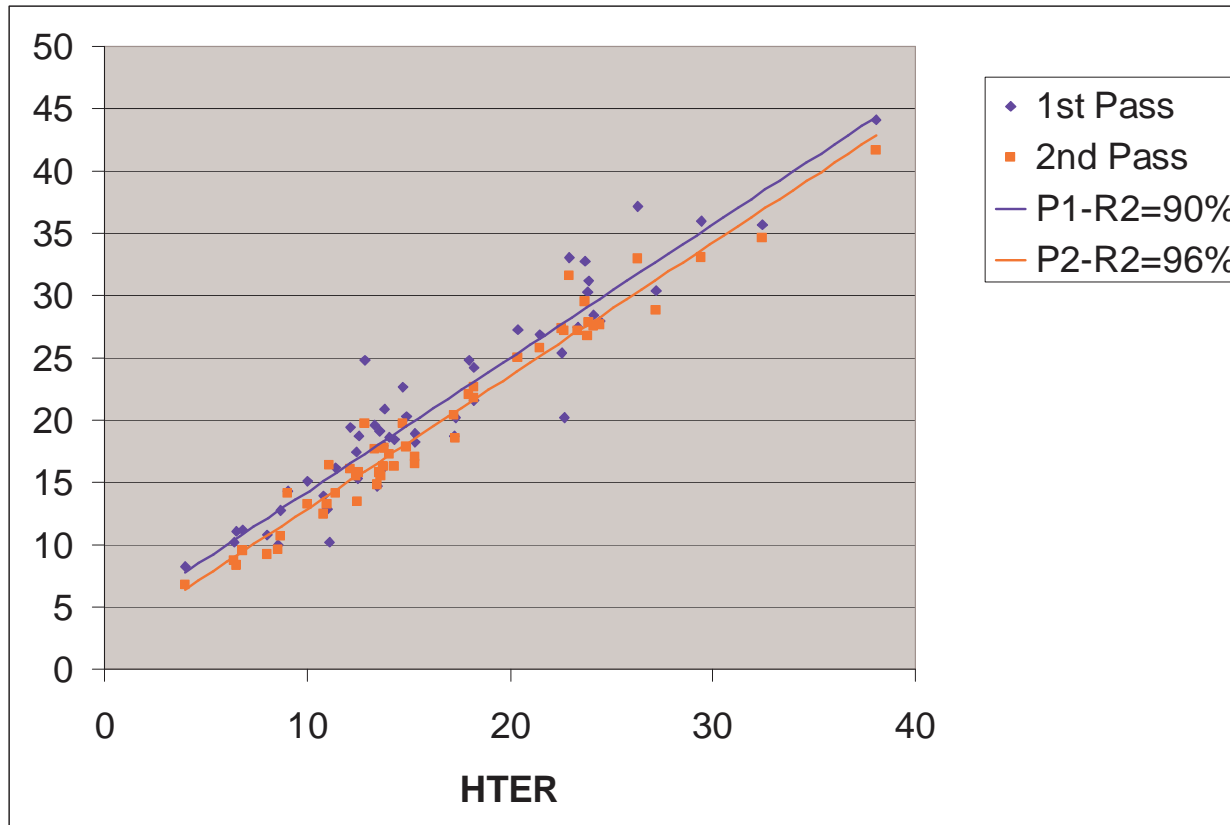
4.3
9.5
10.4
9.9
16.3
5.9
30.4
14.2
12.7
6.3

Final E-MT

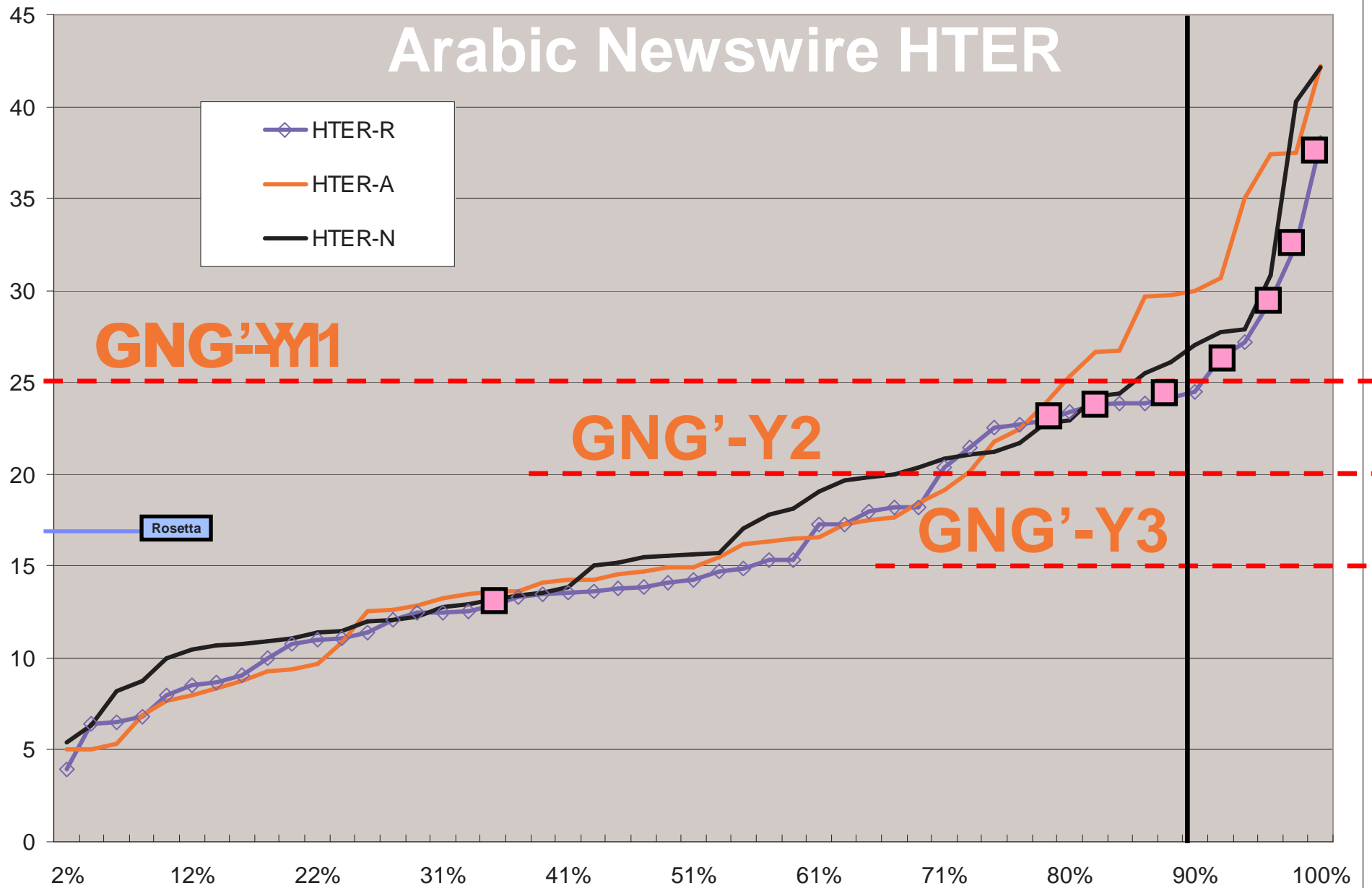
This is MT segment 1.
This is MT segment 2.
This is MT segment 3.
This is MT segment 4.
This is MT segment 5.
This is MT segment 6.
This is MT segment 7.
This is MT segment 8.
This is MT segment 9.
This is MT segment 10.

LDC multipass Post Editing

Rosetta	P1	P2	FINAL
NW	21.2%	19.8%	16.5%
Delta		-1.4%	-4.7%
R2	90%	96%	



Arabic Newswire HTER



The French President to Visit India to Intensify Bilateral Cooperation 0

New Delhi 16 February (Xinhua) said Naftyj Sarna, spokesman for the Indian Foreign Ministry in New Delhi today, Thursday, that the French President, Jacques Chirac will visit India on 19 and 20 Of February
\$ordinal. 1

It is expected to be the signing of a number of agreements and memoranda of understanding during the visit reflectsing the extent of the cooperation between India and France. 1

Such agreements include a declaration on the development of nuclear energy for peaceful purposes, and on cooperation in the field of defense, and a memorandum of understanding on cooperation in the field of tourism. 0

The two countries aim to intensify bilateral cooperation in various fields, including their partnership in the political, economic, defense, space, and civilian nuclear energy. 1

President Jacques Chirac will deliver a keynote speech on economic partnership between India and France. 0

President Chirac is accompanied in the visit by his wife Bernadette Chirac, and the ministers of foreign affairs, defense, economy, finance, industry, foreign trade, tourism as well as some 30 senior managers of major French companies. 0

XIN_ARB_20060212.0073 HTER=15.3% BLEU=.25

The Economic Offer: for Environment-friendly Cars in the Chinese Market/First and Last Addition/ HTER=0%

He pointed out that the two official tests on the Al-Hajeen, which indicates the start of mass production of environment-friendly in China. **HTER=26%**

He added a senior official of the Ministry of Science and Technology that China has achieved remarkable progress in developing the cars will increase local production without doubt their competitiveness in the global market. **HTER=15%**

The Economic Offer: for Environment-friendly Cars in the Chinese Market/First and Last Addition/

Wan pointed out that the two hybrid bus types passed official tests, which indicates the start of mass production of environment-friendly buses in China.

A senior official of the Ministry of Science and Technology added that China has achieved remarkable progress in developing the cars and local production without doubt will increase their competitiveness in the global market.

Can we predict document HTER from document BLEU/TER?

Doc BLEU= 0.25 => Doc HTER= 16.5%+/- SE

NW TEXT		
STD. ERR.	TER	BLEU
Doc=302wd		
Agile	5.0	5.7
Nightingale	5.8	5.7
Rosetta	5.3	5.5

BN AUDIO		
STD. ERR.	TER	BLEU
Doc=770wd		
Agile	4.5	4.9
Nightingale	6.6	4.5
Rosetta	4.2	4.5

To be 95% confident of passing a GNG threshold one needs 100 docs (for a stderr of 0.5% in HTER) around that level:

==> **need DEV SETS of 1000 docs per condition**

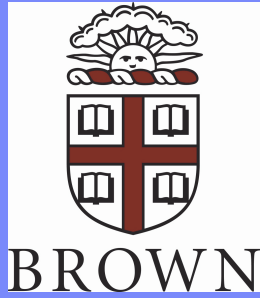
Can we predict document HTER from document Post Editing @IBM?

