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Training a Super Model Look-Alike: Featuring Edit Distance, N-Gram Occurrence, and One Reference Translation

Eva Forsbom, Uppsala University

`evafo@stp.ling.uu.se`

MT Summit IX

Workshop on Machine Translation Evaluation

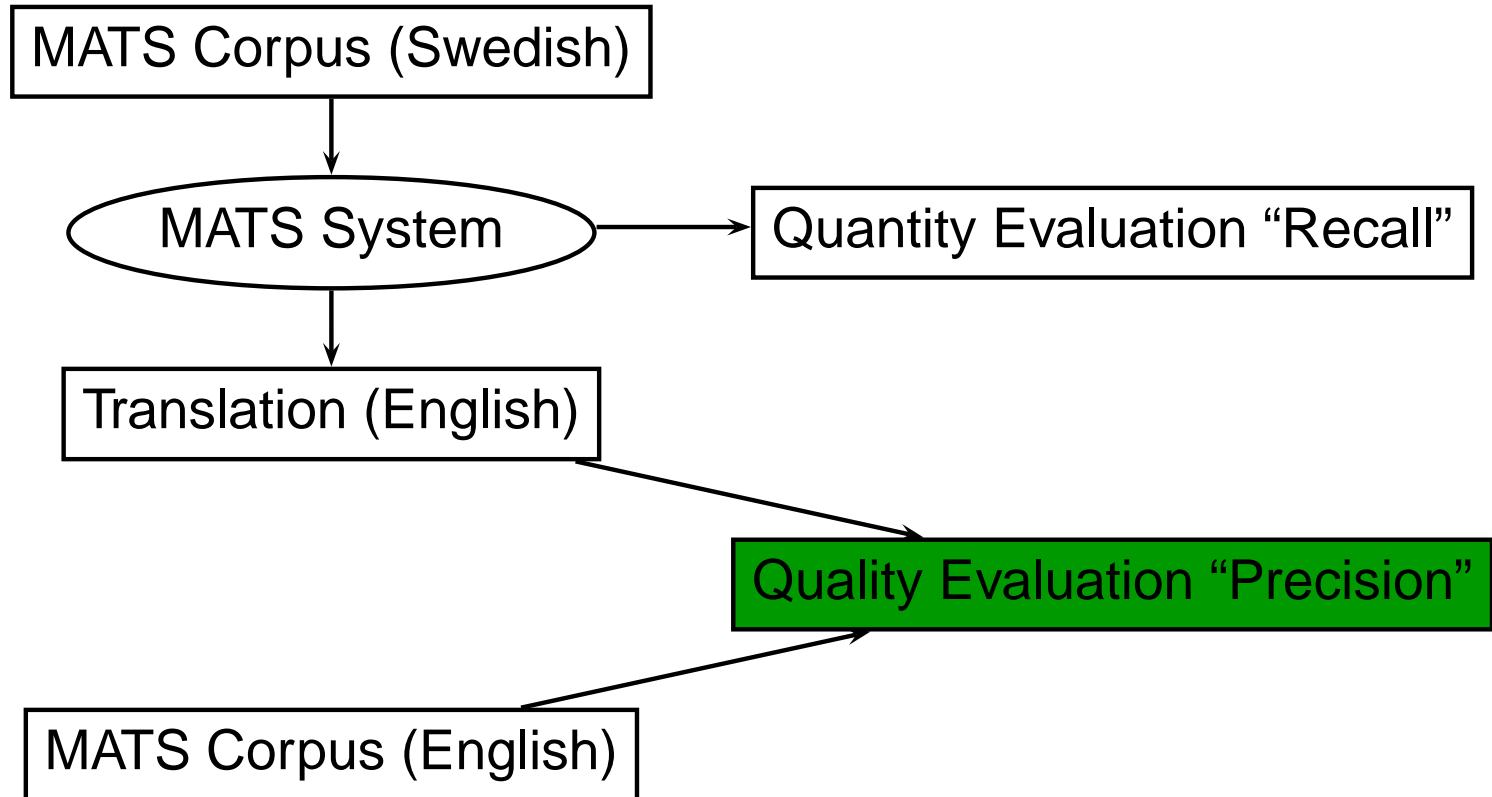
Towards Systematizing MT Evaluation

Saturday, September 27, 2003

New Orleans, Louisiana, USA



Evaluation Context





WANTED: Translation Quality Measure!

- Be automatic.
- Work for various kinds of evaluations:
 - declarative,
 - progressive,
 - diagnostic.
- Work at various levels:
 - system,
 - document,
 - segment.
- Work for various text types (news/technical manuals).
- Work with one reference translation.
- Exist.



Applicants

- Edit distance: Word Accuracy (Alshawhi et al. 1998)
- N-gram occurrence: BLEU (Papineni et al. 2001)
- N-gram occurrence: NIST (Doddington 2002)
- Possible redefinitions...





Heat 1: Edit Distance – Word Accuracy

$$WA = \left(1 - \frac{d + s + i}{r} \right)$$

where

d = deletions

s = substitutions

i = insertions

r = length of reference



WA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment.
- Work for various text types.
- Work with one reference translation. ✓



WA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment. **Failed!**
- Work for various text types.
- Work with one reference translation. ✓



WA Failed for Segments!

