Beam Decoding with Controlled Patience

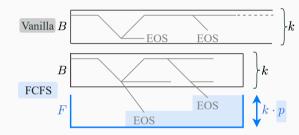
(Kasai et al., 2022)

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- 即https://arxiv.org/abs/2204.05424
- https://github.com/jungokasai/beam_with_patience

Introduction

Beam search is widely used, but the implementation has variations.

- Vanilla: TensorFlow Addons library
- first come, first served; FCFS: fairseq, Transformers
 - This paper proposes a patience factor on FCFS beam search.



Vanilla vs. FCFS with patience factor p:

k denotes the beam size. FCFS stores finished sentences in F, but they stay in (and later may fall off from) beam B during vanilla decoding.

Vanilla Beam Decoding

The top-k operation is applied over all sequences.

- Find the top-k sequences, including both finished and unfinished sequences: Line 5
- Decode until all top-k sequences are finished: Line 13
 - It tends to result in deeper search.

Vanilla Beam Decoding

```
k: beam size, M: maximum length,
          \mathcal{V}: Vocabulary, score(·): scoring function.
 1: B_0 \leftarrow \{\langle 0, \text{Bos} \rangle\}
 2: for t \in \{1, \dots, M-1\} :
 3: for \langle s, \mathbf{y} \rangle \in B_{t-1}:
              if \mathbf{y}.last() = Eos:
                   H.add(\langle s, \mathbf{y} \rangle)
                  continue
              for u \in \mathcal{V}:
                   s \leftarrow \operatorname{score}(\mathbf{v} \circ y), \ H.\operatorname{add}(\langle s, \mathbf{v} \circ y \rangle)
          B_t \leftarrow \varnothing
          while |B_t| < k: # Find top k from H.
10:
11:
              \langle s, \mathbf{y} \rangle \leftarrow H.\max(), B_t.\operatorname{add}(\langle s, \mathbf{y} \rangle)
              H.\text{remove}(\langle s, \mathbf{y} \rangle)
13:
          if \mathbf{y}.last() = Eos, \forall \mathbf{y} \in B_t: # All finished.
14:
                return B_t.max()
15: return B_t.max()
```

First Come, First Served (FCFS)

The top-k operation is applied over unfinished sequences.

- Collect finished sequences in a first come, first served manner and removes them from the beam: Line 11
- FCFS has a wider breadth since it collects *k* unfinished sequences at every step regardless of how many sequences are finished with the EOS symbol.
 - Terminate when a total of k finished sequences is found.

FCFS Beam Decoding with Controlled Patience

```
k: beam size, M: maximum length, \mathcal{V}: Vocabulary score(·): scoring function, p: patience factor.
```

- 2: for $t \in \{1, ..., M-1\}$: 3: $H \leftarrow \varnothing, F_t \leftarrow F_{t-1}$ 4: for $\langle s, y \rangle \in B_{t-1}$: # Expansion.
- 5: for $y \in \mathcal{V}$: 6: $s \leftarrow \text{score}(\mathbf{y} \circ y), \ H.\text{add}(\langle s, \mathbf{y} \circ y \rangle)$
- 7: $B_t \leftarrow \varnothing$ 8: **while** $|B_t| < k$: #Find top k w/o EOS from H.
- 9: $\langle s, \mathbf{y} \rangle \leftarrow H.\max()$ 10: **if** $\mathbf{y}.\operatorname{last}() = \operatorname{EOS}:$

1: $B_0 \leftarrow \{\langle 0, Bos \rangle\}, F_0 \leftarrow \emptyset$

- 10: **if** $\mathbf{y}.\text{last}() = \text{EOS}:$ 11: $F_t.\text{add}(\langle s, \mathbf{y} \rangle)$ # Finished hypotheses.
- 12: **else** B_t .add $(\langle s, \mathbf{y} \rangle)$
 - if $|F_t| \ge k \cdot p$: # Originally, p = 1.
- 14: **return** F_t .max() 15: H.remove($\langle s, \mathbf{y} \rangle$)

13:

16: **return** F_t .max()

Patience Factor for FCFS

Separate the stopping criterion from the search breadth.

- Beam size k in FCFS controls both the breadth and stopping criterion (i.e., depth) of search.
- Proposed patience factor relaxes this assumption: Line 13
 - It can be implemented by the one-line modification.

FCFS Beam Decoding with Controlled Patience

```
k: beam size, M: maximum length, \mathcal{V}: Vocabulary score(·): scoring function, p: patience factor.
```

```
1: B_0 \leftarrow \{\langle 0, Bos \rangle\}, F_0 \leftarrow \emptyset
 2: for t \in \{1, \dots, M-1\} :
 3: H \leftarrow \varnothing, F_t \leftarrow F_{t-1}
          for \langle s, \mathbf{y} \rangle \in B_{t-1}: # Expansion.
             for y \in \mathcal{V}:
                   s \leftarrow \operatorname{score}(\mathbf{v} \circ y), H.\operatorname{add}(\langle s, \mathbf{v} \circ y \rangle)
          B_t \leftarrow \varnothing
          while |B_t| < k: # Find top k w/o EOS from H.
              \langle s, \mathbf{v} \rangle \leftarrow H.\max()
              if \mathbf{y}.last() = EOS:
10:
11:
                   F_t.add(\langle s, \mathbf{y} \rangle)
                                                # Finished hypotheses.
              else B_t.add(\langle s, \mathbf{y} \rangle)
              if |F_t| > k \cdot p: # Originally, p = 1.
13:
14.
                   return F_t.max()
15:
               H.remove(\langle s, \mathbf{y} \rangle)
16: return F_t.max()
```

