In [1]:
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import torch
import torch.nn as nn
import torch.nn.functional as F

class PositionalEncoding(nn.Module):
    def __init__(self, d_model, max_len=5000):
        super(PositionalEncoding, self).__init__()
        pe = torch.zeros(max_len, d_model)
        position = torch.arange(0, max_len, dtype=torch.float).unsqueeze(1)
        div_term = torch.exp(torch.arange(0, d_model, 2).float() * (-torch.log(torch.tensor(10000.0)) / d_model))
        pe[:, 0::2] = torch.sin(position * div_term)
        pe[:, 1::2] = torch.cos(position * div_term)
        pe = pe.unsqueeze(0).transpose(0, 1)
        self.register_buffer('pe', pe)

    def forward(self, x):
        return x + self.pe[:x.size(0), :]

class Transformer(nn.Module):
    def __init__(self, src_vocab_size, tgt_vocab_size, d_model, nhead, num_encoder_layers, num_decoder_layers, dim_feedfo
        super(Transformer, self).__init__()
        self.embedding = nn.Embedding(src_vocab_size, d_model)
        self.pos_encoder = PositionalEncoding(d_model)
        self.transformer = nn.Transformer(d_model=d_model, nhead=nhead, num_encoder_layers=num_encoder_layers,
                                          num_decoder_layers=num_decoder_layers, dim_feedforward=dim_feedforward, dropou
        self.fc_out = nn.Linear(d_model, tgt_vocab_size)

    def forward(self, src, tgt):
        src = self.embedding(src)
        src = self.pos_encoder(src)
        tgt = self.embedding(tgt)
        tgt = self.pos_encoder(tgt)
        output = self.transformer(src, tgt)
        output = self.fc_out(output)
        return output
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In [ ]:
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