Word Accuracy can result in a score less than 0 if the length of the reference is shorter than the length of the candidate:

Src: Tätningsring

Cand: Sealing ring length = 2

Ref: Seal length = 1

$$WA = \left(1 - \frac{1 + 1 + 0}{1} \right) = -1$$



WA – Redefinition

Word Accuracy For Translation:

$$\text{WAFT} = \left(1 - \frac{d + s + i}{\max(r, c)} \right)$$

where

d = deletions

s = substitutions

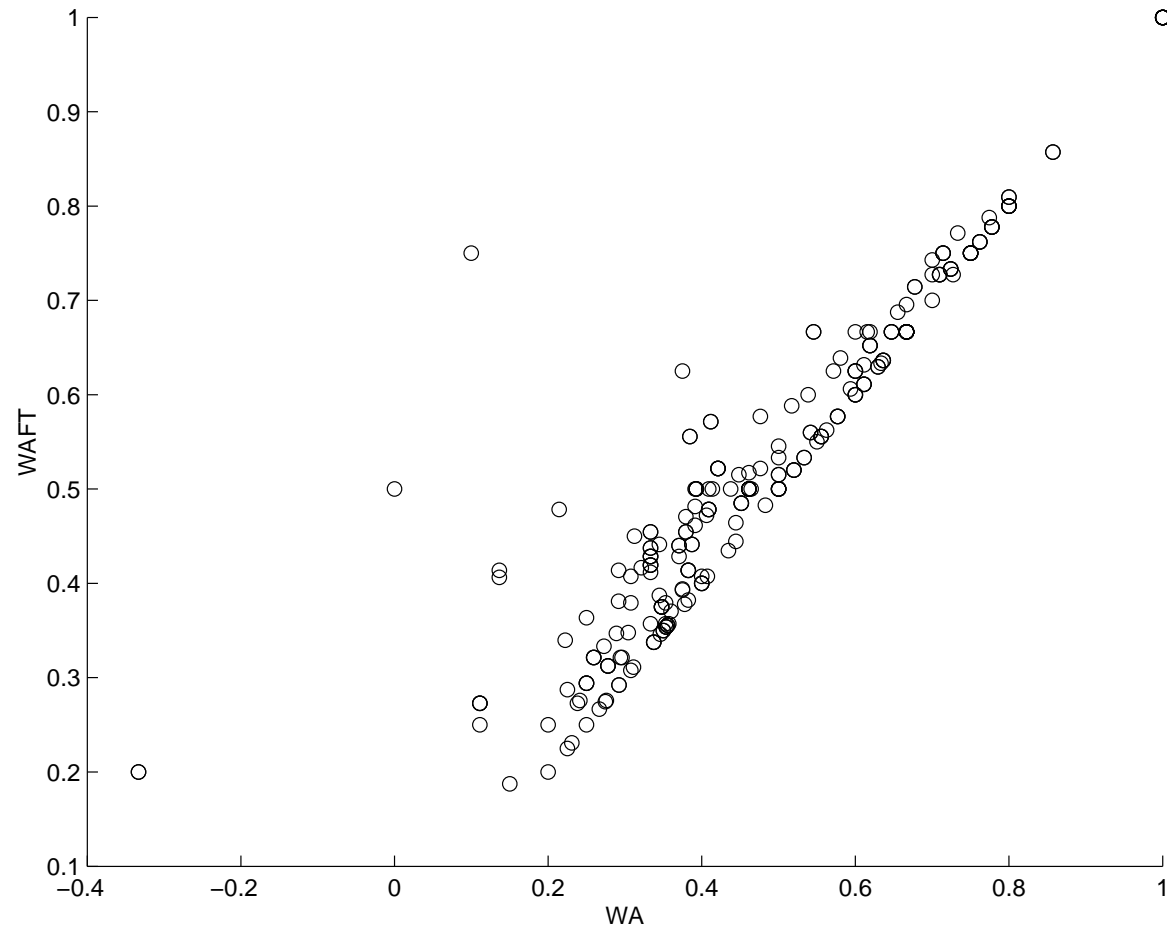
i = insertions

r = length of reference

c = length of candidate

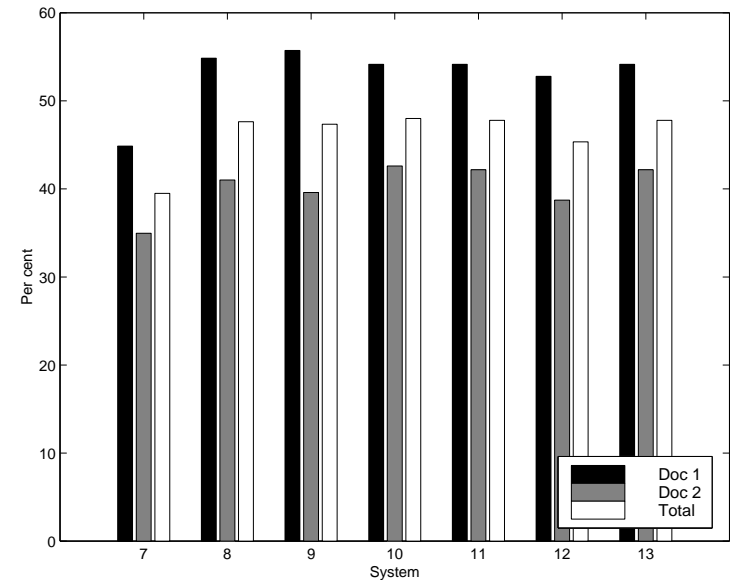
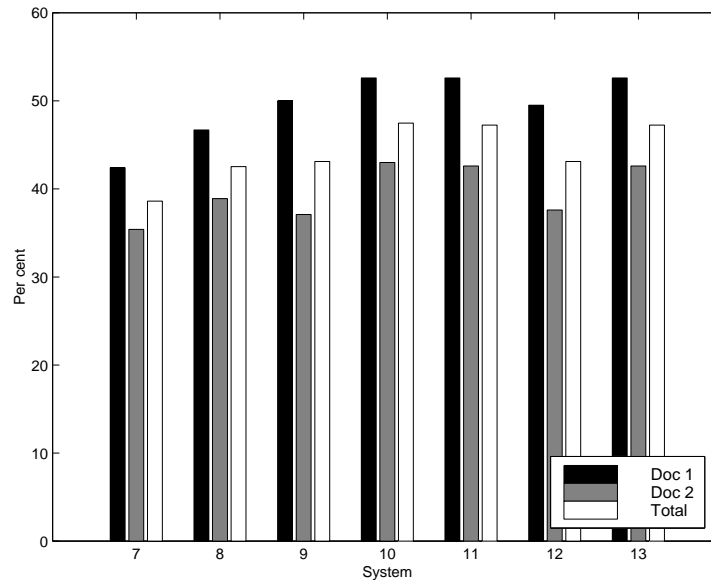


