

Score Calibration in speaker verification

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Score Calibration

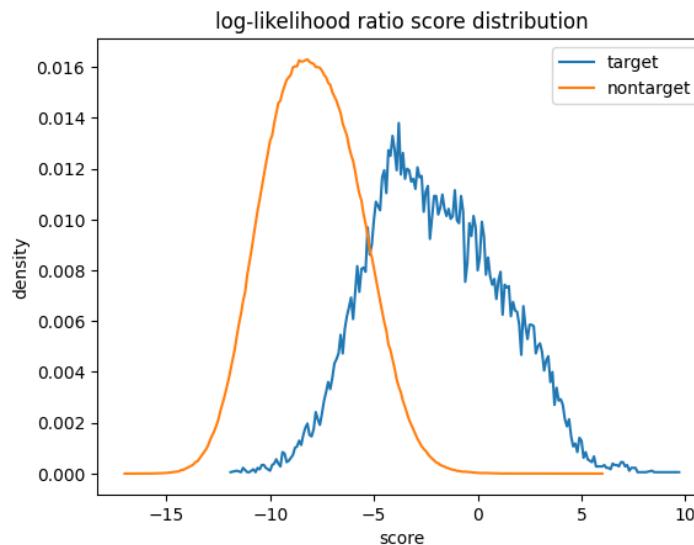
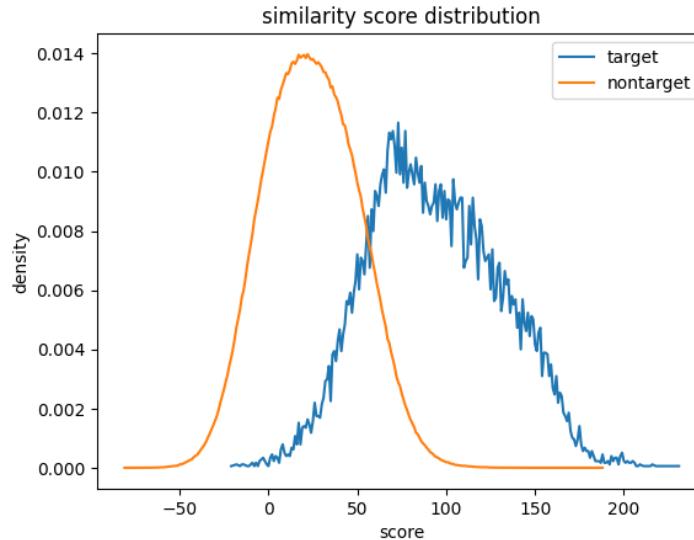
- Introduction
 - Mismatching between enrollment and test utterances leads to shifting of scores and unreliable accept/reject decisions.
 - Therefore, the purpose of score calibration is to suppress the impact of score drift on performance.

Linear Calibration:

