

Self-attention

https://arxiv.org/abs/1706.03762



q: query (to match others)

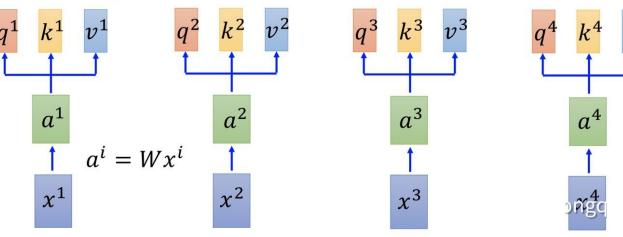
$$q^i = W^q a^i$$

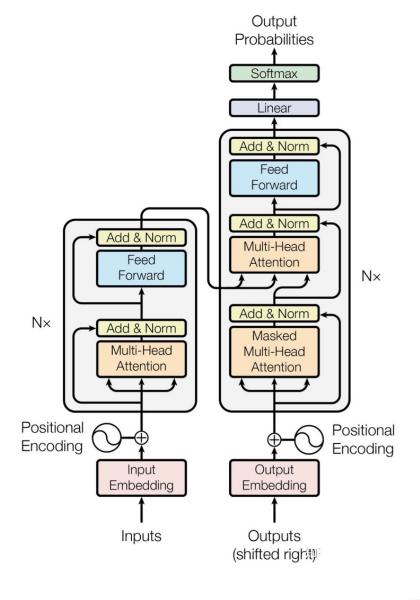
k: key (to be matched)