WA vs. WAFt for Segments (LREC 6/4)





WA vs. WAFT: Docs & Sys (LREC 6/4)



- Ranking is the same (except for systems 8 and 9 on document 1).
- WAFT yields slightly higher scores than WA.



WAFT Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment.
- Work for various text types.
- Work with one reference translation. ✓



WAFT Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment. ✓
- Work for various text types.
- Work with one reference translation. ✓



WAFT Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment. ✓
- Work for various text types. ✓
- Work with one reference translation. ✓



Heat 2: N-Gram Occurrence

BLEU

- Score = $[0, 1]$;
- Compensates for difference in length by a brevity penalty;
- Applies equal weights for all n-grams.

NIST

- Score = $[0, ?]$;
- Compensates for difference in length by another brevity penalty;
- Applies different weights for the n-grams.



NIST Scoring Card

- Work for various kinds of evaluations:
 - declarative
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment.
- Work for various text types.
- Work with one reference translation.



NIST Scoring Card

- Work for various kinds of evaluations:
 - declarative **Failed!**
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment.
- Work for various text types.
- Work with one reference translation.



NIST Failed for Declarative Evaluation!

NIST can yield different scores for equivalent objects of evaluations, due to its weighting method:

Src: Antal
Cand: Number
Ref: Number

Src: Beteckning
Cand: Designation
Ref: Designation

NIST = 4.6267

NIST = 8.0311



Heat 2: N-Gram Occurrence – BLEU

$$\text{BLEU} = \text{BP} \cdot \exp \left(\sum_{n=1}^N w_n \log p_n \right)$$

where

$$\text{BP} = \begin{cases} 1 & \text{if } c > r \\ e^{(1-\frac{r}{c})} & \text{if } c \leq r \end{cases}$$

r = length of reference

c = length of candidate

$$N = N_{\max} (=4)$$

$$w = \frac{1}{N}$$

$$p = \frac{\sum_{C \in \{Cand\}} \sum_{n\text{-grams} \in \{C\}} \text{Count}_{clip}(n)}{\sum_{C \in \{Cand\}} \sum_{n\text{-grams} \in \{C\}} \text{Count}(n)}$$



BLEU Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment.
- Work for various text types.
- Work with one reference translation.



BLEU Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment. **Failed!**
- Work for various text types.
- Work with one reference translation.



BLEU Failed for Segments!

BLEU measure is not defined for segments with a length shorter than N_{max} :

Src: Cylinder, underdel

Cand: Bottom cylinder *length* $< N_{max}$

Ref: Cylinder bottom

$$\text{BLEU} = \text{BP} \cdot \exp \left(\sum_{n=1}^N w_n \log p_n \right) = \textit{undefined}/0$$



BLEU – Redefinition

$$\text{First draft} = \text{BP} \cdot \exp \left(\sum_{n=1}^N w_n \log p_n \right)$$

where

$$N = \begin{cases} N_{max} & \text{if } c \geq N_{max} \\ c & \text{if } c < N_{max} \end{cases}$$



First Draft Still Failed for Segments!

First draft measure is not defined for segments lacking co-occurrence for at least 1 n-gram level:

Src: Ledningsnät för bränslepump

Cand: Cable harness for fuel pump **no 3- or 4-grams**

Ref: Fuel pump cable harness

$$\text{First draft} = \text{BP} \cdot \exp \left(\sum_{n=1}^N w_n \log p_n \right) = \text{undefined}/0$$