Main Results

- **p**atience factor: p=2 for MT, p=0.5 for summarization
- metric: COMET (XLM-RoBERTa) for MT, ROUGE for summarization

	WMT 2020/2021 Machine Translation (p=2)								Summarization $(p=0.5)$					
	EN↔DE		EN⇔JA		$EN \leftrightarrow PL$		EN↔ZH		CNNDM			XSUM		
Algorithm	\rightarrow	\leftarrow	\rightarrow	\leftarrow	\rightarrow	\leftarrow	\rightarrow	\leftarrow	R-2	R-3	R-L	R-2	R-3	R-L
Greedy	43.7	66.2	33.6	9.5	46.0	53.5	32.5	23.5	21.1	11.9	30.7	19.8	10.7	34.3
Vanilla	48.2	66.3	38.7	15.7	52.7	58.2	33.9	29.9	19.2	11.0	28.0	19.5	10.7	33.1
FCFS	47.9	66.2	38.0	15.0	52.1	58.1	33.7	29.6	20.4	11.6	30.3	20.4	11.4	34.4
FCFS w/ p	48.3	66.4	38.4	15.6	53.0	58.4	33.8	30.2	21.4	12.4	31.2	21.0	11.8	35.4

- In summarization, vanilla decoding performs worse than greedy.
 - It might be a reason why FCFS is used instead of vanilla in popular libraries.

Evaluate Translation by BLEU

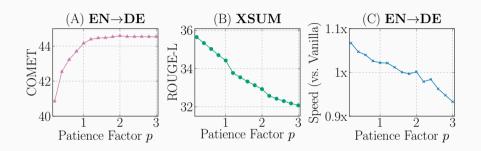
 \blacksquare patience factor: p=2

■ metric: BLEU

	WMT 2020 and 2021 Machine Translation (BLEU)										
	EN↔DE		EN∻	→ JA	EN∻	→PL	$EN \leftrightarrow ZH$				
Algorithm	\rightarrow	\leftarrow	\rightarrow	\leftarrow	\rightarrow	\leftarrow	\rightarrow	\leftarrow			
Greedy	42.9	46.6	20.2	17.4	19.8	30.7	31.2	21.7			
Vanilla	45.1	48.4	21.6	19.7	21.1	32.5	32.5	23.6			
FCFS	45.0	48.4	21.3	19.5	21.0	32.4	32.6	23.4			
FCFS w/ p	45.0	48.5	21.7	19.8	21.1	32.5	32.3	23.7			

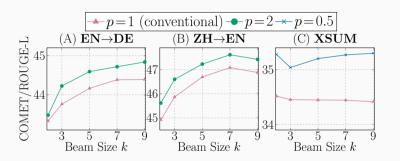
■ In many cases, FCFS w/p slightly better than FCFS.

Analysis: Effects of varying patience factors p



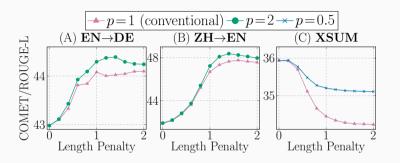
- The translation performance improves with larger patience factors with diminishing gains.
- Summarization benefits more from patience factors smaller than the original value of 1.
 - The nature of the summarization task that aims to generate concise text.

Analysis: Effects of controlled patience over varying beam sizes



■ The amount of improvement changes, but the patience factor is generally beneficial.

Analysis: Effects of controlled patience over varying length penalty



■ The amount of improvement changes, but the patience factor is generally beneficial.

[BTW] Length Penalty

■ Length-normalized log probability:

$$\frac{\log(P(Y|X))}{\text{lenpen}(Y)}$$

- Google NMT: $\operatorname{lenpen}(Y) = \left(\frac{5+|Y|}{5+1}\right)^{\alpha}$, $\alpha = 0.6$
- Fairseq: lenpen $(Y) = |Y|^{\alpha}$, $\alpha = 1.0$
 - ► This paper uses the fairseq version's length penalty.

Related Work

Stopping Criterion for Beam Decoding

- Optimal Beam Search (Huang et al., EMNLP 2017)¹
 - They propose a method to optimally finish beam search.
 - Modify the beam decoding procedure

Breadth of Beam Decoding

■ Analyzing effects of the search breadth (Ott et al., ICML 2018)²

^{1...}When to Finish? Optimal Beam Search for Neural Text Generation (modulo beam size)", Huang et al., 2017.

^{2.} Analyzing Uncertainty in Neural Machine Translation", Ott et al., 2018.

Conclusion

- This paper named two major beam search implementation, vanilla and FCFS.
- This paper proposed patience factor for FCFS, controls the stopping criterion.
 - Experiments shows that the proposed FCFS w/patience gains better translation/summarization performance than FCFS.

