## **Summary (extended)**

The Transformer uses multi-head attention in three different ways. In a self-attention layer all of the keys, values and queries come from the same place. On the WMT 2014 English-to-French translation task, our model establishes a new single-model state-of-the-art BLEU score of 41.0 after training for 3.5 days on eight GPUs. The Transformer allows for signi acantly more parallelization and can reach a new state of the art in machine translation quality. In this work, we presented the Transformer, the arst sequence transduction model based entirely on self-attention. We replace the recurrent layers most commonly used in encoder-decoder architectures with multi-headed self-Attention. In row (E) we replace our sinusoidal positional encoding with learned positional embeddings. We measure the change in performance on English-to-German translation on the development set, newstest2013. The Transformer can be trained signi acantly faster than architectures based on recurrent or convolutional layers. The Transformer is the ■rst transduction model relying on self-attention to compute representations of its input and output without using sequence-aligned RNNs or convolution. At each step the model is auto-regressive[9], consuming the previously generated symbols as additional input when generating the next. The encoder is composed of a stack of N= 6 identical layers. In the embedding layers, we multiply those weights bypdmodel. ing different positions of a single sequence in order to compute a representation.