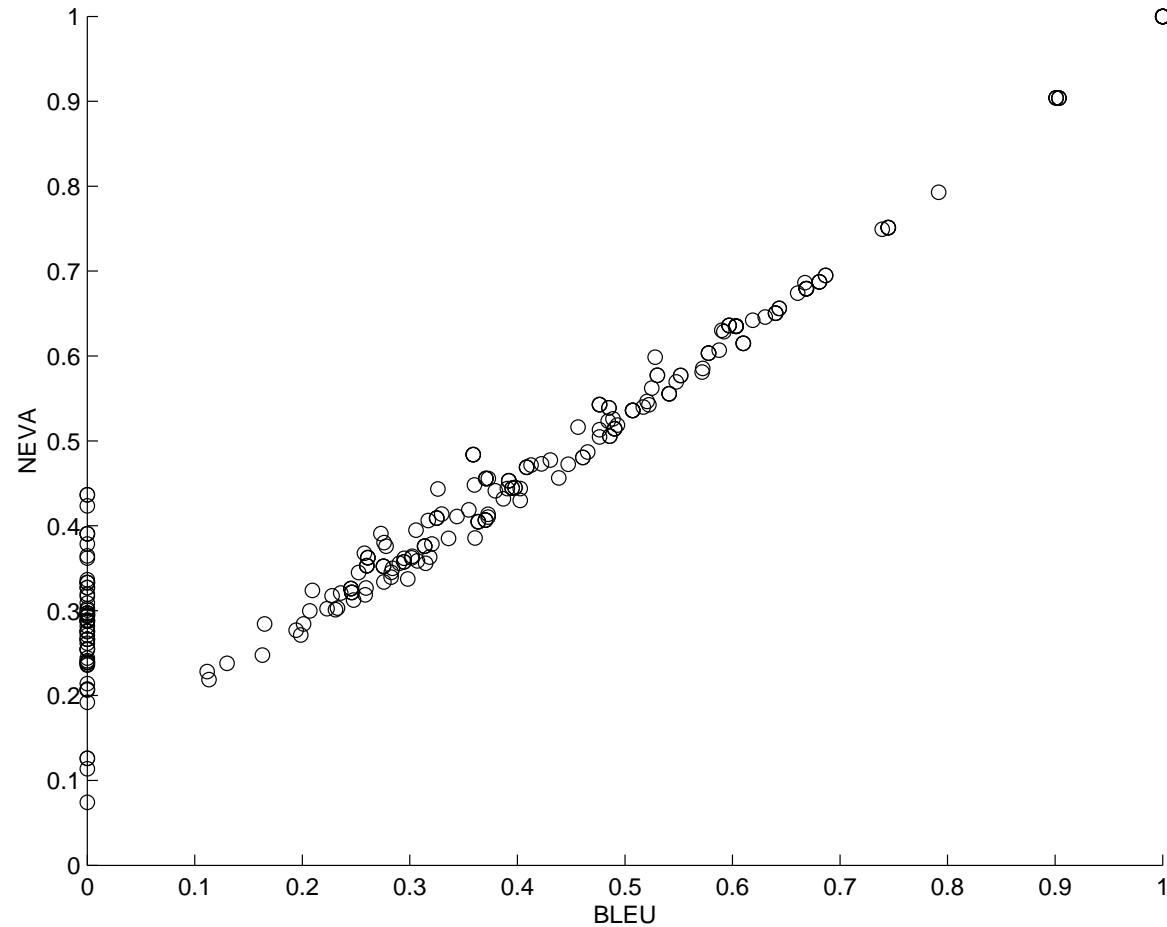
BLEU – Redefinition 2

N-gram EVALuation:

$$\text{NEVA} = \text{BP} \cdot \sum_{n=1}^N w_n p_n$$

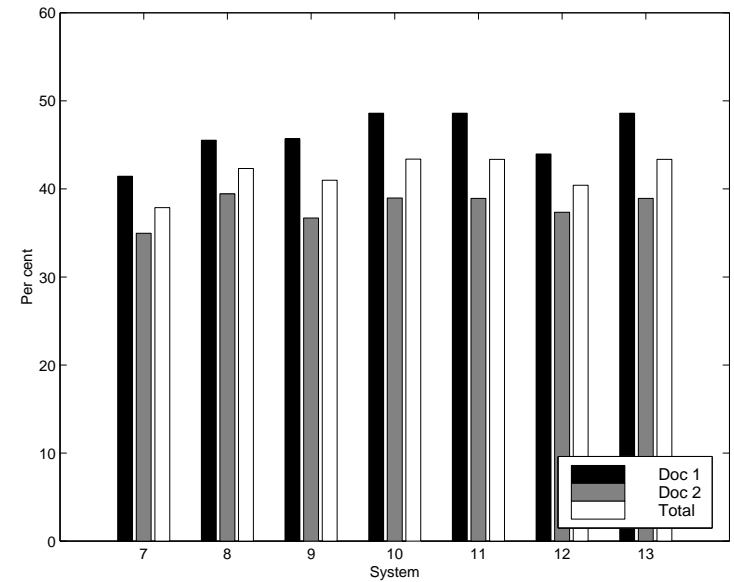
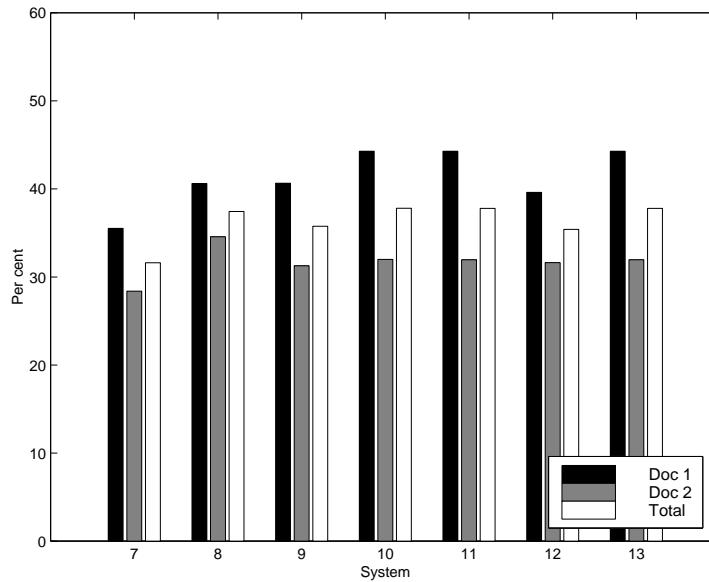


BLEU vs. NEVA for Segments (LREC 6/4)





BLEU vs. NEVA: Docs & Sys (LREC 6/4)



- Ranking is the same.
- NEVA yields slightly higher scores than BLEU.