Subset of Arabic NW: 18 docs Post-Edited @ IBM

Post Editing	Agile	Nightingale	Rosetta
LDC HTER	21.01%	20.18%	19.19%
IBM HTER	34.02%	32.94%	32.91%
R2	62%	59%	58%
STD ERR	5.9%	5.0%	5.9%

+65%

- Similar results for Chinese



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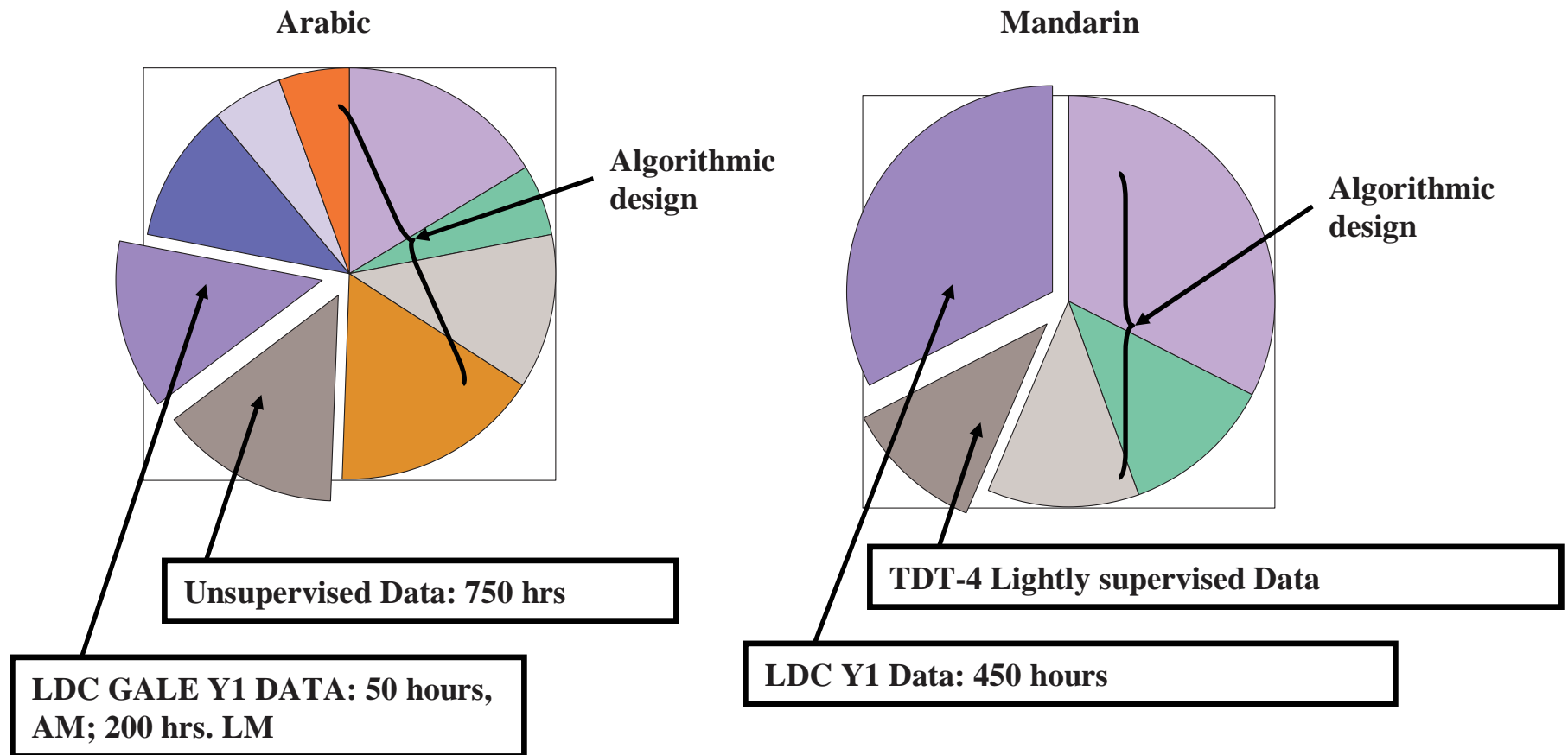
The 2006 Rosetta Transcription Effort



Net Rosetta Progress This Year

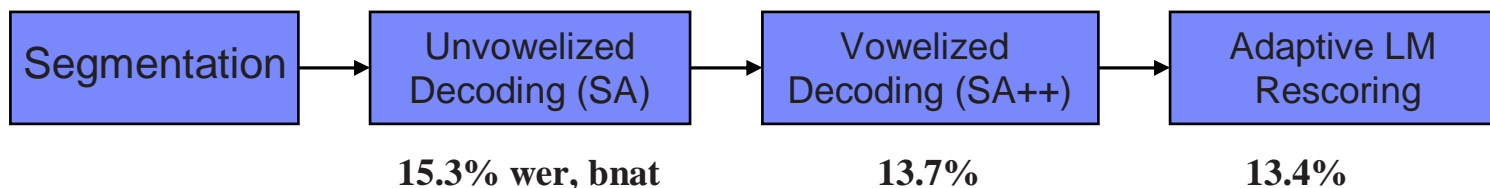
	Mandarin (RT04 Test set)	Arabic (RT04 Test set)
December	23.2%	21.7%
June	13.5%	12.6%
Improvement	42%	42%

Where did the improvement come from?



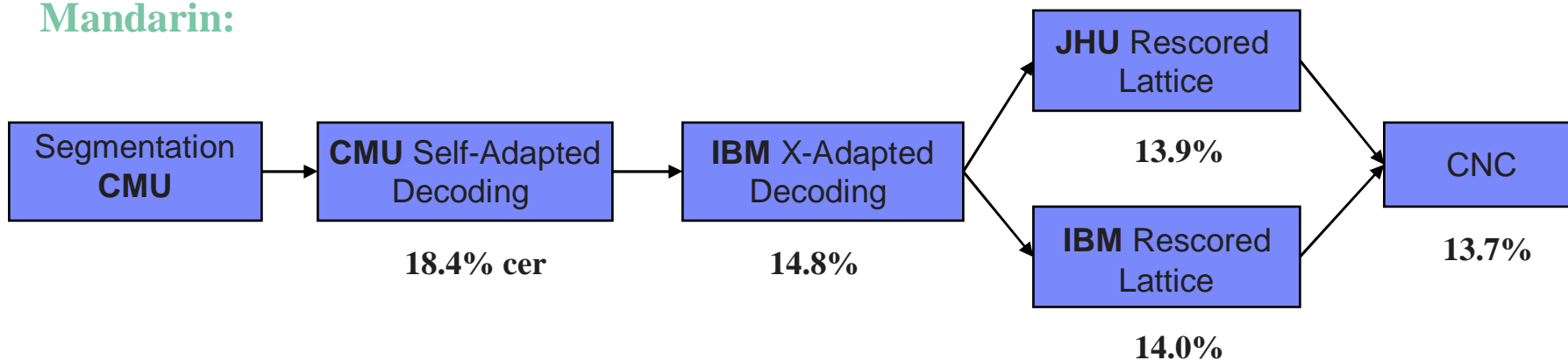
Transcription Flow Charts

Arabic:



* Numbers on subset of BNAT and BCAD

Mandarin:



* Numbers on subset of LDC2006E10 and dev05bcm

What happened between Sep'05 and July'06 ?

- And the improvements come from ...
- LDC data : 1.2%
- Unsupervised Training : 1.3%
- Vowelization : 2.0%
- Big Vocabulary : 1.5%
- Cross-Adaptation Unvowelized-Vowelized : 1.0%

Pronunciation Probabilities

- Vowelized Setup : 617k vocabulary, 2m pronunciations
- Forced alignment on training data (incl. unsupervised BN-03)

Pron. Prob.	RT-04	BNAT-05	BCAD-05
no	16.0%	17.3%	26.0%
yes	14.9%	16.4%	25.1%

- Developed technology to cope with 2 million pronunciations
- Significant improvements from pronunciation probabilities

Vowelization and Broadcast Conversations ..

- ML models : VTLN, FMLLR, MLLR

	RT-04	BNAT-05	BCAD-05
Unvowelized	17.0%	18.7%	25.4%
Vowelized	14.9%	16.4%	25.1%

- Significant improvements on Broadcast News, but not on Broadcast Conversations ! -> Need to investigate:
 - Dialect issue?
 - BC training data with vowelized transcripts?

Evaluation Results

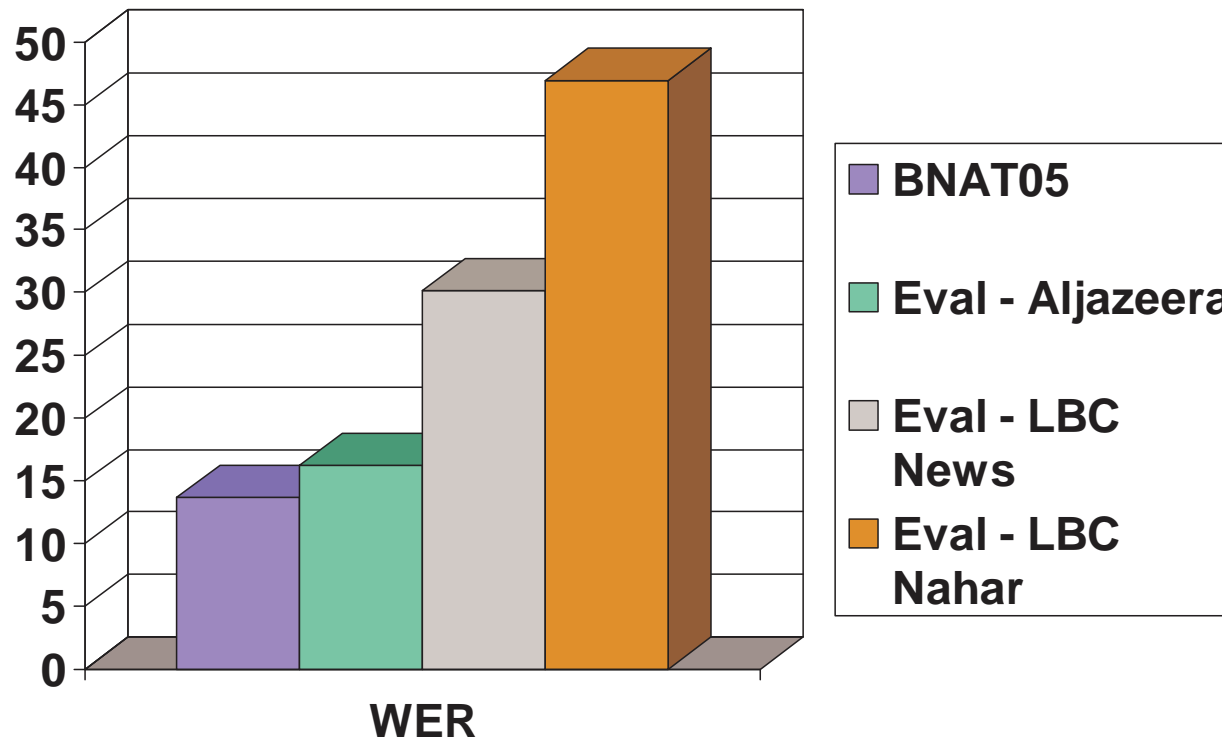
	BC	BN
Arabic - Dev	21.5	13.7
- Test	34.0	24.4
- HTER	35.6	29.2
Mandarin - Dev	20.7	12.9
- Test	24.1	13.4
- HTER	37.1	32.4

Really big mismatch
between dev & test

We hit the target!

