# An Awkward Disparity between BLEU / RIBES Scores and Human Judgements in Machine Translation

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# troduction

MT metrics criticized for various reasc

(Babych and Hartley, 2004; Smith et al. 2014; Graham et al. 2015)

## Hypothesis 1:

Appeared calm when he was taken to the American plane, to Miami , Florida.

# Hypothesis 2:

which will he was , when taken Appeared calm to the Ameri to Miami , Florida

**Reference:** Orejuela appeared calm as he was led to the American plane take him to Miami , Florida.

• Low BLEU != Bad MT (Callison-Burch et al. 2006)

• Higher BLEU -> Better MT (c.f. WMT, WAT, IWS

# BLEU: (Papineni et al. 2002)

- · Precision based
- · Weak recall penalty

## · Disregards order

### Source:

이러한작용을발휘하기위해서는, 각각 0.005% 이상함유하는것이바람직하다.

### Hypothesis:

このような作用を発揮するためには、<u>夫々</u> 0.005%以上含有することが好ましい。

### Baseline:

このような作用を発揮するために は、それぞれ 0. 005%以上含有す ることが好ましい。

### Reference:

このような作用を発揮させるためには、夫々 0.005%以上含有させることが好まし りょ

- · Crude ngram wei
- Over-sensitive to minor difference

<u>Hypothesis</u>		
<b>P</b> <sub>1</sub> : 90.0	P	
<b>P<sub>2</sub></b> : 78.9	Ρ	
<b>P<sub>3</sub></b> : 66.7	Ρ	
<b>P</b> <sub>4</sub> : 52.9	P	
<b>BP:</b> 0.905	B	
BLEU: 64.03	B	
HUMAN: -5	┠	

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	RIB	ES (Isozaki et al. )	2014)		
ons	• Kenda	ll Tau prior on unigram	· Adequacy not	measured	
)	·Overc	omes reordering	· Correlates with	n BLEU ( <i>natu</i>	r(
which will <b>Same BLEU?!</b> ican plane	Source: T용융(E 5℃/분∘ Hypothe Tme でDS( (T)。	OSC) = 89.9℃; T결정화(DSC) = 72℃( 에서DSC 로측정). esis: 1 t (DSC) = 7 2℃ (5℃/分 C 測定 (DSC) =89.9 結晶化度	Hypothesis RIBES: 94.04 BLEU: 53.3 HUMAN: -5		
e which will	Baseline T溶融 (DS( 定)。	Baseline: T溶融(DSC)=89.9℃;T結晶化 (DSC)=72℃(5℃/分でDSCで測 定)。			
/SLT, OpenMT)	Tmel yst Cを用い	t (DSC) =89.9℃;Tcr (DSC) =72℃ (5℃/分でDS いて測定)。	BLEU: 58.8 HUMAN: 0		
	Syst	em Setup + Resu	lts		
		Parameters	Organizers	Ours	
ights		Input document length Korean tokenizer Japanese tokenizer	40 MeCab Juman	80 KoNLPy MeCab	
<u>Baseline</u>		LM <i>n</i> -gram order Distortion limit	5 0	5 20	
<b>P</b> <sub>1</sub> : 84.2		Quantized & binarized LM	no	yes	
<b>a</b> : 66.7		devtest.txt in LM	no	yes	
		MEDT runs	no 1	yes	
'₂:4/.⊥		WIEKTTUIIS	1	2	

5 •₄:25.0 **3P:** 0.854 **3LEU: 43.29 HUMAN: 0** 

• Organizers:

 $\cdot$  **Ours** :

**Note:** This is our Unicode2String submission for KO->JA patent subtask in WAT 2015; the other results of other subtasks are presented in Tan and Bond (2014) and Tan et al. (2015)







# Conclusion

- Higher BLEU/RIBES correlates with +ve HUMAN, not -ve HUMAN
- · Minor lexical diff. cause huge diff. in BLEU; RIBES mostly measures fluency
- Minor metric score diff. not reflecting major translation inadequacy
- Higher BLEU != Better MT

rally)

# RIBES = 94.13; BLEU = 69.22; HUMAN = 0.0

# RIBES = 95.15; BLEU = 85.23; HUMAN = -17.75!!!

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# Marie Curie Actions

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