NEVA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment.
- Work for various text types.
- Work with one reference translation.



NEVA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment. ✓
- Work for various text types.
- Work with one reference translation.



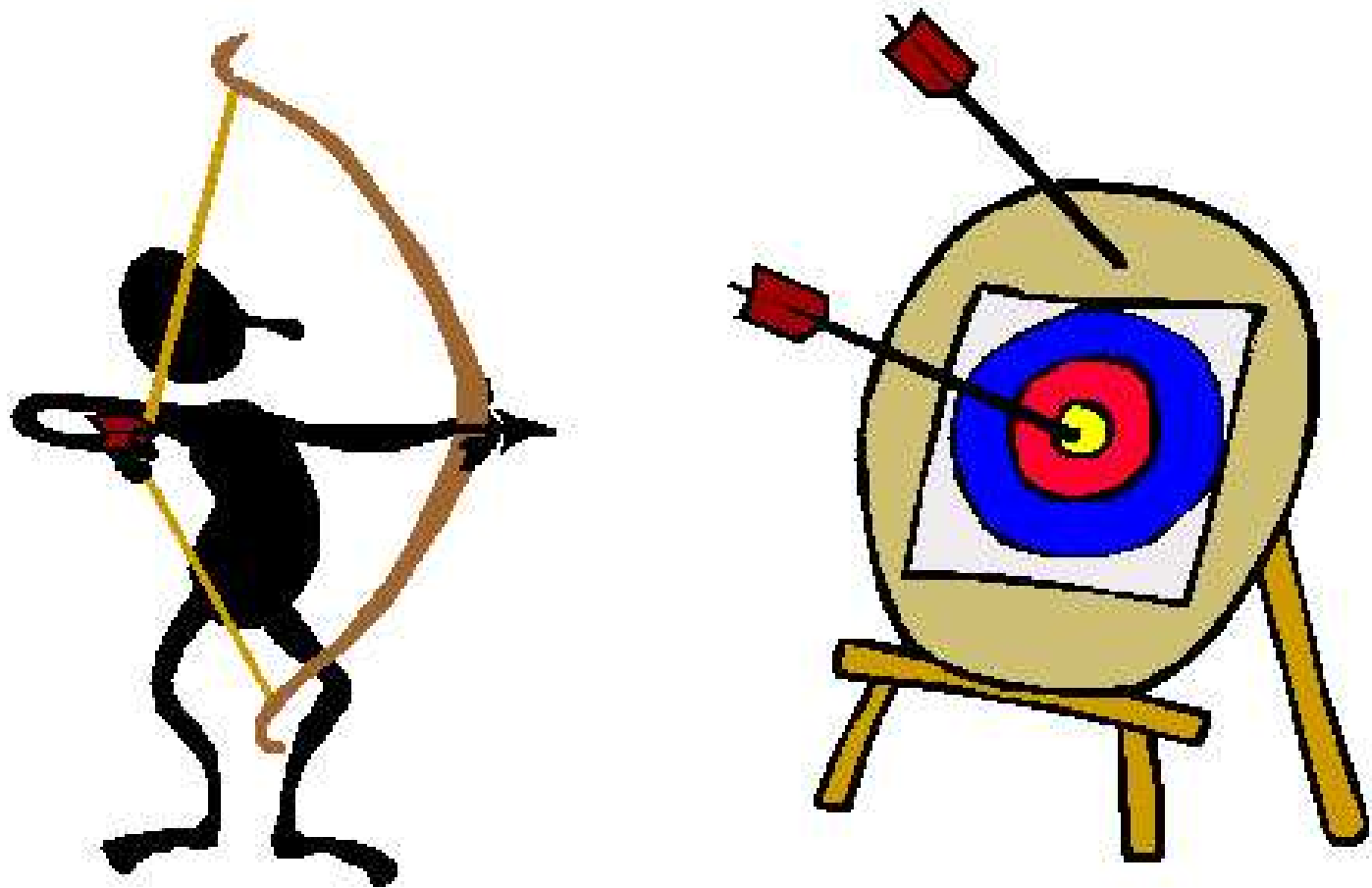
NEVA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels:
 - system, ✓
 - document, ✓
 - segment. ✓
- Work for various text types. ✓
- Work with one reference translation.



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One Reference Translation?





6/4 vs. 1 Reference Translation

- Ranking is basically the same (except for 8 and 9 on document 1 for WAFT).
- Scores are higher for 6/4 references (much higher for NEVA).
- Scoring levels for document 1 and 2 are reversed.

Level	WAFT	NEVA
System	0.8589	0.9857
Document 1	0.6854	0.9983
Document 2	0.9348	0.9632
Segment	0.6215	0.7274



NEVA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels: ✓
- Work for various text types. ✓
- Work with one reference translation.