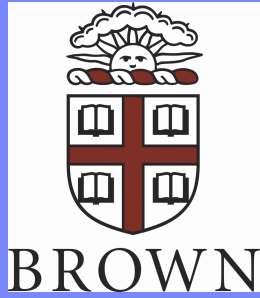
Some mismatch
between dev & test

One Key Lesson: Need wider variety of training data



Very little training data for LBC – poor results on test set.
In the future we would like to have at least 10h of speech from each source.

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Predicting the WER on New Test Sets

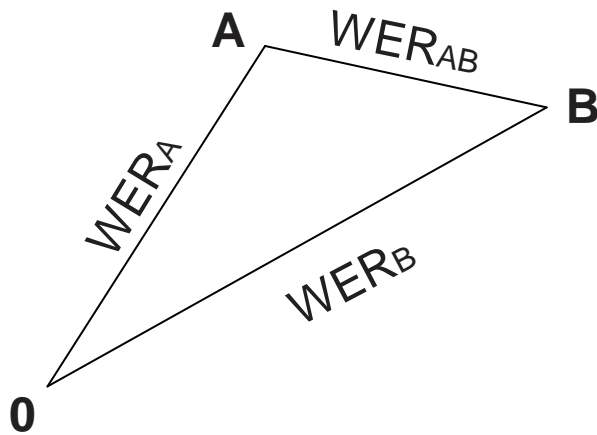


Motivation

- Rapidly assess the performance of an ASR system on a new test set without the need of a reference transcript
- Creating an accurate reference is a time-consuming process
 - Expertise may not be readily available (e.g. foreign languages)
 - Have to rely on other institutions to provide reference (e.g. NIST)
- Applications
 - Predict system performance in government evaluations 😊
 - Select data for (un)supervised training (active learning)
 - Change system configuration to minimize predicted WER

How can we compute $WER_{A'}$?

Training: all WERs known



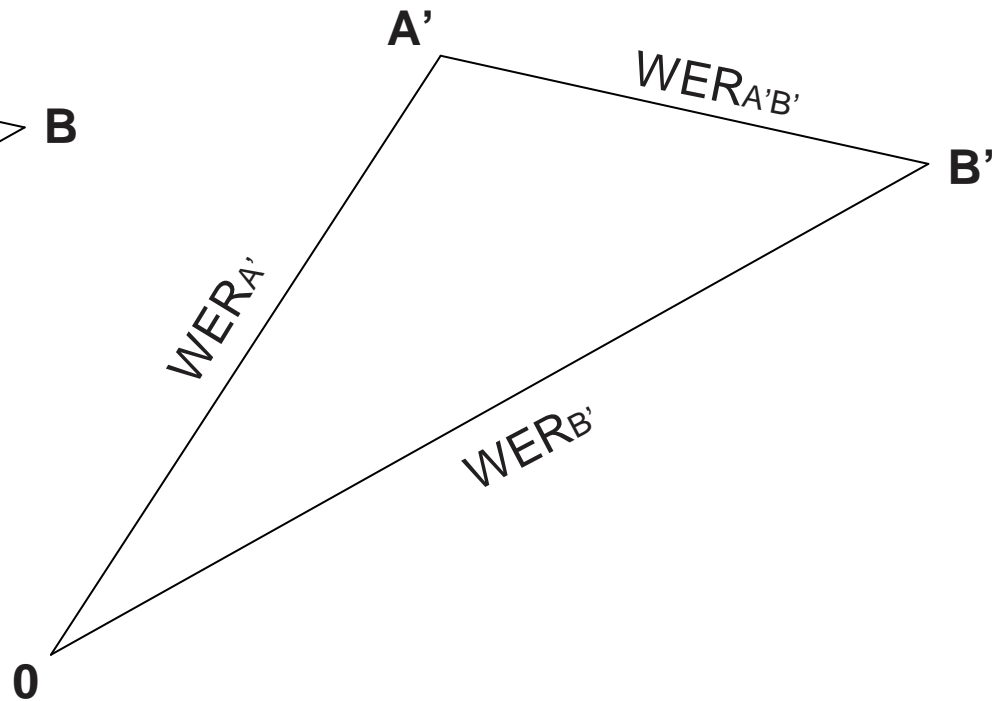
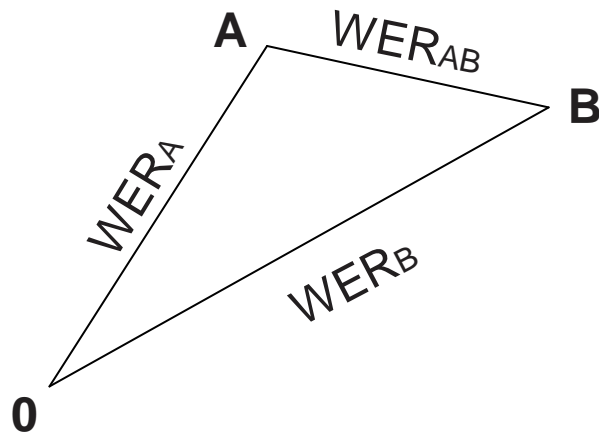
Test: only $WER_{A'B'}$ known



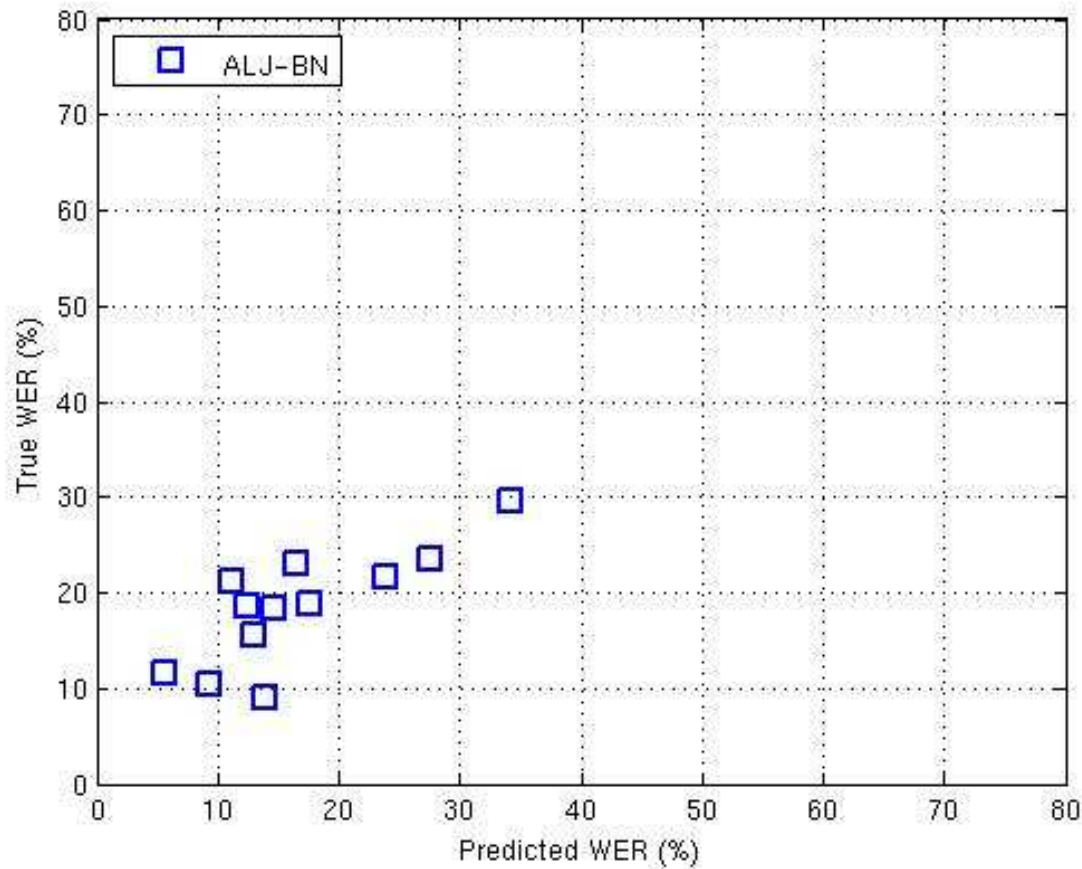
How can we compute $WER_{A'}$?

