https://colab.research.google.com/drive/1u-610KA-urqfJjDH5O0pecwfP--V9DQs?usp=sharing

模型任務:聲音特徵分類原本的模型架構:

```
class Classifier(nn.Module):
   def __init__(self, d_model=80, n_spks=600, dropout=0.1):
       super().__init__()
       # Project the dimension of features from that of input into d_model.
       self.prenet = nn.Linear(40, d model)
           Change Transformer to Conformer.
           https://arxiv.org/abs/2005.08100
       self.encoder_layer = nn.TransformerEncoderLayer(
           d model=d model, dim feedforward=256, nhead=2
       # self.encoder = nn.TransformerEncoder(self.encoder layer, num layers=2)
       # Project the the dimension of features from d model into speaker nums.
       self.pred layer = nn.Sequential(
           nn.Linear(d model, d model),
           nn.Sigmoid(),
           nn.Linear(d_model, n_spks),
   def forward(self, mels):
       args:
           mels: (batch size, length, 40)
       return:
           out: (batch size, n_spks)
       # out: (batch size, length, d_model)
       out = self.prenet(mels)
       # out: (length, batch size, d model)
       out = out.permute(1, 0, 2)
       # The encoder layer expect features in the shape of (length, batch size, d model).
       out = self.encoder_layer(out)
       # out: (batch size, length, d_model)
       out = out.transpose(0, 1)
       # mean pooling
       stats = out.mean(dim=1)
       # out: (batch, n_spks)
       out = self.pred layer(stats)
       return out
```

增加 self-attention pooling、head 數改成 8

```
class SelfAttentionPooling(nn.Module):
        def __init__(self, d_model):
                super(SelfAttentionPooling, self).__init__()
                self.attention = nn.Linear(d_model, 1)
        def forward(self, x):
                # x: (length, batch size, d_model)
                attn_weights = torch.softmax(self.attention(x), dim=0)
                attn_output = torch.sum(attn_weights * x, dim=0)
                return attn_output
 class Classifier (nn. Module) :
    def __init__(self, d_model=80, n_spks=600, dropout=0.1):
        super().__init__()
        self.prenet = nn.Linear(40, d_model)
        self.encoder_layer = nn.TransformerEncoderLayer(
            d_model=d_model, dim_feedforward=256, nhead=8
        self.self_attention_pooling = SelfAttentionPooling(d_model)
        self.pred_layer = nn.Sequential(
            nn.Linear(d_model, d_model),
            nn.Sigmoid(),
            nn.Linear(d_model, d_model),
            nn. Sigmoid(),
            nn.Linear(d_model, n_spks),
        )
    def forward(self, mels):
        out = self.prenet(mels)
        out = out.permute(1, 0, 2)
        out = self.encoder_layer(out)
        out = self.self_attention_pooling(out)
        out = self.pred_layer(out)
修改過後的效果:53.2%
```

```
Train: 100% 2000/2000 [00:42<00:00, 46.67 step/s, accuracy=0.59, loss=1.87, step=7e+4] Valid: 100% 5664/5667 [00:02<00:00, 1911.74 uttr/s, accuracy=0.53, loss=2.17] Train: 0% 0/2000 [00:00<?, ? step/s] Step 70000, best model saved. (accuracy=0.5323)
```

心得:

一開始加了 self-attention pooling 沒甚麼效果,後來把 head 數改成 8 之後效果才變好。