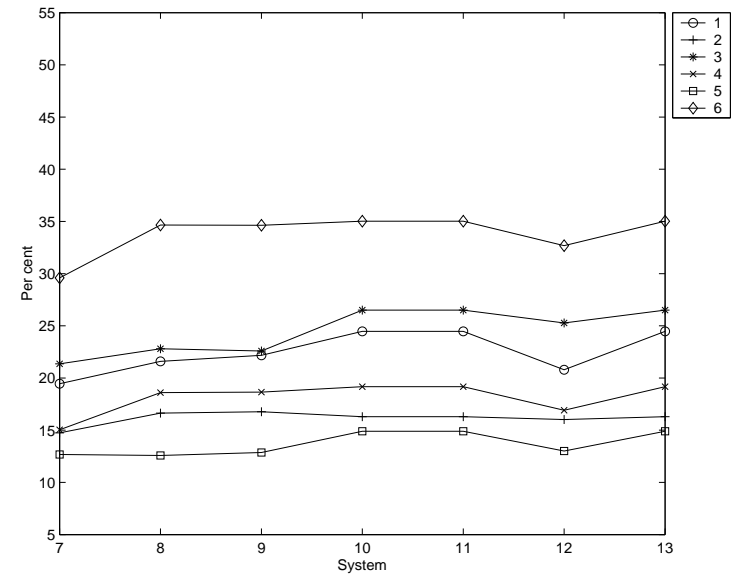
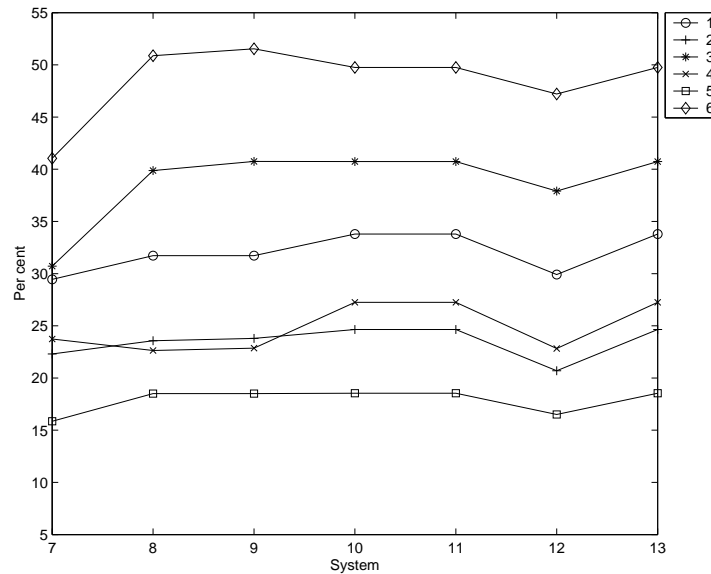


NEVA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels: ✓
- Work for various text types. ✓
- Work with one reference translation. ✓



Quality of Reference Translation



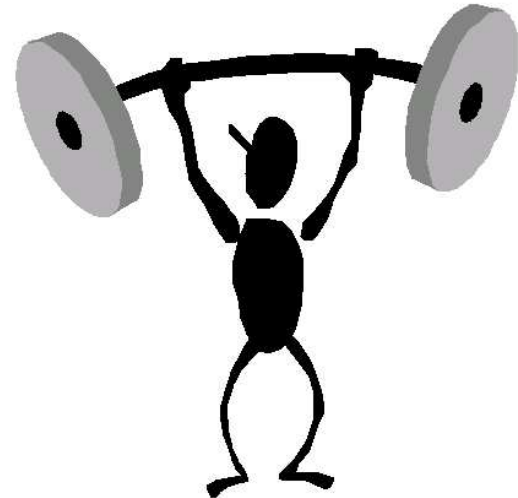
- Quality of reference translation matters for scoring.
- Quality of reference translation does not matter much for ranking.



The Super Model: Requirements

Free from

- errors (spelling, grammar, etc.);
- inconsistencies (variant forms, unwanted synonyms, etc.);
and
- interpretations, additions, deletions, etc.





The Super Model: Cloning

Original

1. Trouble shooting
2. The fluid is cleaned by passing through a filter.
3. Failure to follow this instruction can ...

Clone 1: Errors

1. Trouble shooting
2. The fluid is cleaned **via** a filter.
3. Failure to follow this instruction can ...

Clone 2: Inconsistencies

1. **Trou-
bleshooting**
2. The **oil** is cleaned via a filter.
3. Failure to follow this instruction can ...