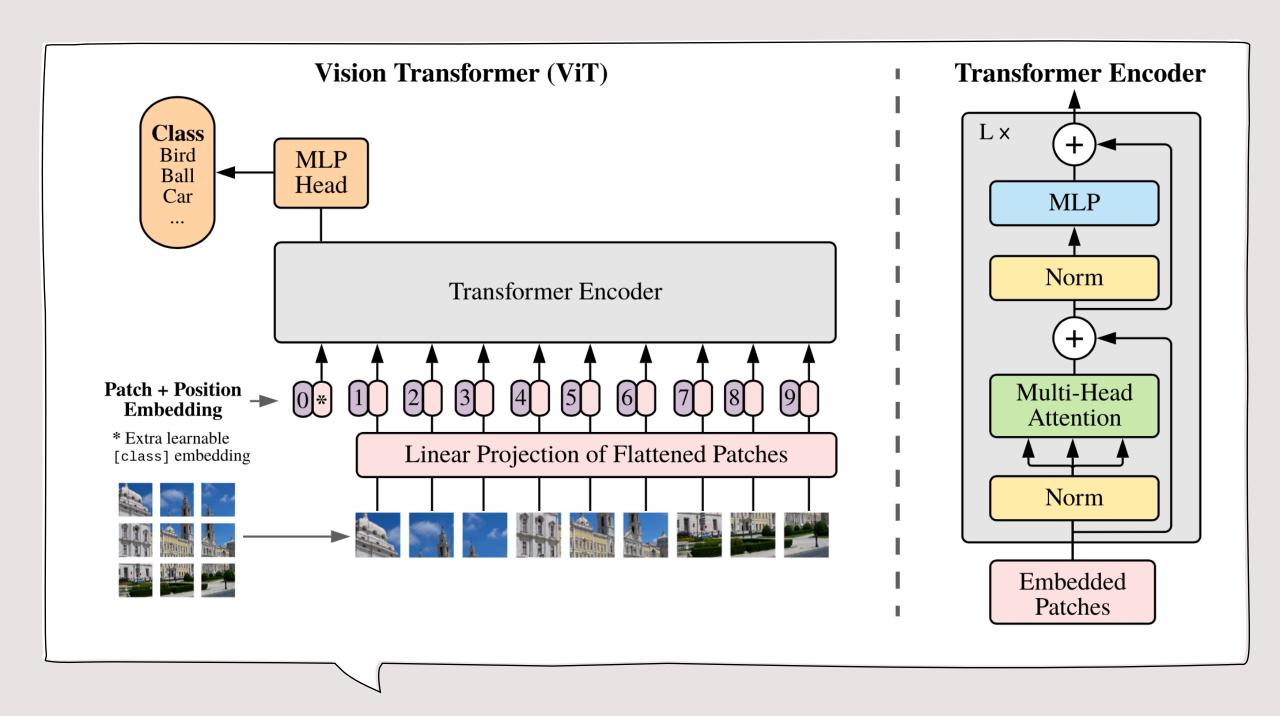
$$k^i = W^k a^i$$

v: information to be extracted

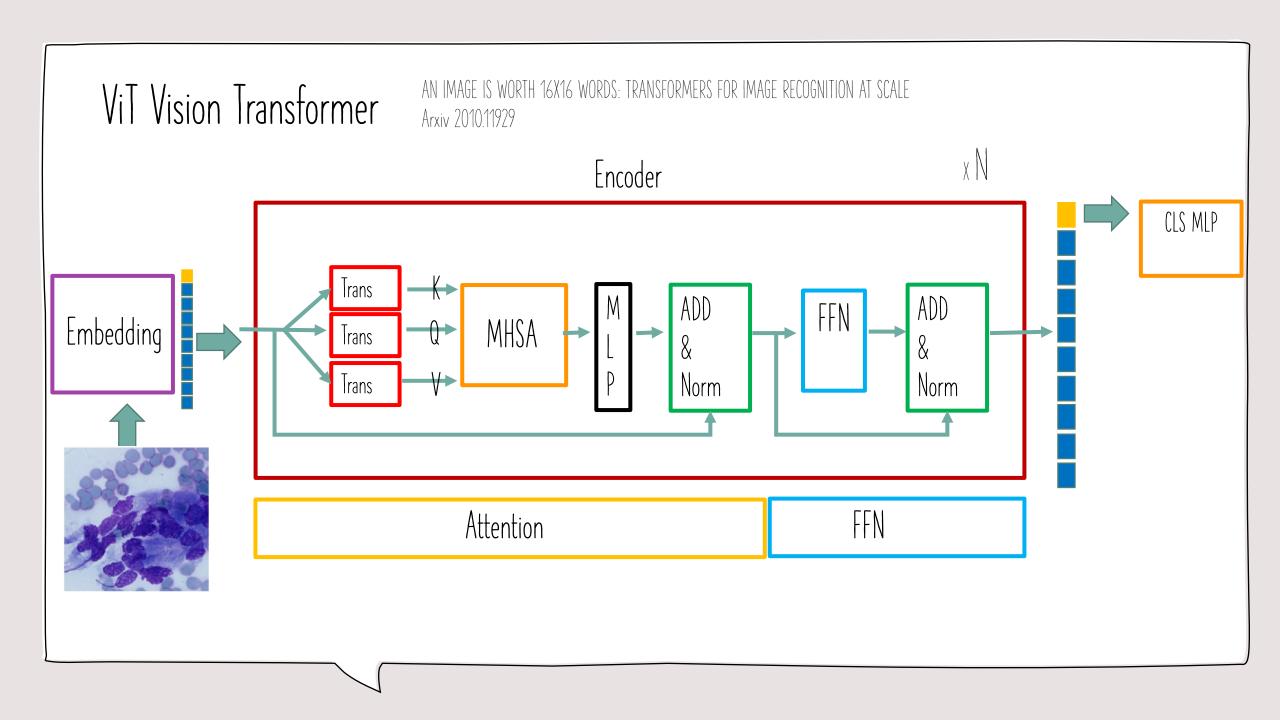
$$v^i = W^v a^i$$





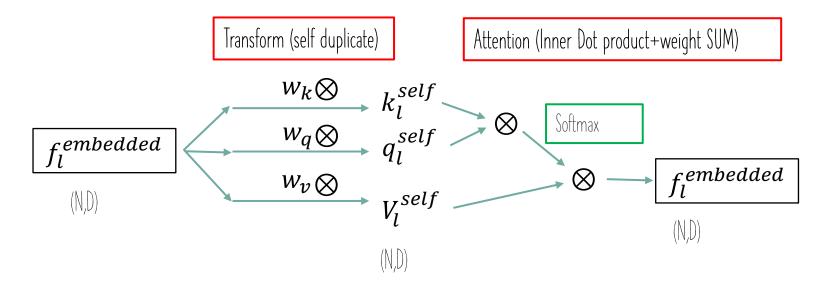






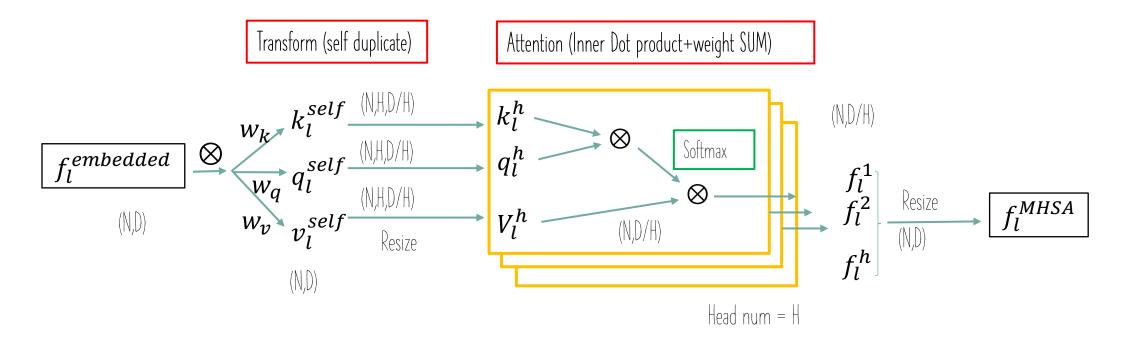
CLS token Patch embedding (1,768) Zeros Use CNN to split patches (patch_num +1 Embed_dim) Concatenate + positional encoding ReArrange CNN projection zeros : learnable posional (Batch, 3, 224, 224) (Batch, encoding Embed_dim, Patch_num_col, Kernal size = Stride