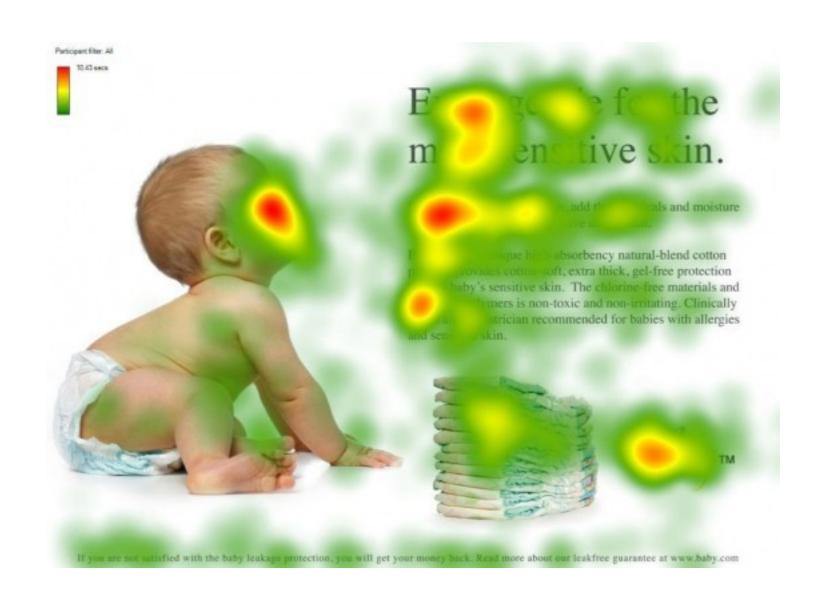
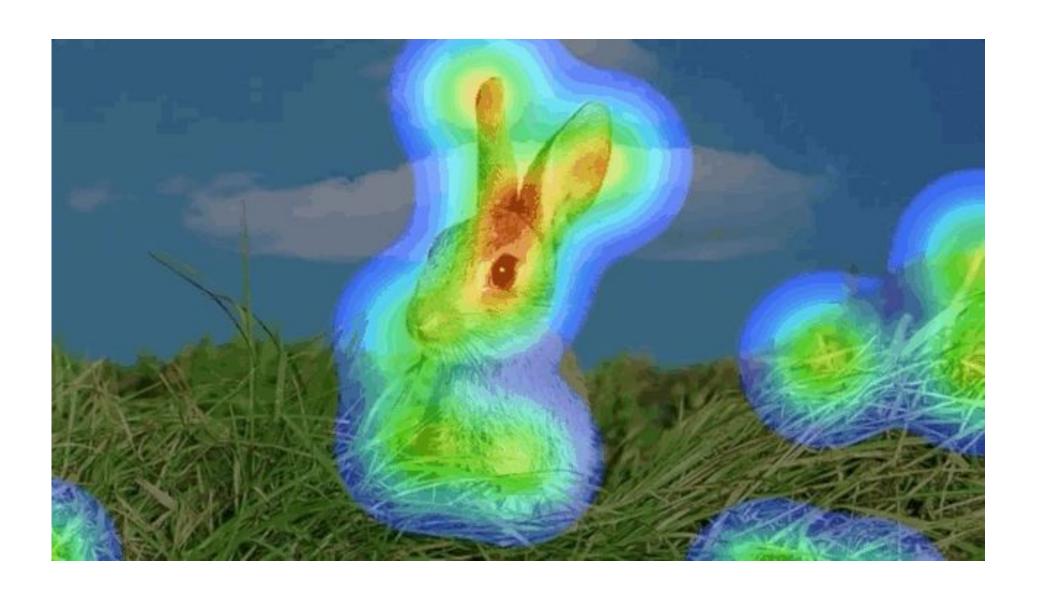
Transformer and GNN