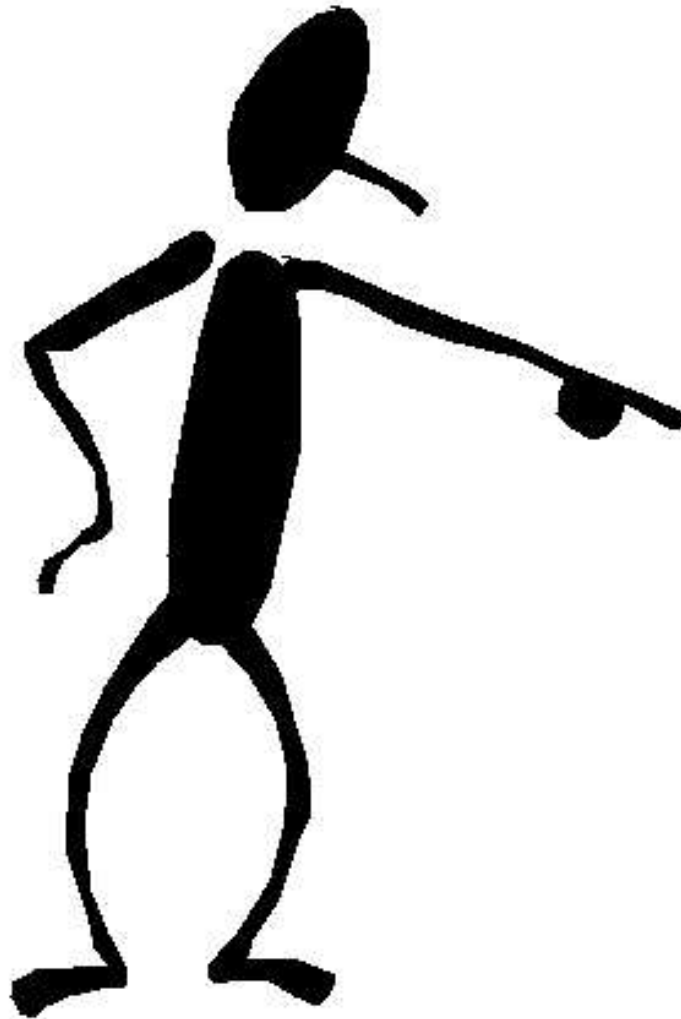
Clone 3: Interpretations

1. Trou-
bleshooting
2. The oil is cleaned via a filter.
3. **It** can ...



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Diagnostic Evaluation?





Diagnostic Evaluation: Edit Distance

Weakness:

- Sensitive to word order reversal.

Src: Cylinder, underdel

Cand: Bottom cylinder

Ref: Cylinder bottom

Advantages:

- Possibility to align edit operations, and to find
 - variant forms and synonyms (*clip/clamp*);
 - inflectional errors (*tensioner/tensioners*);
 - word errors (*in/into*);
 - differences in definiteness (*the*);
 - specifications or generalisations (nominal modifiers);



WAFT Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels: ✓
- Work for various text types. ✓
- Work with one reference translation. ✓



WAFT Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic. ✓
- Work at various levels: ✓
- Work for various text types. ✓
- Work with one reference translation. ✓



Diagnostic Evaluation: N-Gram Occurrence

Weakness:

- Sensitive to word errors (particularly mid-segment)

Src: Kontrollera backventilen.

Cand: Check the check valve.

Ref: Check the non-return valve.

Advantages:

- If something is right, it always yields a score above 0.
- It would be possible to report all n-grams not found.



NEVA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic.
- Work at various levels: ✓
- Work for various text types. ✓
- Work with one reference translation. ✓