Training: all WERs known

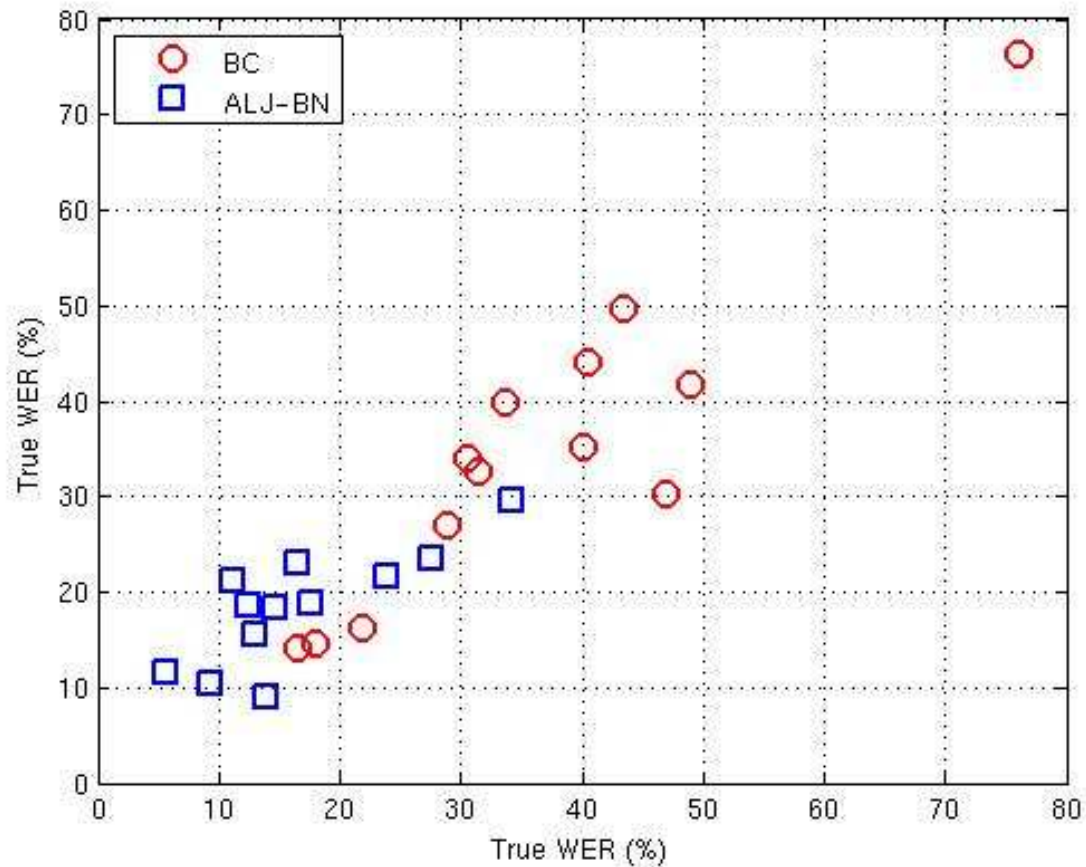
Test: only $WER_{A'B'}$ known



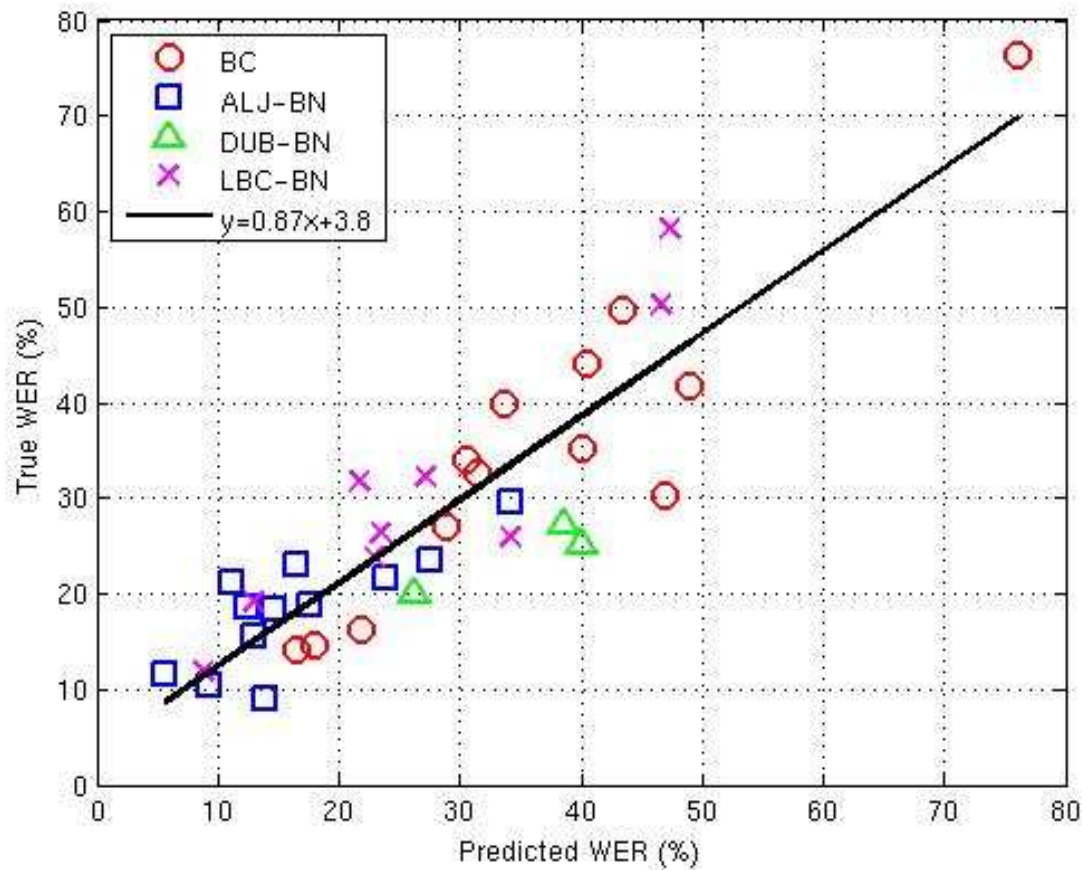
Performance on the 2006 GALE evaluation data



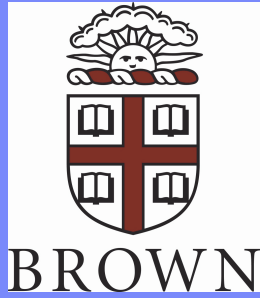
Performance on the 2006 GALE evaluation data



Performance on the 2006 GALE evaluation data



True WER=29.2%, predicted WER=30.0%, CORR=0.87, MAD=5.4



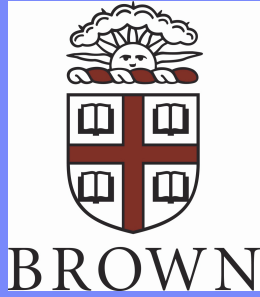
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Rosetta:
MT GALE GnG06 Report



IBM



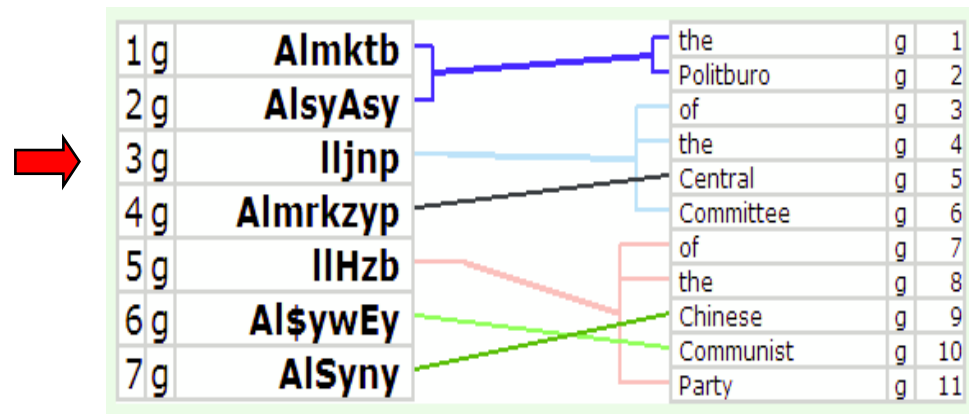
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A Direct Translation Model II



How many phrases do we need?



- N-M blocks (Used by most SMT systems)
 - General
 - All possible blocks extracted
 - 40-50M blocks in Arabic
 - Sparsity problems

Iljnp <i>committee</i> of the commission commission of the committee the committee of the commission on the commission committee of	Almrkzyp central the central <i>of the central</i> of central and the central and central , central 's central
<i>of the central committee (11)</i> of the central committee of (11) the central committee of (8) central committee (7) committee central (2)	

Direct Translation Model

- **Joint future: Jump, Target Sequence**

$$p(T, j | S)$$

- j =jump, which is the number of positions from the previously translated source word position
- Integrates Distortion and Word-selection model

- **Features**

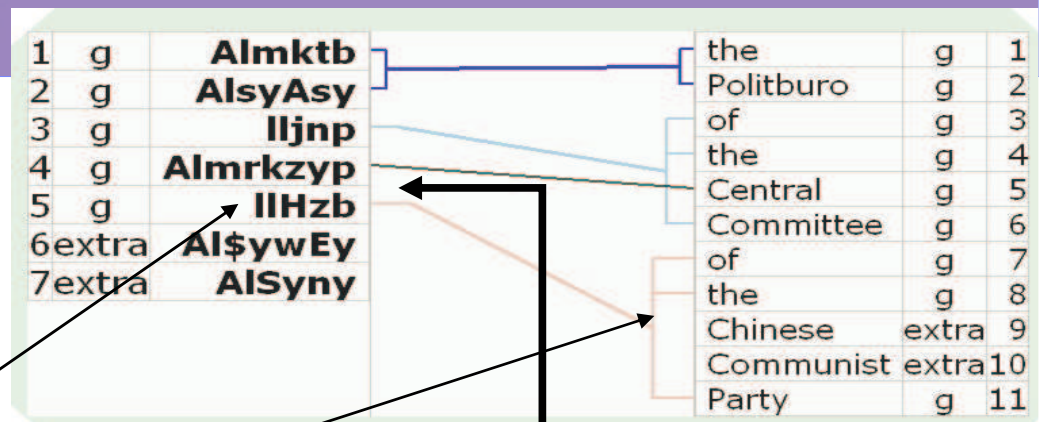
- Lexical:
 - Left and Right context of source sequences
 - Questions about the left context of a target sequence
- Part-of-speech, Segmentation

- **Features shared across phrase blocks**

- Feature parameters trained to maximize log-likelihood
 - **No direct optimization of any translation quality metric (BLEU, TER)**

- **Details in an upcoming paper**

Features



- **MaxEnt Block Example**

33 0.0876793 0.0274136 | llHzb | of the VAR_1 party | 0 0 -1 0 || l# l# Hzb

- **Block Internal: Seg Features**

Cnt	Alpha	Jump	Tgt	Seg
1107	1.047	-2	of	l#
3120	0.989	-1	of	l#
55461	1.319	1	of	l#
7009	1.225	2	of	l#