Patch_num_row) Input Chanel = 3 (Batch, (Batch, (Batch, Output Chanel = Embed_dim patch_num, patch_num +1, patch_num +1, Embed_dim) Embed_dim) Embed_dim) (8, 196, 768) (8, 197, 768) (8, 197, 768)

SA self attention



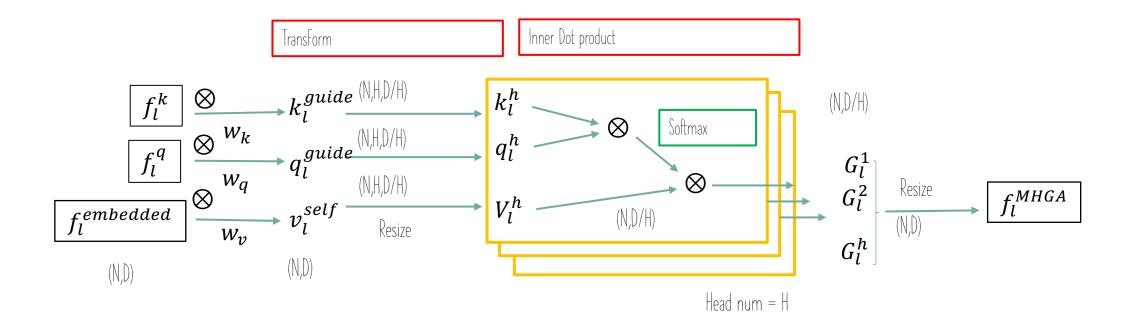
For a given embedded patch $f_l^{embedded}$, $f_l^{MHSA} = SoftMax(q_l^{self} \cdot k_l^{self})^T \cdot V_l^{self}$, where $q_l^{self} = w_q \cdot f_l^{embedded}$, $k_l^{self} = w_k \cdot f_l^{embedded}$, $V_l^{self} = w_v \cdot f_l^{embedded}$.