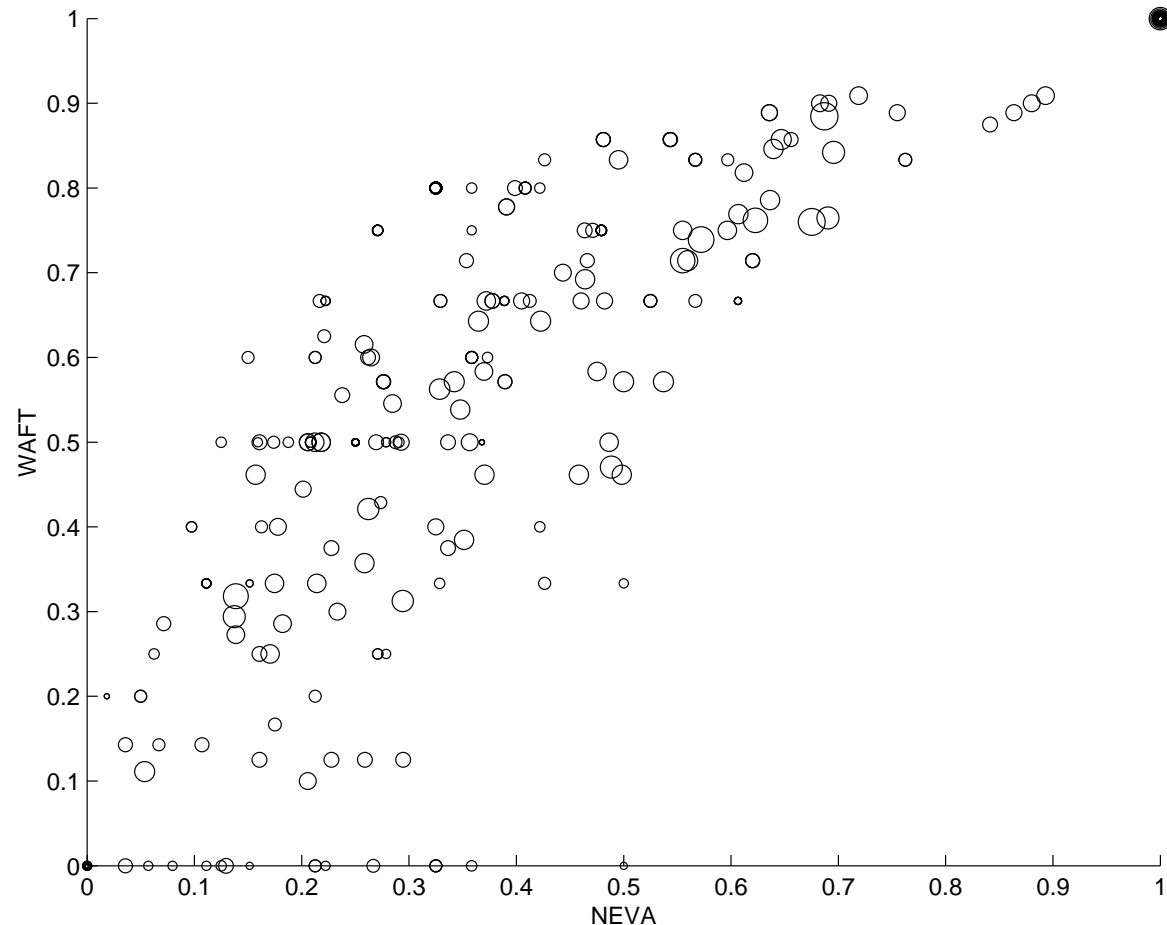
NEVA Scoring Card

- Work for various kinds of evaluations:
 - declarative, ✓
 - progressive, ✓
 - diagnostic. ✓
- Work at various levels: ✓
- Work for various text types. ✓
- Work with one reference translation. ✓



Diagnostic Evaluation: Correlation

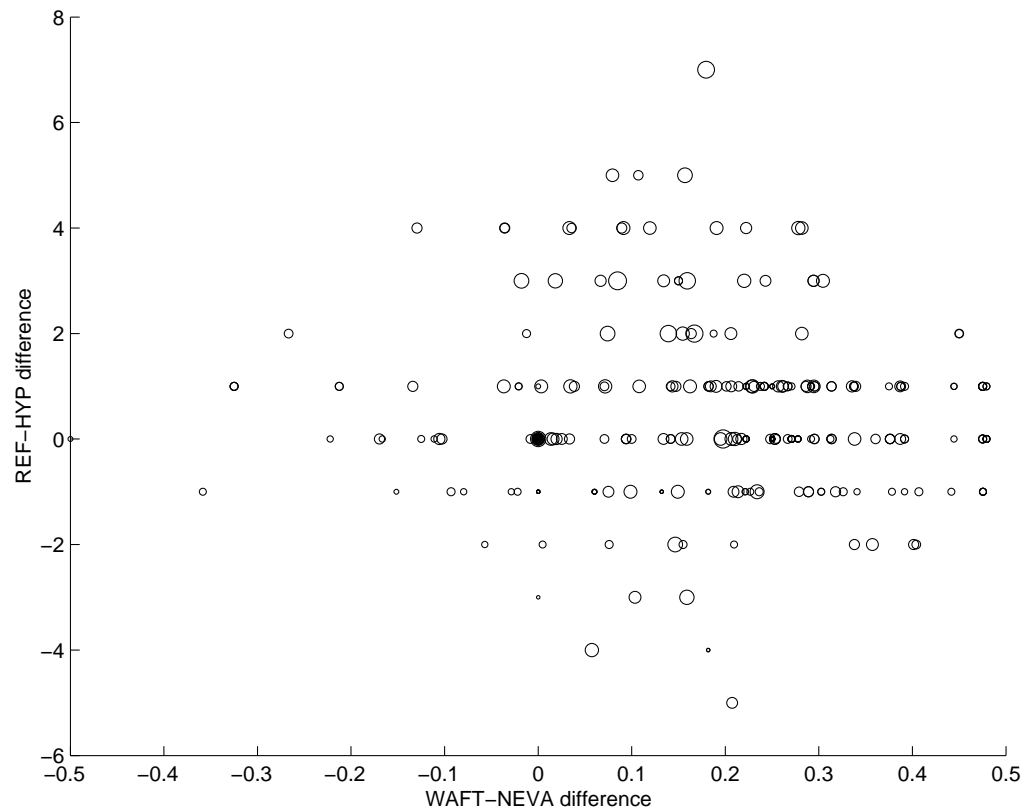
All segments in MATS where NEVA scores were greater than WAFT scores displayed a reversed word order problem:





Diagnostic Evaluation: Function?

Diagnostic score is possibly a function involving difference in WAFT and NEVA scores and difference in candidate and reference length.





Conclusions 1

- Edit distance and n-gram co-occurrence measures are applicable
 - for declarative, progressive, and diagnostic evaluations;
 - at the system, document, and segment level;
 - for news text and technical manuals; and
 - for use with a single reference translation.
- The existing measures (WA and BLEU) needed redefinition to be applicable at the segment level.
- The redefined measures (WAFT and NEVA) gave higher scores, but kept the ranking.
- The measures gave higher scores when used with several reference translations, but kept the ranking.



Conclusions 2

- WAFT gave higher scores than NEVA, except for major word-order reversals.
- NEVA was more sensitive to word-level errors.
- The differences could be used to single out certain error types in diagnostic evaluation.
- The differences could be used to find inconsistencies in a single reference translation.



References 1

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- Doddington. Automatic evaluation of machine translation quality using n-gram co-occurrence statistics. In *Proceedings of HLT 2002*, pp. 128–132, San Diego, USA, 2002.
- Papineni et al. BLEU: a method for automatic evaluation of machine translation. IBM RC22176 (W0109-022), IBM Research Division, T. J. Watson Research Center, 2001.



References 2

- Popescu-Belis. Meta-evaluation of evaluation metrics. tentative synthesis on the evaluation exercise. Talk presented at *Workshop on Machine Translation Evaluation: Human Evaluators Meet Automated Metrics (LREC'02)*, Las Palmas de Gran Canaria, Spain, 2002.
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