- **Block Context Feature**

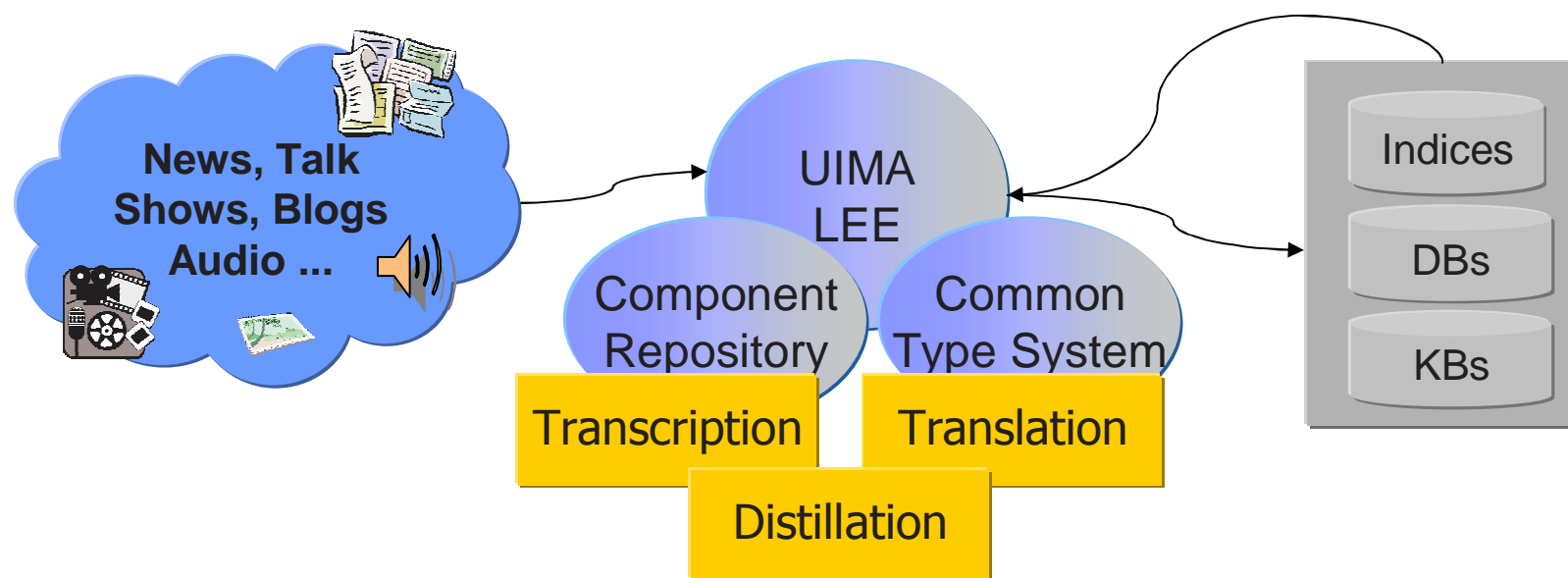
- 11 1.66021 0.0330579 1024 -1 party llHzb // communist Al\$ywEy chinese AlSyny

- **New Feature ~ coding time + 8 hours training + 1 hr decode time**

Experiments - NIST

Feature Types	# of feats (MT05)	MT-05	MT-06 (NIST)
MaxEnt Decoder Lexical Feats	520,210	48.21	
+Lexical Context	1,551,582	49.24	
+Segmentation Feats	3,063,023	49.51	
+Part-of-Speech Feats	3,370,901	49.87	
+Distortion Feats	3,412,210	49.98	38.61
Block Decoder		49.06	36.92

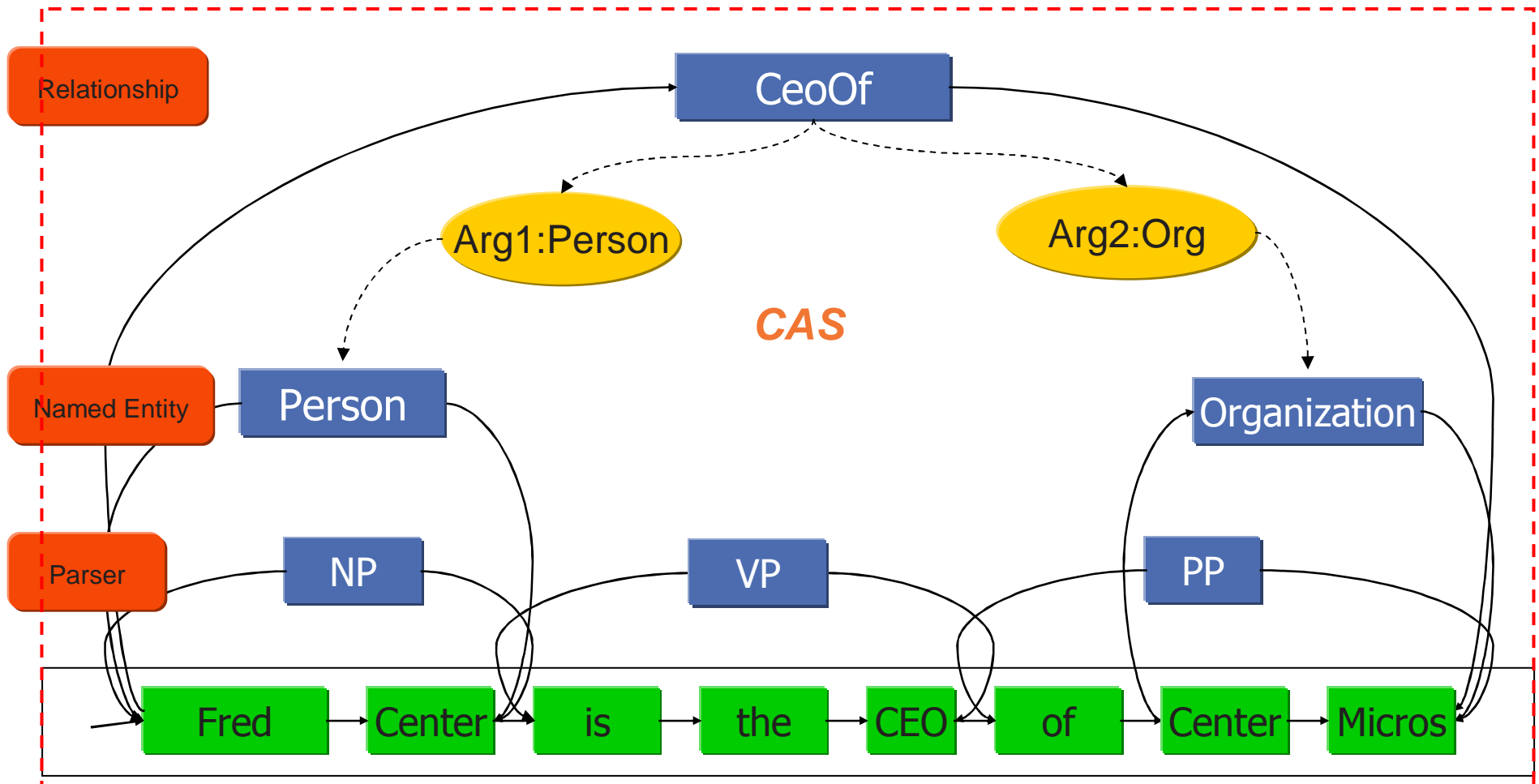
UIMA: ARCHITECTURE FOR DARPA GALE



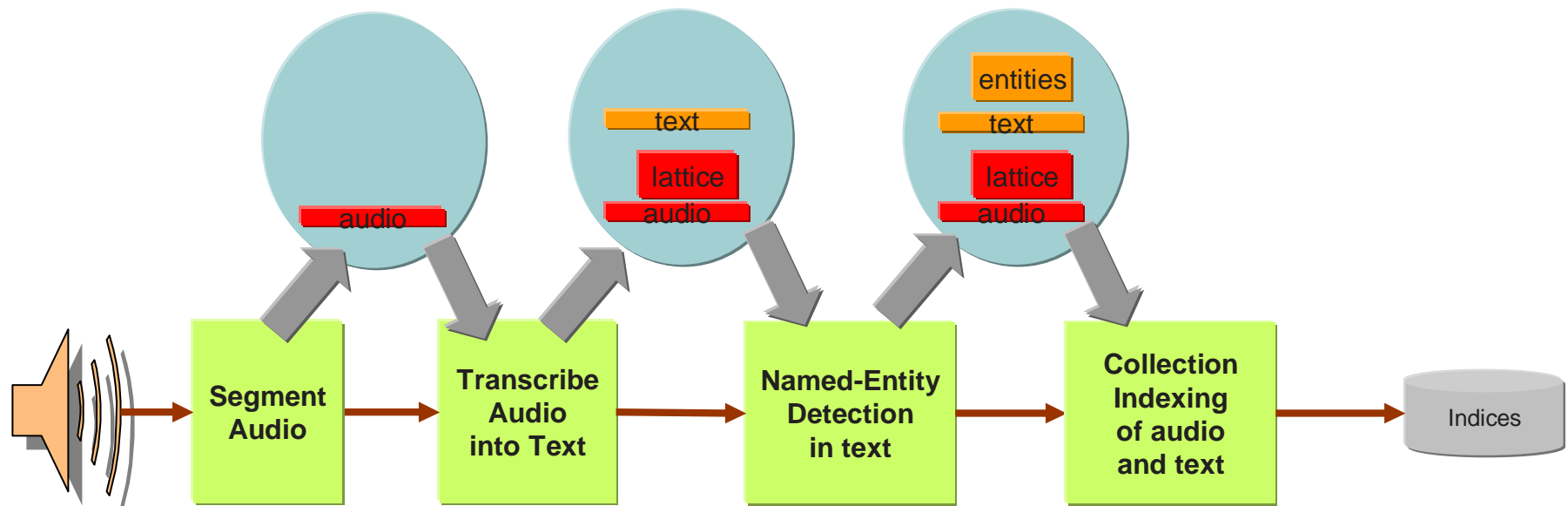
- Highly-distributed plug-and-play architecture
- Support for multi-modal sources
- Support for local/remote heterogenous components
- Open Source



UIMA's Basic Building Blocks are Annotators. They iterate over an artifact to discover new types based on existing ones and update the Common Analysis Structure (CAS) for upstream processing.



