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The University of Tokyo

2025-07-12



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			65

1. XXXXXXXXX

1.1 **XXXXXXXXX vs XX**

1. XXXXXXX

XX

- MXXXXXXChomsky & Halle 1968\(\text{\text{M}}\)
- MXXXXIPrince & Smolensky 1993

 \mathbb{Z} [sak \mathbb{Z} [sak \mathbb{Z}] \rightarrow /sakura/

- MM/s/, /a/, /k/, /u/, /r/, /a/MM/MM/M
- MM[+consonantal], [-voice], [+coronal]......MMMMMMM



- word2vec, Skip-gram\(Mikolov et al. 2013\(\)

GAN XXXXXX (Begus, 2020)X

Vector Quantization (VQ) XXX → XXXXXXXX (Higy et al., 2021)

- k-means XXXXXXXX → XXXXXXXX

- **MXXXXX**/p/-/p^h/**XXX** (Medin et al., 2024)

1.5 XXXXXXXXX

1. XXXXXXXX

2. XXXXX



NLP XXXXXX



- XXXXXXXXXXXXXXX

NLP XXXXXX

 $\begin{array}{c} & & & \\ &$

XX

 $\frac{2}{2}$

- XXXXXXXXXXX

SSL XXXX

wav2vec 2.0 Meta/Facebook MM = MMMMMMM SSL MM (Baevski et al., 2022) M

- 3. **XXXXXXXXXXXXX**



3. SSLXXXXXXXXX

XXX (Astrach & Pinter, 2025; Venkateswaran et al., 2025)∑

4. $VQXXX \iff XXXXXXX$

4.1 VQ

 $4. \ VQ \overline{XXX} \Longleftrightarrow \overline{XXXXXX}$

VIXION VOLUMEVector Quantization, VQXX

= \times

 $\boxtimes\boxtimes\boxtimes\boxtimes$ [0.73, -0.45, 1.23] $\longrightarrow\boxtimes\boxtimes\boxtimes\boxtimes$ \boxtimes \boxtimes

XXXXXX

4.2 K-means WW VQ W

XXXX

- 1. wav2vec 2.0
- 2. K-means 🛭 128 🖾 🖾 🖽
- 3. MXXXXX ID XXXXX
- 4. XXXXXXXXXXXXXXXXX



- XXXXXn_clusters=128, random_state=42, batch_size=2048, n_init=3

- ⊠⊠⊠joblib.dump ⊠⊠ pickle ⊠

- 1. XXXXXXX: all_frames.shape = (15,234, 768)
- 2. KMeans XX: 128 XXXXXX
- 3. **XXXXXXX**: cluster_centers_.shape = (128, 768)



RQ1



- XXXXX wav2vec2-base-960h XXXXX
- XXXXXXXX vs VQ XXX(vs XXX)

XXXX

- XXXXXXXXXX

RQ2

Zubiaga, 2024)

6. XXXXXXXX



XXXX

- Docker + Poetry
- XXXXXCPU XXXXMacBook ProX- GPU XXXXXXXX

LibriSpeech	100	

6. XXXXXXXX

Common Voice	100	

6.2 XXXX

6. XXXXXXX

RQ1

XXXXXX

- XXXX wav2vec2 XXXXX 768 XXX
- MXXXVQ XXXX IDX0-127 XXXX

XXXXXXX

- **XX**train-test split (70%-30%)
- XXX63 XXXXXXXXXXXXXX

6.2 XXXX

6. XXXXXXX

RQ2

- XXXXStandardScaler XX

- XXXXIIIbrosa.pyin XXXXXXXX
- XXXXXXXX 772 XXXXXXXXXX
- XXXXXXXXX StandardScaler XX

LibriSpeech test.clean ⊠RQ1 XXX

- XXX 100 XXXXXX

Common Voice 13.0 🛮 RQ2 🖾 🛣

- XXXXXIteens, twenties, thirties, forties, fifties, sixties, seventies, eighties

6.3 XXXXXXXXXXX

```
"file": "6930-75918-0000.flac",
  "audio": {"array": [-6.10e-05, 9.15e-05, ...], "sampling_rate": 16000},
  "text": "CONCORD RETURNED TO ITS PLACE AMIDST THE TENTS",
  "speaker id": 6930
Common Voice
  "audio": {"array": [0.001, -0.002, ...], "sampling rate": 48000},
  "sentence": "The quick brown fox jumps over the lazy dog",
  "age": "twenties"
```

