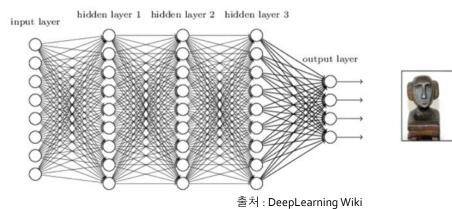
Vision Transformer(ViT)

발표자: 문경환

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About CNN



Translation Invariance

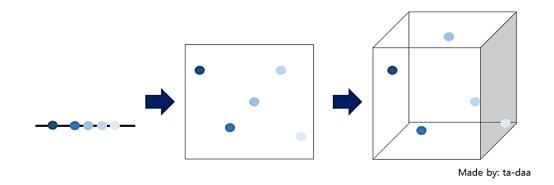




출처 : 데이콘

DNN 모든 퍼셉트론이 연결되어 있는 구조

특징 이미지의학습과정에서이미지가이동하면예측정확도가좋지않음.

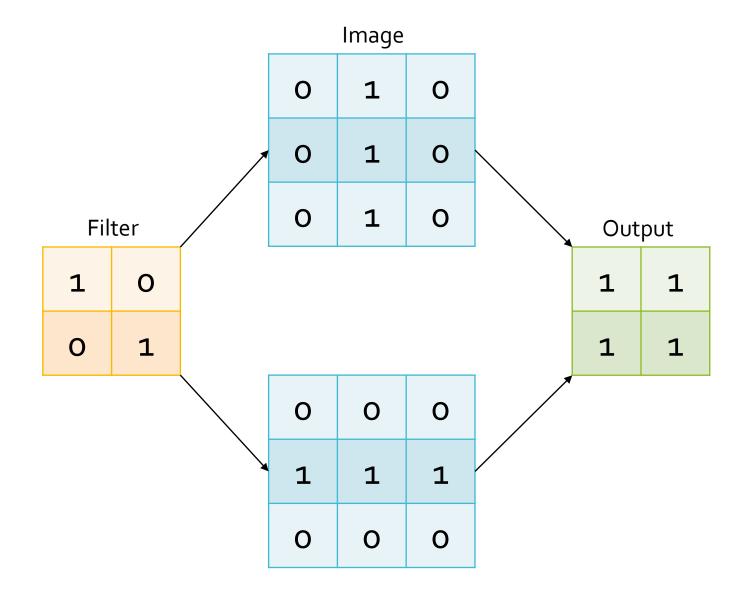


차원의 저주

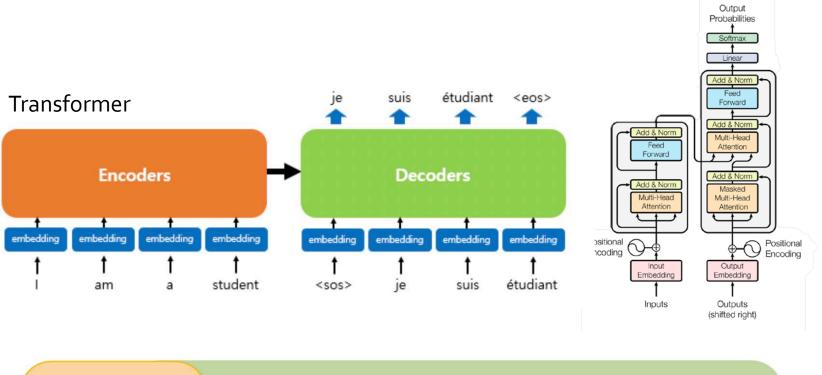
차원이클수록예측이어려우며, DNN의문제점을개선할방안이필요하였음.



About CNN



About ViT

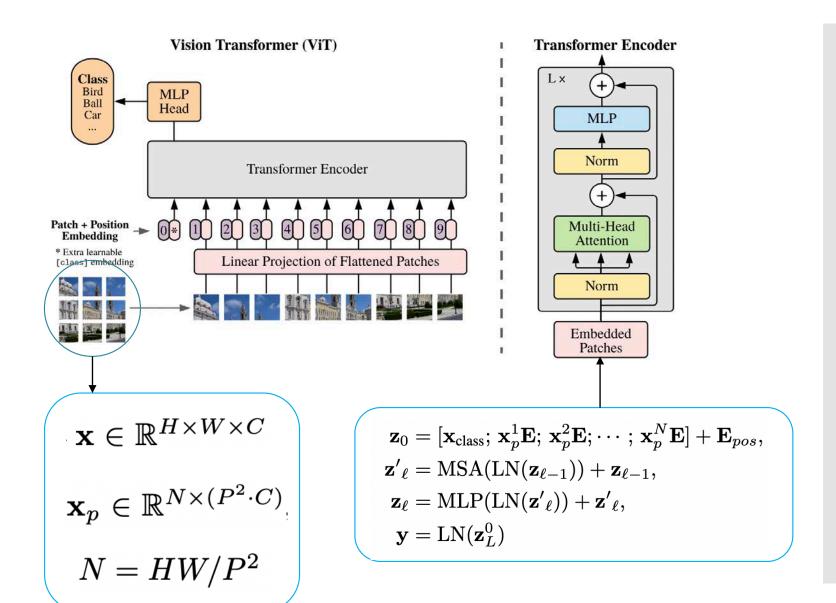


Transformer 자연어 분야에서 문장 번역을 위해 고안된 모델.