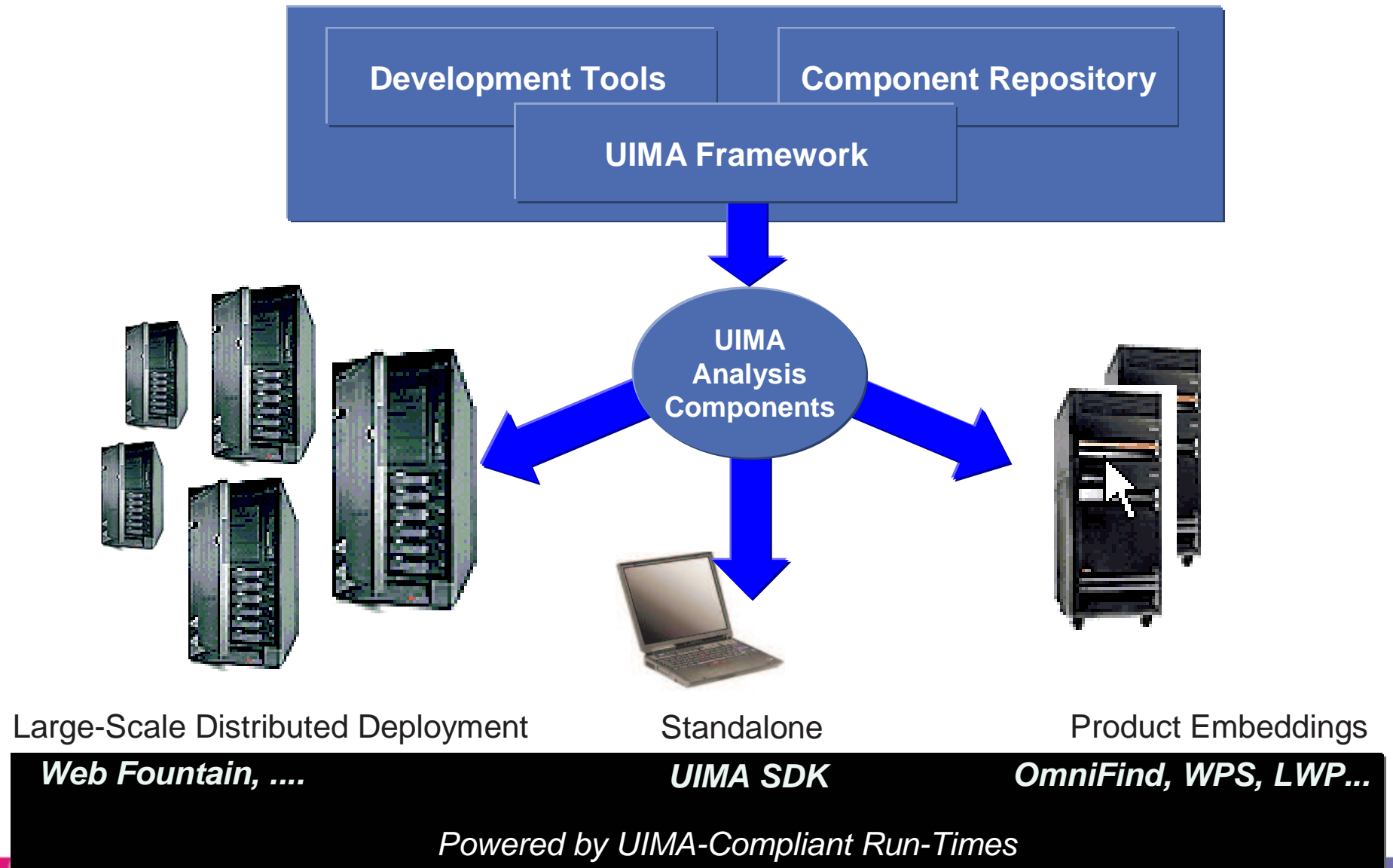
Common Annotation Structure (CAS): *Multiple Subject of Analysis (SOFA) in CAS Supports Multi-Modal Analysis*



- Multiple views of an artifact can each support independent sets of attributes
- Focus can change from audio to text to both
- Attributes directed to one or more SOFAs



A common platform for development, composition and deployment of multi-modal analytics into different carriers.



Track thousands of Web sites in one place: [Newsburst](#)

SEARCH **ADVANCED SEARCH**

Enterprise Software >> Open source

IBM dives deeper into corporate search

Published: August 7, 2006, 8:01 PM PDT

By **Edgar Mills**
Staff Writer, CNET News.com

TalkBack | E-mail | Print | TrackBack

IBM is promoting a new standard to allow interoperability between software that helps corporations search for and analyze unstructured data across their corporate networks, including e-mails, Word documents and anything that is not formatted in columns and rows.

The company was set to release on Monday a new version of its WebSphere Information Integration OmniFind Edition corporate information management tool. It integrates technology called Unstructured Information Management Architecture (UIMA) that IBM designed to improve the processing of text within documents and other unstructured content sources to help find relationships and meaning beyond just keywords.

IBM, a longtime supporter of the open-source movement in which developers freely write and modify software and share code, also is presenting UIMA to the Open Source Technology Group, a network of online technology resources. The updated software tool is available from IBM now and is expected to be available through the [SourceForge developers Web site](#) by the end of the year.

"IBM has been investing in a huge initiative since 2001 in information

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Today in News.com **EXTRA**

Mutant mice created with ease. Also: Top 10 technologies we miss.

[Read all about it](#)