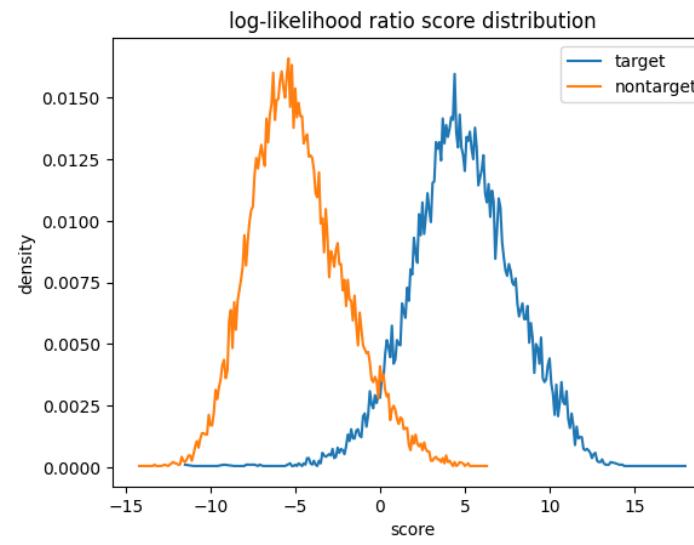
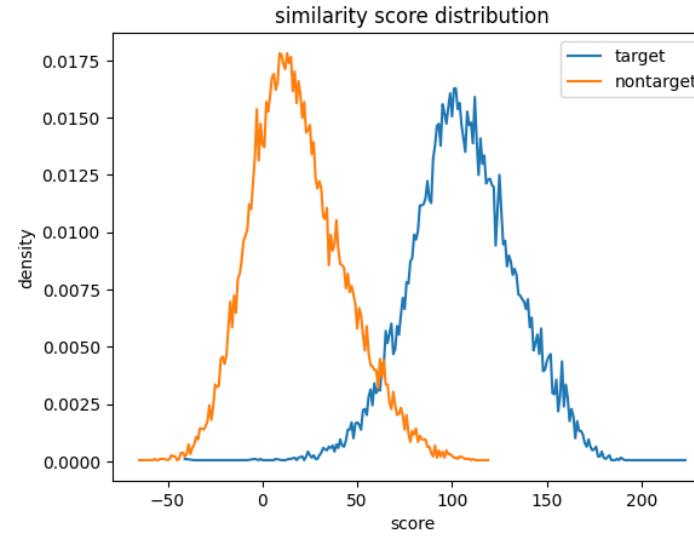
- $y = l(s) = w_0s + b$
- Converting **similarity score** to **log-likelihood ratio** $l(s) = \log \frac{p_1(s)}{p_0(s)}$
- Fusion: $y = l(s) = \sum_{i=1}^n w_i s_i + b$

Score Distribution

CN-Celeb



VoxCeleb



QMF (Quality Measure Function)

QMF is used to measure the speaker identity-independent information difference between enroll and test.

Calibration with **QMFs (Quality Measure Functions)**: $y = l(s) = w_0s + w_1x_1 + w_2x_2 + \dots + w_nx_n + b$

其中：

- y 为 1 (target) or 0 (nontarget)
- s 为 enroll 与 test 的 similarity score
- $l(s)$ 为 s 对应的 log-likelihood ratio
- x_i 为 enroll 与 test 的 quality measure, 例如 embedding length ratio、acoustic distance 等
- 学习参数: 权重 w_0, w_1, w_2 与偏置 b

$Q(q_e, q_t)$ q_e : quality measure of enroll, q_t : quality measure of test

- $Q = w * \frac{q_e}{q_t}$
- $Q = w * \left| \log \frac{q_e}{q_t} \right|$
- $Q = w * \log^2 \frac{q_e}{q_t}$

Experiments on CN-Celeb

- Pretrain model: Training on 200 speaker sampling from CN-Celeb

Calibration \ CNC1.Eval	EER (%)	minDCF (0.01)	minDCF (0.001)
non (similarity score)	14.785	0.71413	0.80018
non (similarity score after AS-norm)	14.148	0.71928	0.86103
magnitude	13.861	0.72208	0.86935
duration	13.495	0.72189	0.86616
mean imposter score	13.534	0.70337	0.85751
duration + imposter	13.050	0.70520	0.86345
magnitude + duration + imposter	13.027	0.70256	0.86161

- magnitude, duration QMF can decrease EER, but will increase minDCF
- These three QMFs are somewhat complementary