Transformer and GNN

- self-attention
- Transformer
- Graph transformer

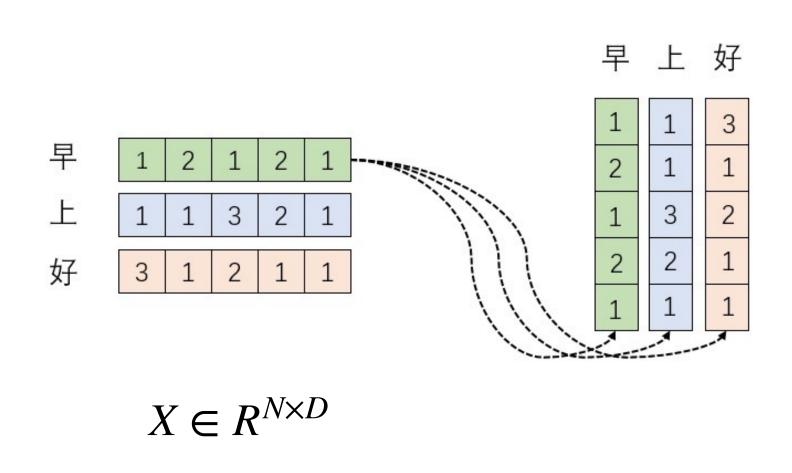
attention

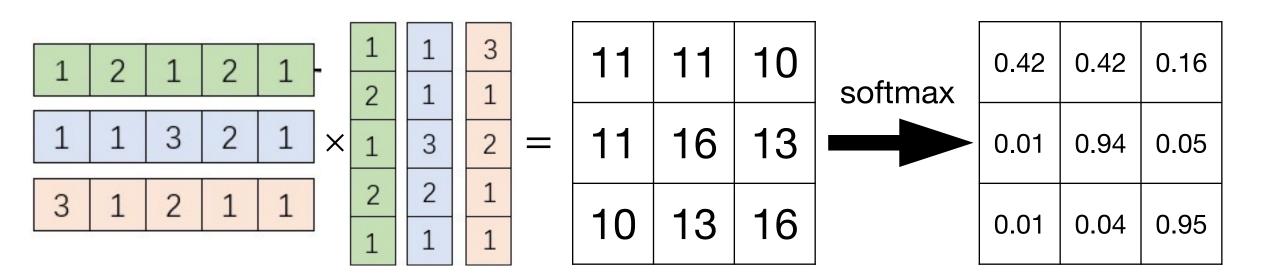




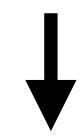
注意力机制: 让模型学习不同区域的重要性, 忽略无关信息, 关注重要信息

simple self-attention





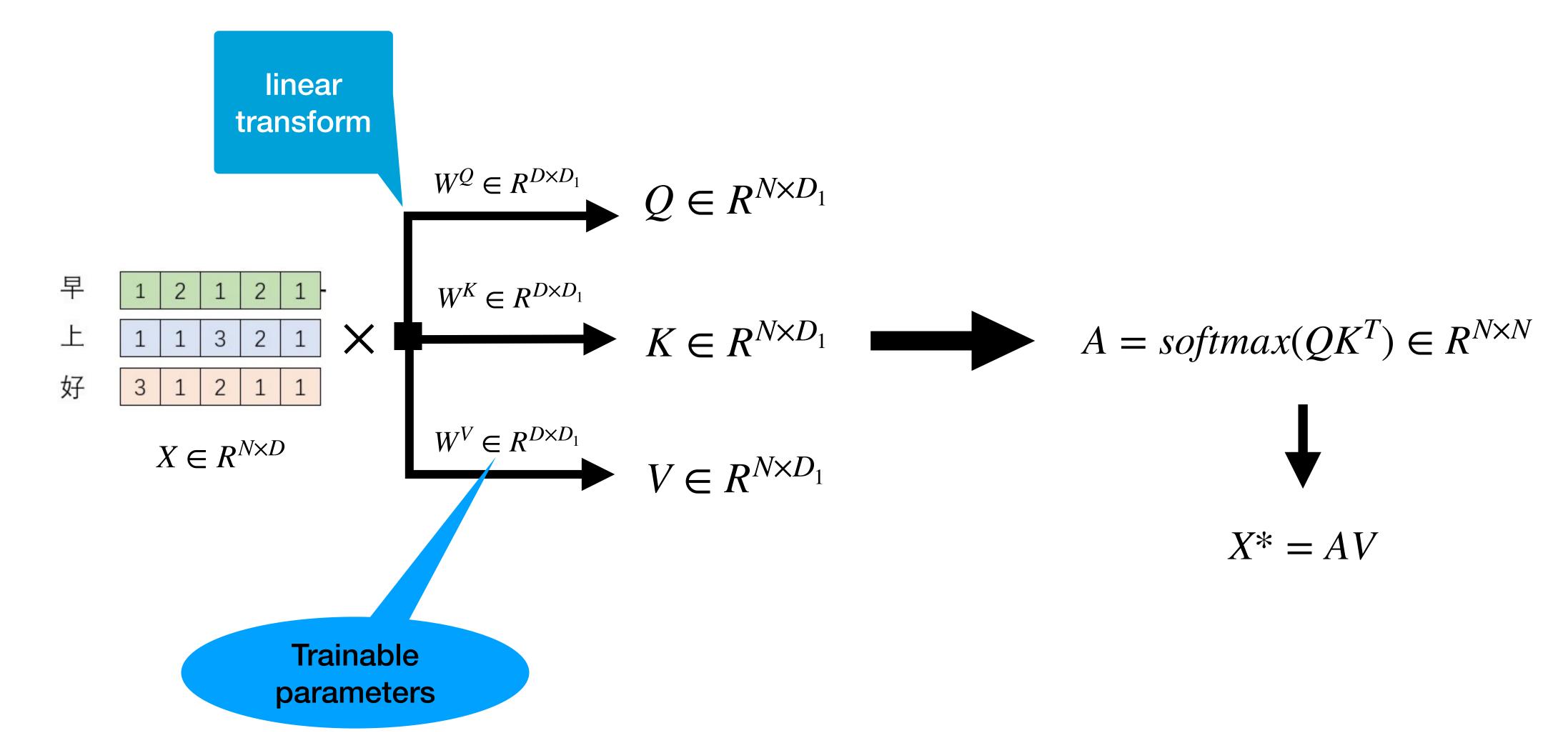
$$A = softmax(XX^T) \in R^{N \times N}$$