MHSA multi-head self attention

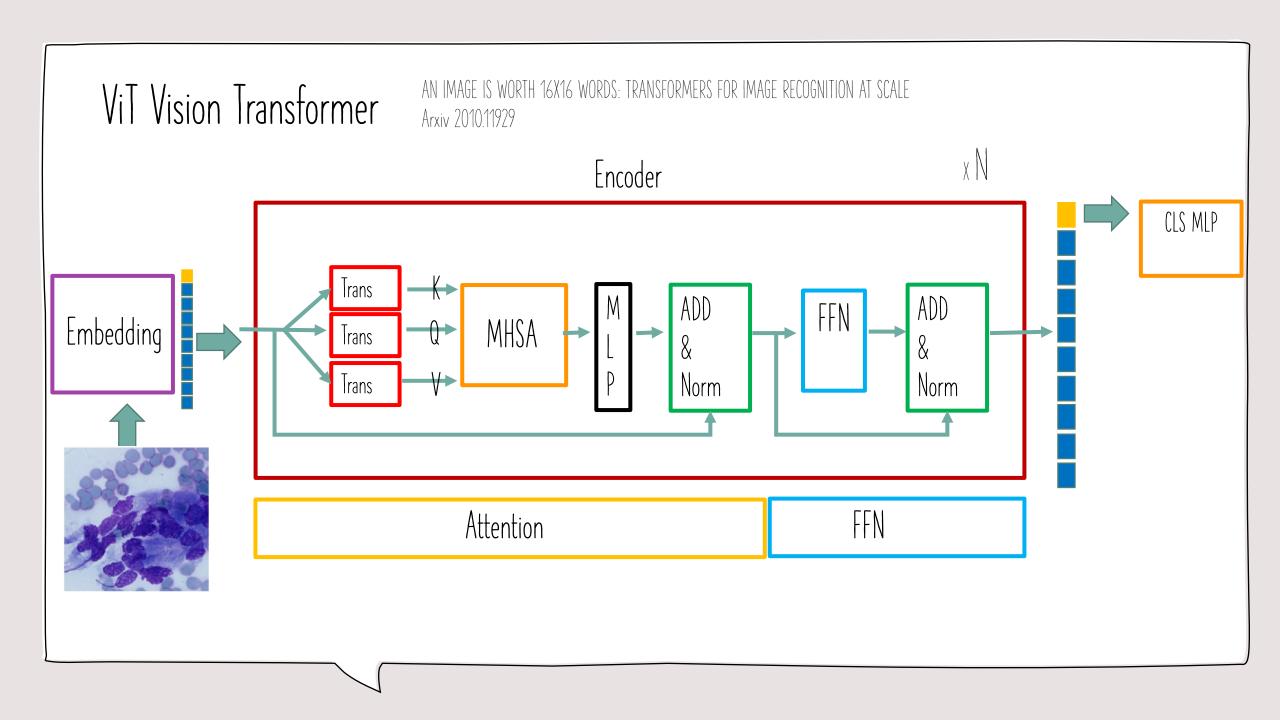


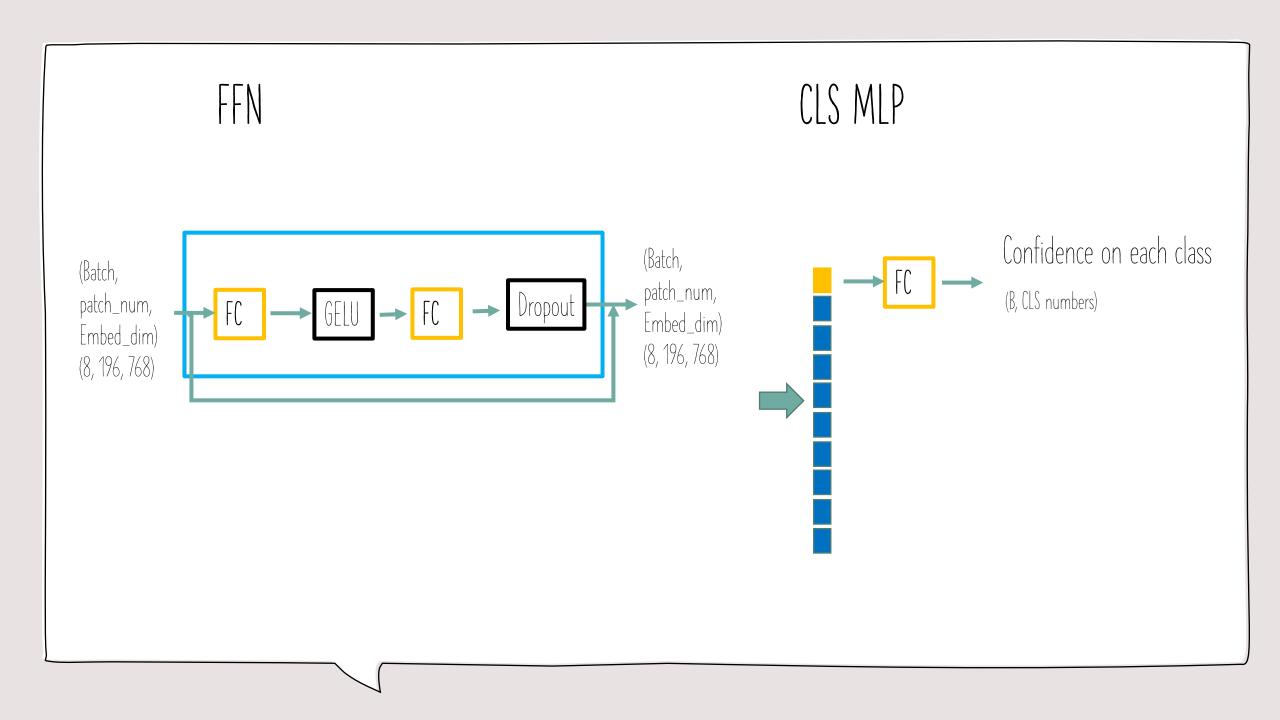
For a given embedded patch in each head $f_l^{embedded}$, $f_l^{MHSA} = SoftMax(q_l^{self^T} \cdot k_l^{self})^T \cdot V_l^{self}$, where $q_l^{self} = w_q \cdot f_l^{embedded}$, $k_l^{self} = w_k \cdot f_l^{embedded}$, $V_l^{self} = w_v \cdot f_l^{embedded}$.