Experiments on CN-Celeb

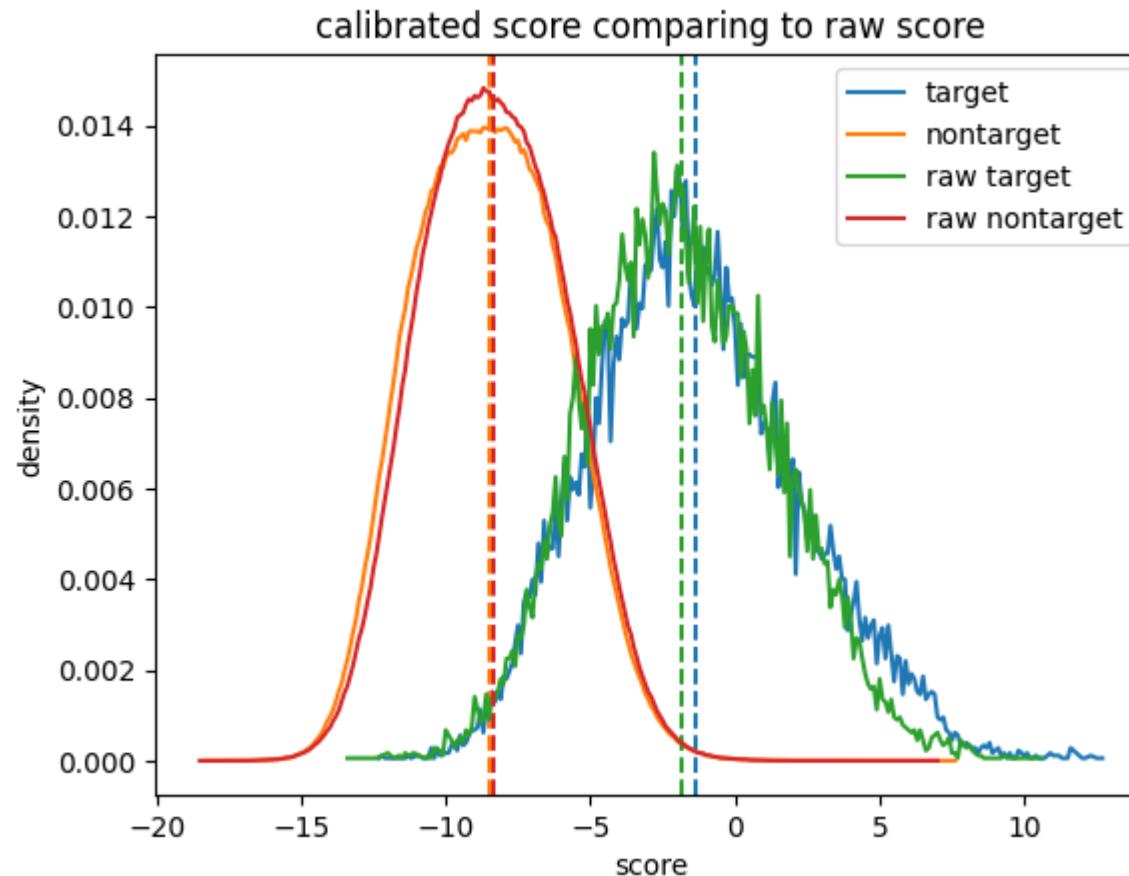
- Pretrain model: Training on 200 speaker sampling from CN-Celeb
- Scoring with multi-enroll embedding average

Calibration \ CNC1.Eval	EER (%)	minDCF (0.01)	minDCF (0.001)
non (similarity score)	14.002	0.70074	0.81121
non (similarity score after AS-norm)	13.309	0.70554	0.84984
magnitude	13.675	0.70682	0.84985
duration	12.757	0.70729	0.86269
mean imposter score	12.565	0.66610	0.83116
duration + imposter	12.278	0.66824	0.83542
magnitude + duration + imposter	12.199	0.66683	0.83499