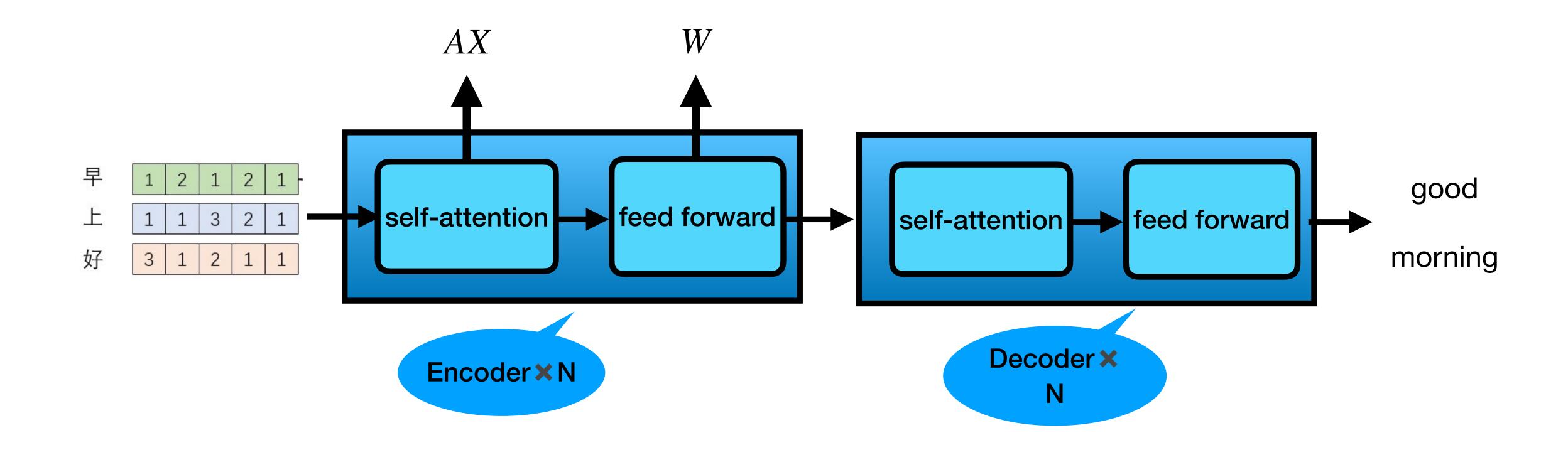
$$X^* = AX$$

No trainable parameters

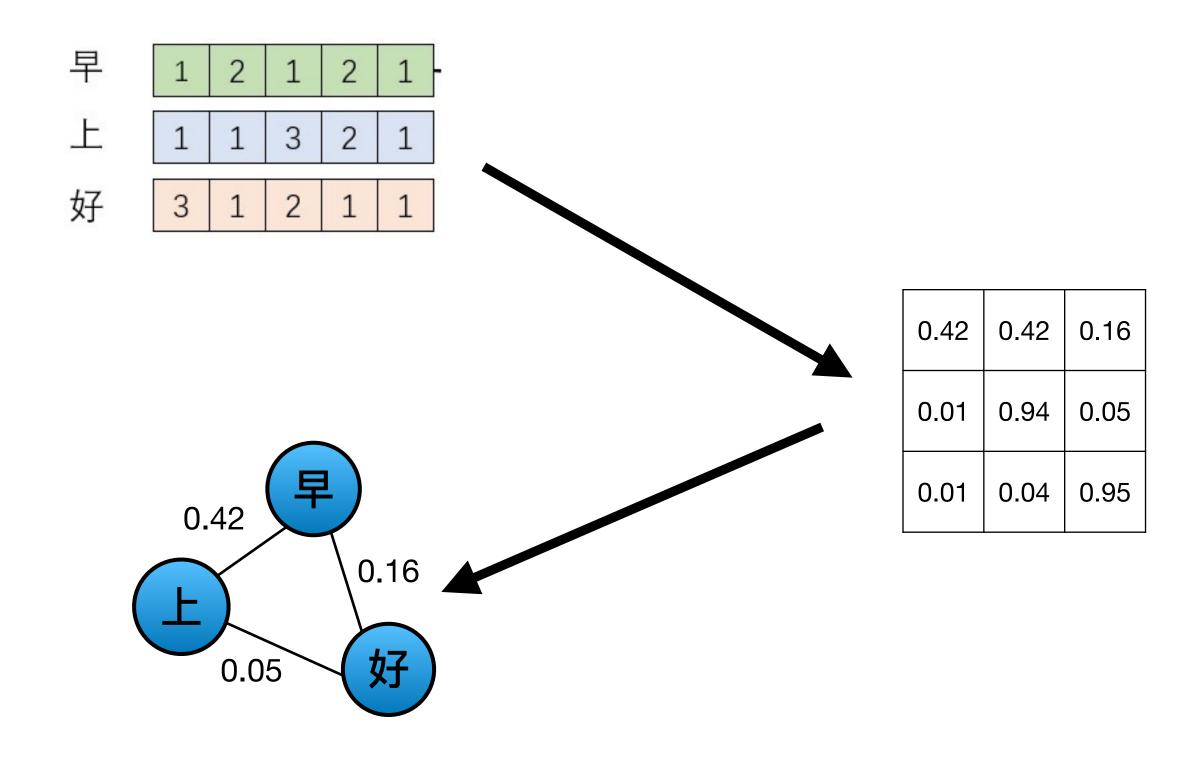
trainable self-attention



Transformer



attention matrix is adjacency matrix



如果将一个句子看成一个Graph,每个词是一个node,注意力矩阵A实际上是整个图的带权邻接矩阵,并且整个图是全连接的

attention matrix is adjacency matrix

A: 邻接矩 阵, 0 or 1

 $X^* = AXW$

A simple GNN

A: 自注意力, [0, 1], 全连接

$$X^* = AXW$$