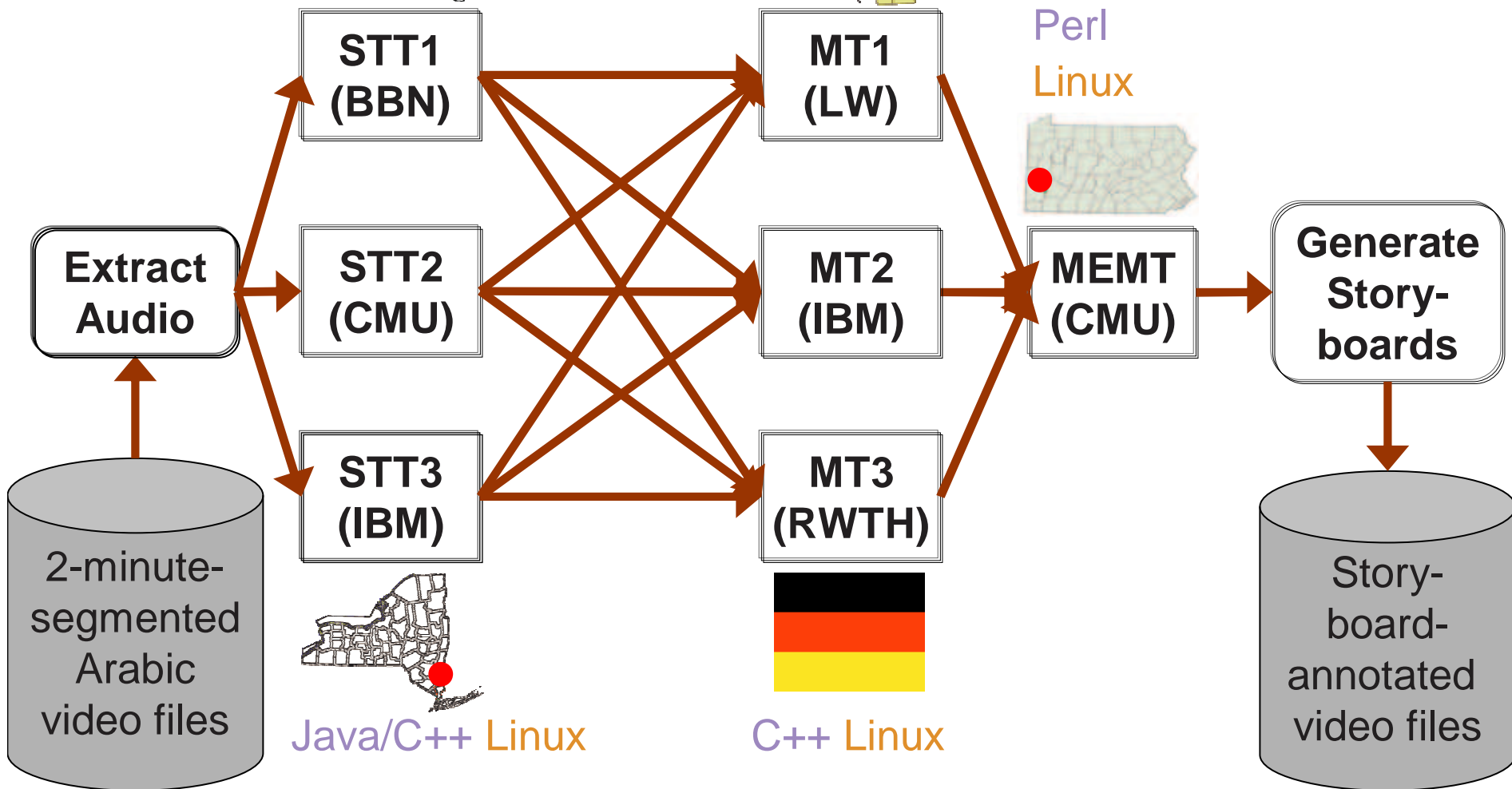
THE PULSE Most popular headlines

- Another way past Windows antipiracy found

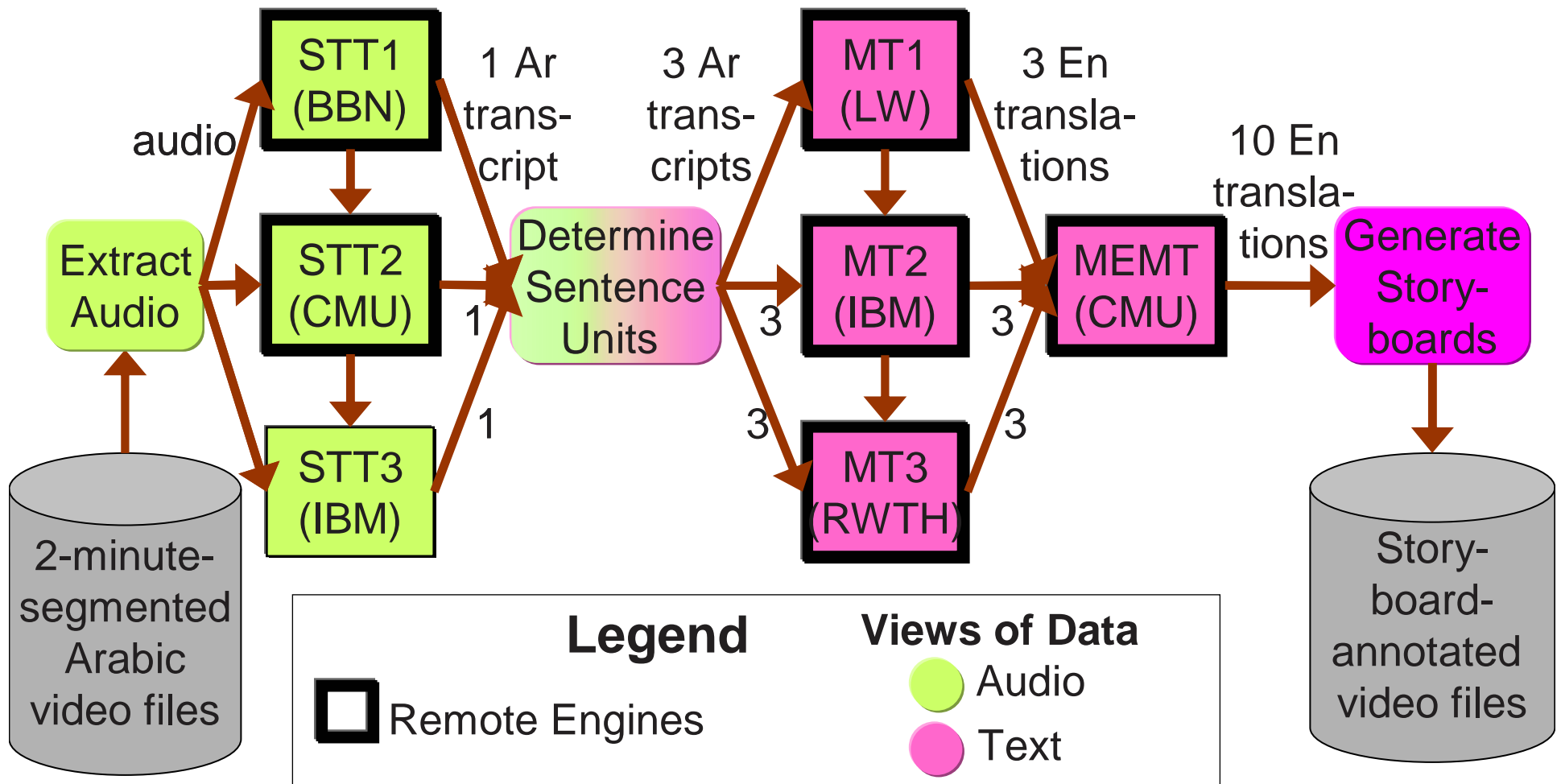
Perl
Windows
2003
Server

Java/C++
Windows
2003
Server

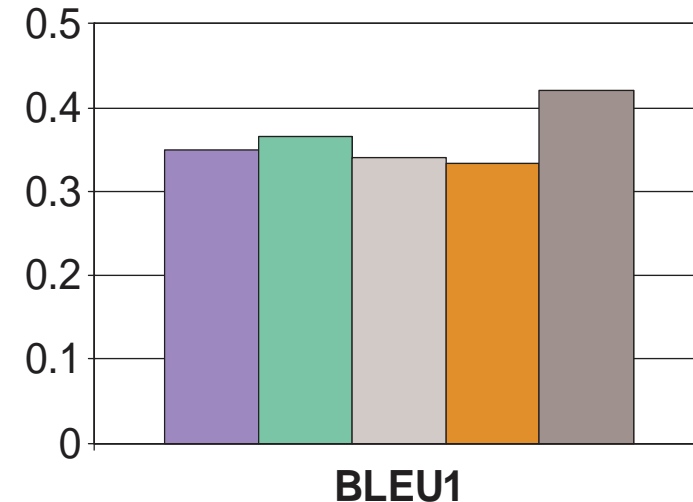
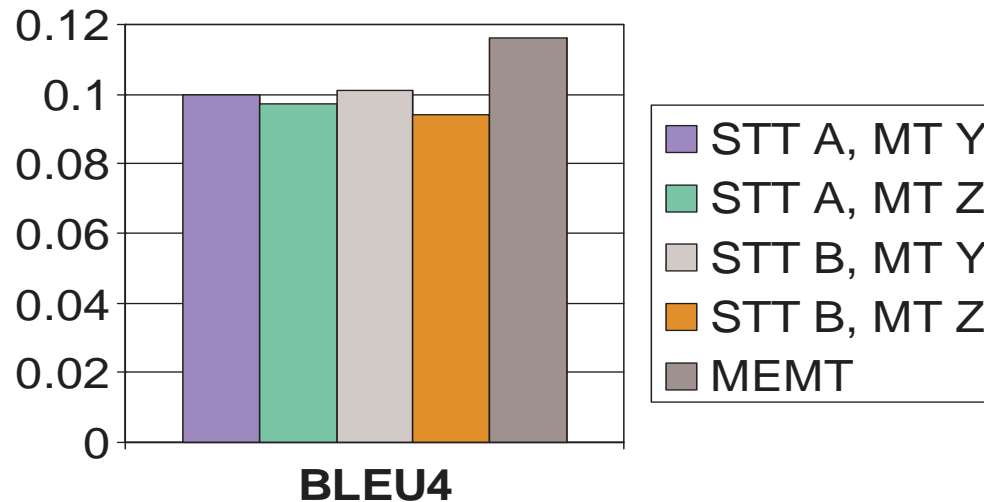
Java/C++/
Perl
Linux



ProData Flow: Serial



IOD Enables On-Line MEMT, Increased Accuracy

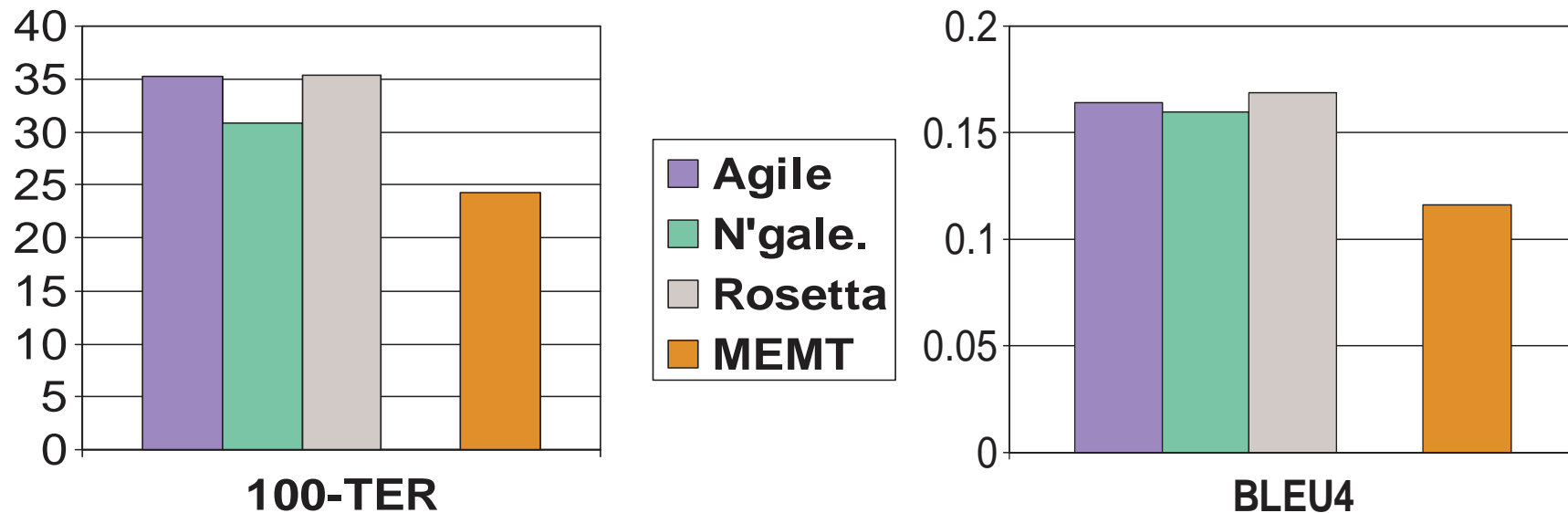


- GNG Arabic speech test set (34 of 37 audio files)

- Case-insensitive evaluation

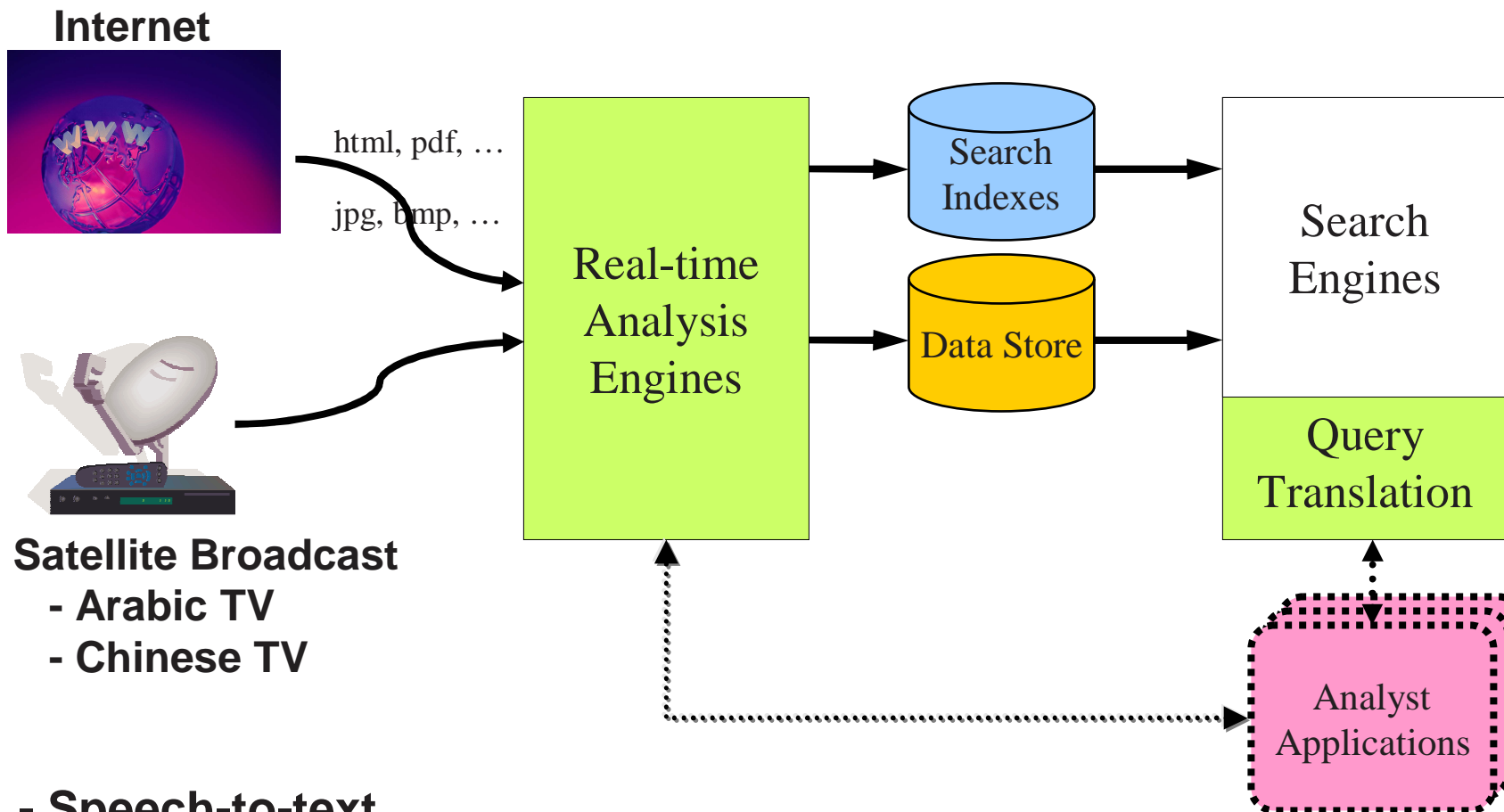
System	TER	BLEU4	BLEU1	METEOR
STT A, MT Y	75.9	0.100	0.349	0.405
STT A, MT Z	75.4	0.097	0.366	0.396
STT B, MT Y	74.7	0.101	0.340	0.405
STT B, MT Z	74.7	0.094	0.334	0.395
MEMT	75.7	0.116	0.421	0.440
MEMT % gain	-1	+15	+15	+9