- Imposter QMF can bring about a big performance improvement, especially minDCF

Score Distribution

CN-Celeb EER: 14.002% --> 12.199%



Experiments on VoxCeleb

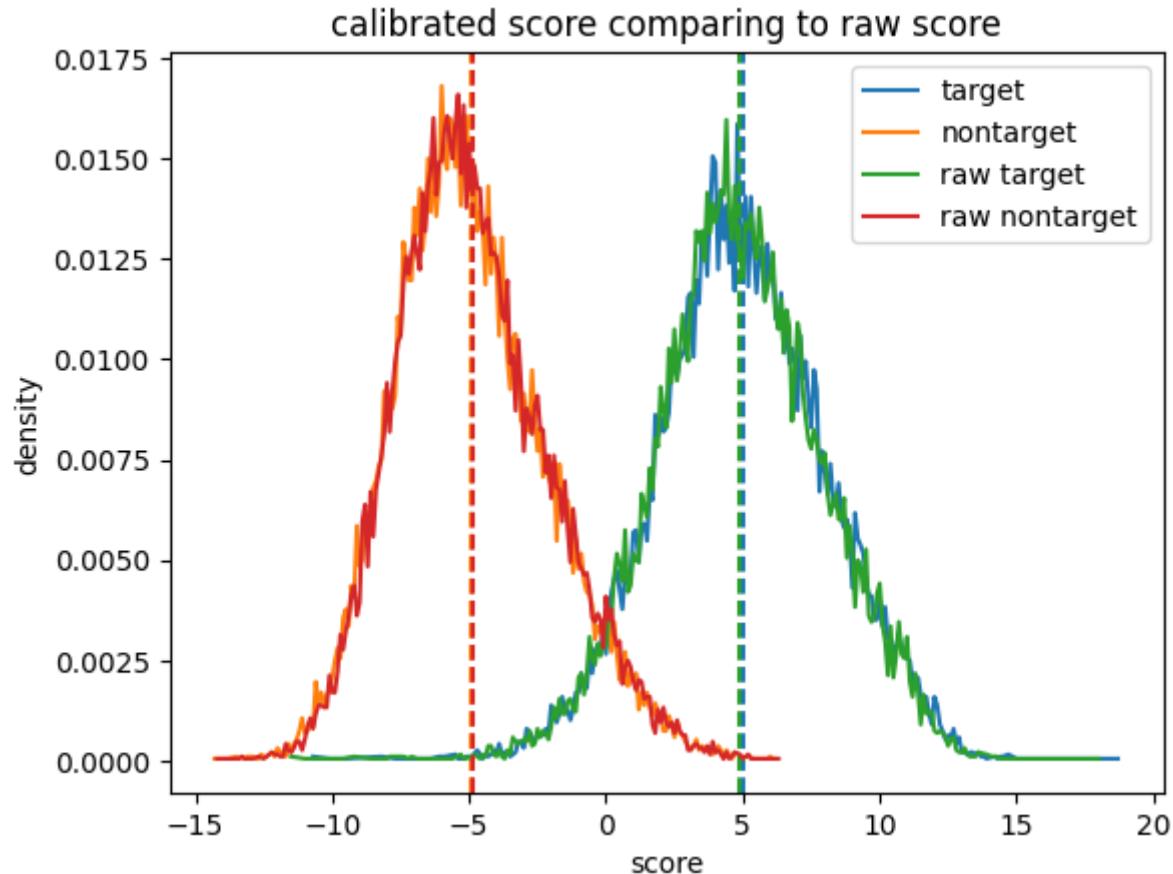
- Pretrain model: Training on 200 speaker sampling from VoxCeleb

Calibration \ Vox1-Clean	EER (%)	minDCF (0.01)	minDCF (0.001)
non (similarity score)	5.611	0.51322	0.65550
non (similarity score after AS-norm)	5.340	0.56127	0.68971
magnitude	5.321	0.56488	0.68897
duration	5.340	0.56368	0.69455
mean imposter score	5.340	0.56127	0.68971
duration + imposter	5.167	0.54229	0.66541
magnitude + duration + imposter	5.172	0.54868	0.66504

- The complementarity of the three QMFs is limited, especially the magnitude

Score Distribution

VoxCeleb EER: 5.611% --> 5.172%



Experiments on acoustic distance

- Pretrain model: Training on 200 speaker sampling from CN-Celeb

model \ CNC1.Eval	EER (%)	minDCF (0.01)	minDCF (0.001)
non (similarity score after AS-norm)	14.148	0.71928	0.86103
vggish	14.103	0.72282	0.86547
ECAPA-TDNN_softmax	14.137	0.71661	0.85624
ResNet34L_softmax	14.109	0.71693	0.85752