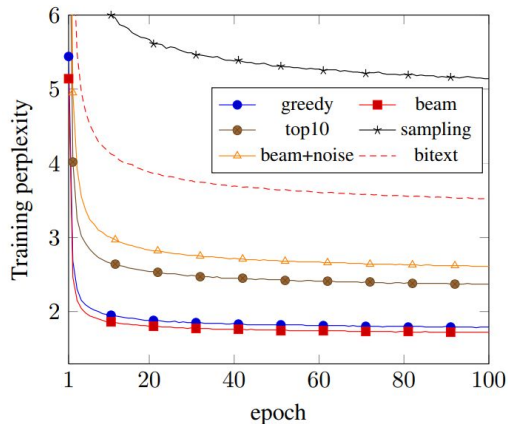
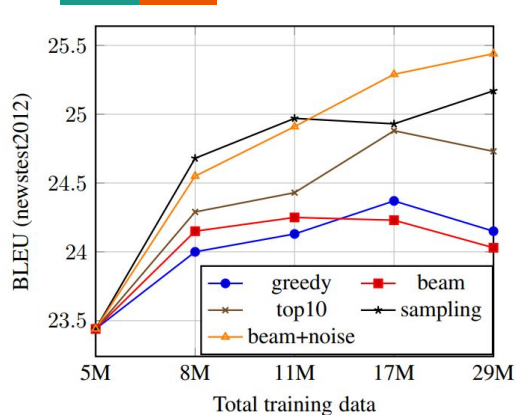
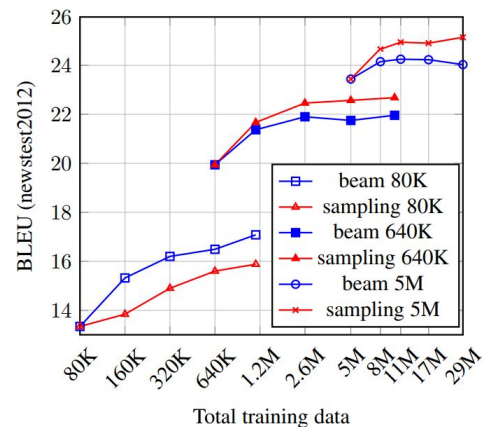


Understanding Back-Translation at Scale

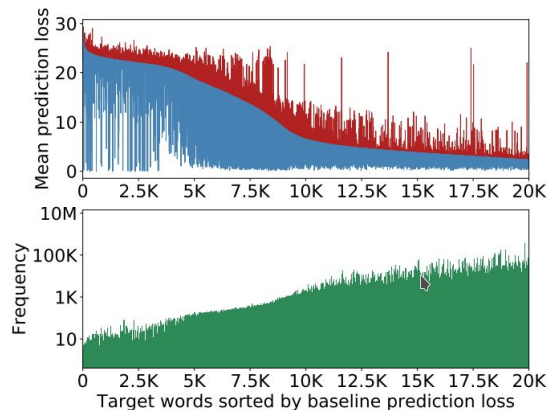


source	Diese gegenstzlichen Auffassungen von Fairness liegen nicht nur der politischen Debatte zugrunde.
reference	These competing principles of fairness underlie not only the political debate.
beam	These conflicting interpretations of fairness are not solely based on the political debate.
sample	<i>Mr President</i> , these contradictory interpretations of fairness are not based solely on the political debate.
top10	Those conflicting interpretations of fairness are not solely at the heart of the political debate.
beam+noise	conflicting BLANK interpretations BLANK are of not BLANK based on the political debate.