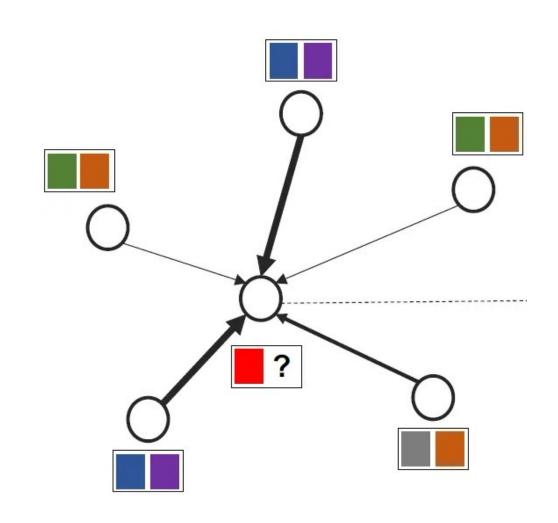
A simple Transformer

Transformer is a special GNN

- Transformer不把图的邻接矩阵(先验信息)作为模型输入,而是通过注意力机制自己学习'邻接矩阵',仅输入节点表示,因此每个节点会通过自注意力(学习到的邻接矩阵)聚合来自其它所有节点的信息。
- GNN(MPNN)同时接受节点特征和邻接矩阵矩阵(先验信息)作为模型输入,每个节点在学习过程中只聚合邻居节点的信息(使用了邻接矩阵先验)。

Transformer 本身就是一种不依赖先验邻接矩阵的GNN

Transformer is equivalent to GAT



GAT: 使用自注意力机制,为节点的每个邻居节点学习一个权重,在聚合邻居节点特征的时候,乘上权重

当GAT处理的图是全连接图的时候,GAT就是 Transformer