MHGA multi-head quided attention



For a given embedded patch in each head $f_l^{embedded}$, $f_l^{MHSA} = SoftMax(q_l^{self} \cdot k_l^{self})^T \cdot V_l^{self}$, where $q_l^{self} = w_q \cdot f_l^{embedded}$, $k_l^{self} = w_k \cdot f_l^{embedded}$, $V_l^{self} = w_v \cdot f_l^{embedded}$.





POINTS TO DECLEAR

- 1. Who's learning the most of the hard works?
- 2. Who's the heavyest?
- 3. Why its sooo stirring?
- 4. A slight view of the future ? (Nay, actually my works lol)

Hard Math

HOW DO VISION TRANSFORMERS WORK?

Arxiv 2202.06709

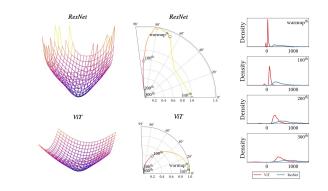
Inspiring Experiments

NeurPS2021 Intriguing Properties of Vision Transformers
Arxiv 2105.10497

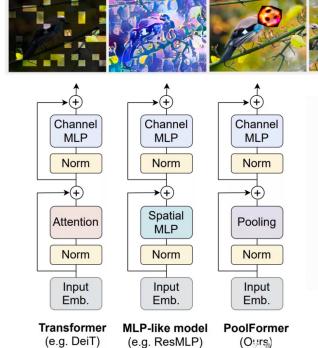
Inspiring Model play

MetaFormer is Actually What You Need for Vision Poolformer Arxiv 2111.11418

Patches Are All You Need https://openreview.net/pdf?id=TVHS5Y4dNvM



(d) Permutation



(b) Distribution Shift (c) Adversarial Patch

(a) Occlusion

Under review as a conference paper at ICLR 2022

Convolutions Attention MLPs
Patches Are All You Need?

(e) Auto-Segment (f) Off-the-shelf Feats.

Anonymous authors
Paper under double-blind review



Model structure

Learning strategy

Initiallization

https://github.com/sagizty/Multi-Stage-Hybrid-Transformer

Arxiv: 2112.13513

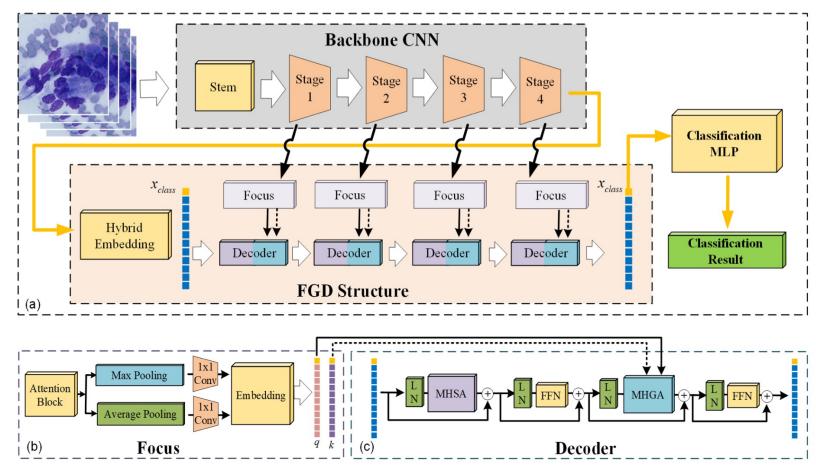


Fig. 1. The architecture of the proposed Multi-stage Hybrid Transformer (MSHT) model for the classification of ROSE images of pancreatic cancer. (a) The architecture of MSHT. (b) The focus block of the FGD structure (c) The decoder of the FGD structure. MHSA denotes multi-head self-attention, MHGA denotes multi-head guided-attention, LN denotes layer norm block, FFN denotes the feed-forward network, and MLP denotes multi-layer perceptron.

Thanks

PPT: https://github.com/sagizty/Insight/blob/main/Dive%20into%20Transformer.pdf Code: https://github.com/sagizty/Insight/blob/main/ZTY_Torch/ViT.py