	En-De	En-Fr
a. Gehring et al. (2017)	25.2	40.5
b. Vaswani et al. (2017)	28.4	41.0
c. Ahmed et al. (2017)	28.9	41.4
d. Shaw et al. (2018)	29.2	41.5
DeepL	33.3	45.9
Our result	35.0	45.6
<i>detok. sacreBLEU³</i>	33.8	43.8



Back-Translation Sampling by Targeting Difficult Words in Neural Machine Translation



	Size	2014	2015	2016	2017
Baseline	4.5M	26.7	27.6	32.5	28.1
+ synthetic (1:1)	9M	28.7	29.7	36.3	30.8
+ synthetic (1:4)	23M	29.1	30.0	36.9	31.1
+ synthetic (1:10)	50M	22.8	23.6	29.2	23.9

System	De-En				En-De			
	test2014	test2015	test2016	test2017	test2014	test2015	test2016	test2017
BASLINE [†]	26.7	27.6	32.5	28.1	21.2	23.3	28.0	22.4
RANDOM [†]	28.7	29.7	36.3	30.8	24.0	26.0	30.7	24.8
Difficulty criterion	Context	Similarity						
FREQ	TOKENS	EMB		30.0	30.8	37.6	31.7	24.4
PREDLOSS	SWORDS	MATCH		29.1	30.1	36.9	31.0	26.3
PREDLOSS	TOKENS	MATCH		29.7	30.6	37.6	31.8	28.8
PREDLOSS	TOKENS	EMB		29.9	30.8	37.7	31.9	27.4
PREDLOSS	SENTENCE	EMB		24.9	25.5	30.1	26.2	24.3
MEANPREDLOSS	TOKENS	EMB		30.2	31.4	37.9	32.2	24.5
								27.2
								31.8
								25.6

	Size	2014	2015	2016	2017
Baseline	4.5M	21.2	23.3	28.0	22.4
+ synthetic tgt	9M	22.4	25.3	29.8	23.7
+ synthetic src	9M	24.0	26.0	30.7	24.8

	Size	2014	2015	2016	2017
Baseline	2.25M	24.3	24.9	29.5	25.6
+ synthetic	4.5M	26.0	26.9	32.2	27.5
+ ground truth	4.5M	26.7	27.6	32.5	28.1

Improving Neural Machine Translation Models with Monolingual Data

a lot of monodata

dataset	sentences
WMT _{parallel}	4 200 000
WIT _{parallel}	200 000
WMT _{mono_de}	160 000 000
WMT _{synth_de}	3 600 000
WMT _{mono_en}	118 000 000
WMT _{synth_en}	4 200 000
dataset	sentences
WIT	160 000
SETimes	160 000
Gigaword _{mono}	177 000 000
Gigaword _{synth}	3 200 000

quality comparison when +mono. +synthetic

		BLEU			
name		newstest2014		newstest2015	
		single	ens-4	single	ens-4
syntax-based (Sennrich and Haddow, 2015)		22.6	-	24.4	-
Neural MT (Jean et al., 2015b)		-	-	22.4	-
parallel	37m (parallel)	19.9	20.4	22.8	23.6
+monolingual	49m (parallel) / 49m (monolingual)	20.4	21.4	23.2	24.6
+synthetic	44m (parallel) / 36m (synthetic)	22.7	23.8	25.7	26.5
name		fine-tuning		BLEU	
		data	instances	tst2013	tst2014
				tst2013	tst2014
NMT (Luong and Manning, 2015) (single model)				29.4	-
NMT (Luong and Manning, 2015) (ensemble of 8)				31.4	27.6
1	parallel	-	-	25.2	22.6
2	+synthetic	-	-	26.5	23.5
3	2+WIT _{mono_de}	WMT _{parallel} / WIT _{mono}	200k/200k	26.6	23.6
4	2+WIT _{synth_de}	WIT _{synth}	200k	28.2	24.4
5	2+WIT _{parallel}	WIT	200k	30.4	25.9

6 BLEU difference in back-translation quality leads to a 0.6–0.7 BLEU difference in translation quality

back-translation	BLEU		
	DE→EN	EN→DE	
	2015	2014	2015
none	-	20.4	23.6
parallel (greedy)	22.3	23.2	26.0
parallel (beam 12)	25.0	23.8	26.5
synthetic (beam 12)	28.3	23.9	26.6
ensemble of 3	-	24.2	27.0
ensemble of 12	-	24.7	27.6

system	BLEU	
	WMT	IWSLT
parallel	20.1	21.5
+synthetic	20.8	21.6
PBSMT gain	+0.7	+0.1
NMT gain	+2.9	+1.2