특징

학습 데이터의 양이 증가할수록 모델의 성능 개선

About ViT



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Attention

Input Attention







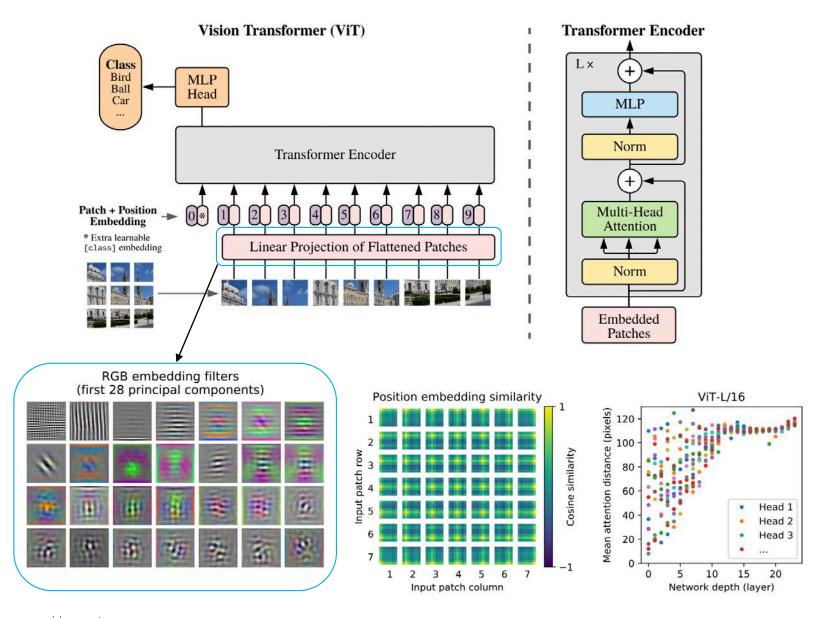






- A learned Position embedding is added to the patch representation.
- Closer patches tend to have more simmilar position embeddings.
- Patches in the same row/column have simmilar embeddings.
- This allows VIT to integrate information across the entire image even in the lowest layers

Self-Supervision



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Hyperparameters

Models	Dataset	Epochs	Base LR	LR decay	Weight decay	Dropout
ViT-B/{16,32}	JFT-300M	7	$8 \cdot 10^{-4}$	linear	0.1	0.0
ViT-L/32	JFT-300M	7	$6\cdot 10^{-4}$	linear	0.1	0.0
ViT-L/16	JFT-300M	7/14	$4 \cdot 10^{-4}$	linear	0.1	0.0
ViT-H/14	JFT-300M	14	$3 \cdot 10^{-4}$	linear	0.1	0.0
$R50x\{1,2\}$	JFT-300M	7	10^{-3}	linear	0.1	0.0
R101x1	JFT-300M	7	$8 \cdot 10^{-4}$	linear	0.1	0.0
$R152x\{1,2\}$	JFT-300M	7	$6 \cdot 10^{-4}$	linear	0.1	0.0
R50+ViT-B/{16,32}	JFT-300M	7	$8 \cdot 10^{-4}$	linear	0.1	0.0
R50+ViT-L/32	JFT-300M	7	$2\cdot 10^{-4}$	linear	0.1	0.0
R50+ViT-L/16	JFT-300M	7/14	$4 \cdot 10^{-4}$	linear	0.1	0.0
ViT-B/{16,32}	ImageNet-21k	90	10^{-3}	linear	0.03	0.1
ViT-L/{16,32}	ImageNet-21k	30/90	10^{-3}	linear	0.03	0.1
ViT-*	ImageNet	300	$3 \cdot 10^{-3}$	cosine	0.3	0.1

Dataset	Steps	Base LR
ImageNet	20 000	{0.003, 0.01, 0.03, 0.06}
CIFAR100	10 000	$\{0.001, 0.003, 0.01, 0.03\}$
CIFAR10	10000	$\{0.001, 0.003, 0.01, 0.03\}$
Oxford-IIIT Pets	500	$\{0.001, 0.003, 0.01, 0.03\}$
Oxford Flowers-102	500	$\{0.001, 0.003, 0.01, 0.03\}$
VTAB (19 tasks)	2 500	0.01

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