GNG Results vs. IOD



- **Research systems ~50% better than product engines**
- **Case-sensitive GNG vs. case-insensitive IOD**
- **→ Significant work to productize**

TALES: Multimodal Trans-lingual Analytics



Satellite Broadcast
- Arabic TV
- Chinese TV

- Speech-to-text
- Statistical machine translation
- Cross-lingual search

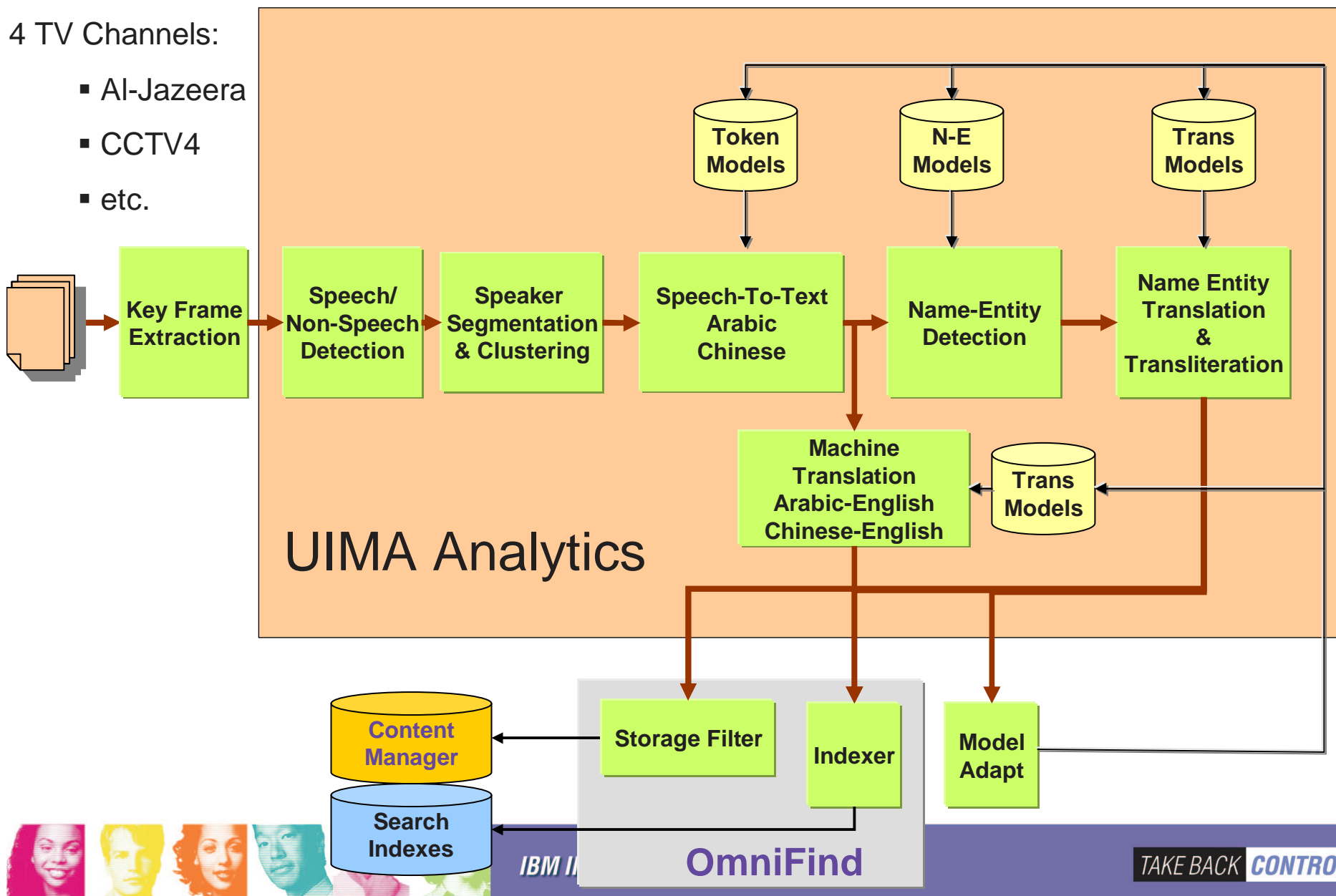
Data available as quickly as acquired
- 5 min delay on video content
- 15 min delay on web pages



Video Processing Flow

4 TV Channels:

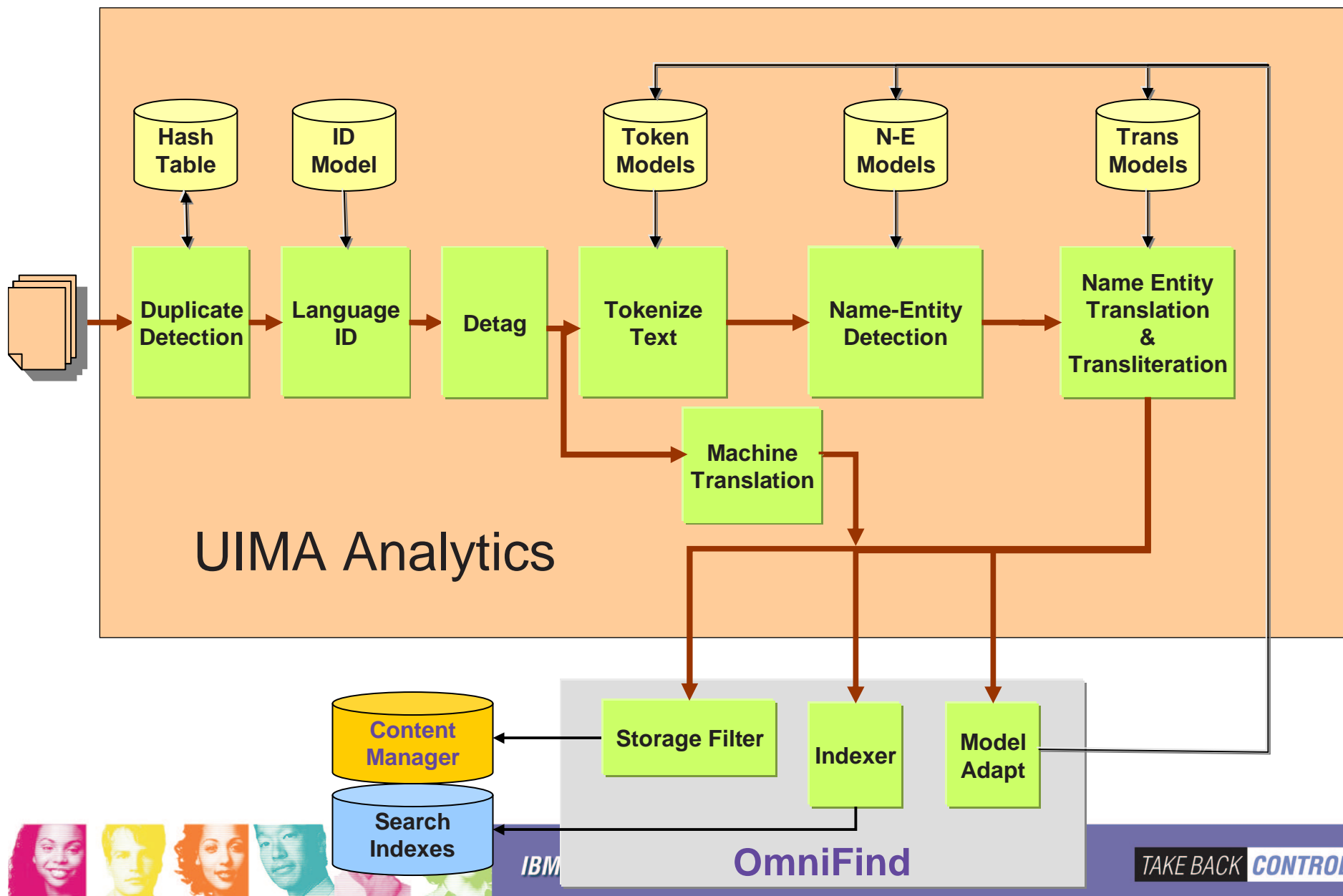
- Al-Jazeera
- CCTV4
- etc.



IBM I

TAKE BACK CONTROL

Text Processing Flow



TALES Foreign Broadcast Video Monitoring and Search System

English Query

Arabic Video in pop-up player

Arabic Text

Translated Speech

English Translation

- UIMA-based trans-lingual search technology:
 - Speech-to-Text
 - Machine Translation (English, Arabic, Chinese)
 - Advanced Text Analysis (language identification and translation, named entity extraction and translation)
 - Cross-lingual Information Retrieval



Thankyou



شکرا ، 謝謝