6. XXXXXXX

XXXXXXXX

- 1. wav2vec2-base-960h
- 3. **XXXXXXXX**Shape (**XXXXX**, 768 **XX**)
- 4. Milibrispeech_micro_continuous.npy

VQ XXXXX

- 2. MiniBatchKMeans XXX128 XXXXXX
- 3. XXXXXXXXXVq_kmeans_128_micro.pkl
- 4. **XXXXXXXXXXXXXX**

XXXXXXX

- 100 \boxtimes $(15,234,768) \rightarrow \boxtimes 15,234 \boxtimes$

VQ XXXX

- XXXXXX: [0.73, −0.45, 1.23, ...] (768 XX)
- \rightarrow XXXXX ID: 25 (0-127 XXX)

XXXX 3XXXXXX notebook XXXX

Notebook	
rq1_probing_pipeline	XXX vs VQ XXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXX

6.3 XXXXXXXXXX

6. XXXXXXX

Notebook	
rq2_hybrid_model_poc	




```
[DDD] K-MeansDall_framesDshape: (15234, 768), Dtype: float32
VQDD (KMeans) D128 DDDDDD...
[DDD] DDDDDDShape: (128, 768)
VQDDD outputs/models/vq kmeans 128 micro.pkl
```

- Poetry XXX Python XXXX
- Docker XXXXXXXX OS XXXXXXX
- requirements XXXXX XXXXXXX
- Hugging Face datasets/transformers

7. XXXXXXX

XXXXX

- XXXXX63 XXXXXXXXXXXX
- XXXXXXXLibriSpeech micro (100 XXXXXX 33,464 XXXX)
- XXXXXtrain-test split (70%-30%) XXXXXXXXXX

XXXXXXXXX

- G2P-ENXXXXX "A MAN SAID..." \rightarrow XXX ['AH0', '', 'M', 'AE1', 'N', ...]
- $\boxtimes 150 \boxtimes \boxtimes \boxtimes 10 \boxtimes \longrightarrow \boxtimes \boxtimes 15 \boxtimes \boxtimes \boxtimes \boxtimes \boxtimes$

- MXXXXI"CONCORD RETURNED TO ITS PLACE AMIDST THE TENTS"
- XXXX42 XXX['K', 'AA1', 'N', 'K', 'AO2', 'R', 'D', ' ', 'R', 'IH0', 'T', 'ER1', 'N', 'D', ...]
- MXXXXXXX175 MXXXX

VQ XXXXXXXXXXX

VQ XXXXXXXX

VQ XXXXXXXXXX

XXXXXXXX

- VQ XX:XXXX ID=52 (XXXXXXX)
- XXXcatX: [ID:52, ID:23, ID:78] → XXXXXX



XXXXX

- XXXXXXXXXXX VS XXXXXX
- Matrain-test split + Matrain-test split + classification_report

7. XXXXXXX

- XXXXStandardScaler

- XXXXXnp.hstack XXXXX → 772 XXXXXX
- MXXXXXXX StandardScaler XX

XXXXXXXX

- X_acoustic: (100, 4) [mean_f0, std_f0, jitter, shimmer]
- XXXXXXX ≈ 0.0, XXXX ≈ 1.0 XXXXXXX

7. XXXXXXX

- **XXXXXXX** 'twenties'
- **XXXX** [mean_f0: 192.33, std_f0: 15.7, jitter: 0.02, shimmer: 0.1]



RQ1

RQ2



- 2. MXXXXXXIII torchaudio.transforms.Resample
- 4. MXXXXIII torch.no_grad() MX GPU MXXXXIII
- 5. XXXXCPU XXXnumpy XXXXXXXXXX

XXXXX

G2P-EN XXXXXXXXX	Montreal Forced Aligner 🖾
100 XXXX 2 XXXXXX	
CPU ⊠⊠	MXXX GPU XX
wav2vec2-base	WavLM-Large ₩

8. XXXXXXXX

XXXX

- WavLM-Large
- XXXXXX

XXXX

XXXXXXXX

9. XXX

- 2. **XXXXXXXXXXXXXX**

- 5. **XXXXXXXXXXXXXXXX**

XXXX

- XXXXXXXXXX

9.2 XXXXXXX

9. XXX





- data/processed/librispeech micro continuous.npy
- outputs/models/vq_kmeans_128_micro.pkl\XXXXX VQ XXX

- XXXXX63 XXXX ', 'AA0', 'AA1', 'AE1', 'AH0', ...X
- XX/XXXXXX23,424/10,040 XXXX

XXXXXX

- MXXX notebooks/prepare.ipynb XXXXXX
- Docker



Note: Reference file path needs to be adjusted for compilation

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