

The Evolved Transformer

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Can we apply Neural Architecture Search to feedforward sequence models?



- Evolution
 - simple
 - works well in vision domain



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 - Progressive Dynamic Hurdles (PDH): discard bad models for cheap



Evolved Transformer Performance





| Model | Embedding Size | BLEU | Δ BLEU |
|-------------|-------------------|-----------------------|-----------------|
| Transformer | 128 | 21.3 ± 0.1 | 10 4 |
| ET | 128 | $\textbf{22.0}\pm0.1$ | + 0.7 |
| Transformer | 432 | 27.3 ± 0.1 | |
| ET | 432 | $\textbf{27.7}\pm0.1$ | + 0.4 |
| Transformer | 512 | 27.7 ± 0.1 | - |
| ET | 512 | $\textbf{28.2}\pm0.1$ | + 0.5 |
| Transformer | 768 | 28.5 ± 0.1 | - |
| ET | 768 | $\textbf{28.9}\pm0.1$ | + 0.4 |
| Transformer | 1024 | 28.8 ± 0.2 | - |
| ET | 1024 | $\textbf{29.0}\pm0.1$ | + 0.2 |

Evolved Transformer Performance



• State of the Art on WMT En-De

| Work | Model | Params | BLEU | SacreBLEU (Post, 2018) |
|-----------------------|---------------------------------------|--------|------|---------------------------|
| Gehring et al. (2017) | Convolutional Seq2Seq | 216M | 25.2 | - |
| Vaswani et al. (2017) | Transformer | 213M | 28.4 | - |
| Ahmed et al. (2017) | Weighted Transformer | 213M | 28.9 | - |
| Chen et al. (2018) | RNMT+ | 379M | 28.5 | - |
| Shaw et al. (2018) | Relative Attention Transformer | 213M | 29.2 | - |
| Ott et al. (2018) | Transformer | 210M | 29.3 | 28.6 |
| Wu et al. (2019) | Dynamic Lightweight Convolution | 213M | 29.7 | - |
| | Evolved Transformer | 218M | 29.8 | 29.2 |

• Generalizes to Other Tasks

| TASK | Size | TRAN PERP | ET Perp | TRAN BLEU | ET BLEU |
|------------------------------|-------------|---|---|---|---|
| WMT'14 EN-FR WMT'14 En-Fr | Base Big | $\begin{array}{c} 3.61 \pm 0.01 \\ 3.26 \pm 0.01 \end{array}$ | $\begin{array}{c} {\bf 3.42} \pm 0.01 \\ {\bf 3.13} \pm 0.01 \end{array}$ | $\begin{array}{c} 40.0 \pm 0.1 \\ 41.2 \pm 0.1 \end{array}$ | $\begin{array}{c} \textbf{40.6} \pm 0.1 \\ \textbf{41.3} \pm 0.1 \end{array}$ |
| WMT'14 En-Cs WMT'14 En-Cs | Base Big | $\begin{array}{c} 4.98 \pm 0.04 \\ 4.43 \pm 0.01 \end{array}$ | $\begin{array}{c} \textbf{4.42} \pm 0.01 \\ \textbf{4.38} \pm 0.03 \end{array}$ | $\begin{array}{c} 27.0 \pm 0.1 \\ 28.1 \pm 0.1 \end{array}$ | $\begin{array}{c} \textbf{27.6} \pm 0.2 \\ \textbf{28.2} \pm 0.1 \end{array}$ |
| LM1B | BIG | 30.44 ± 0.04 | $\textbf{28.60} \pm 0.03$ | - | - |

Architecture Comparison



Conv 1x1 : 512 RELU Conv 1x1 : 2048 Layer Norm 8 Head Attend to Encoder : 512 Layer Norm 8 Head Self Attention : 512 Layer Norm Conv 1x1 : 512 RELU Conv 1x1 : 2048 Layer Norm 8 Head Attend to Encoder : 512 Layer Norm 8 Head Self Attention : 512 Layer Norm

Transformer

Architecture Comparison





Evolved Transformer

Architecture Comparison





Evolved Transformer

Summary



- First work applying NAS on feedforward sequence model.
- Discovered the Evolved Transformer, which shows better efficiency.
- Open sourced in Tensor2Tensor.



Scan to see the paper and code.

Poster: Pacific Ballroom 6:30 pm to 9:00 pm