Seeing What You Said: Talking Face Generation Guided by a Lip Reading Expert

Jiadong Wang, Xinyuan Qian, Malu Zhang, Robby T. Tan, Haizhou Li Paper ID: 10744 Session: WED-PM-220



Introduction





 Visual quality and lip-speech sync are widely concerned aspects of talking face generation.

Introduction





 Reading intelligibility indicates how much text content can be interpreted from lip movements.

Introduction



| Stimuli | | Human Responses (count) | | | | | | |
|----------|--------|-------------------------|--------|-------|-------|-------|--|--|
| Auditory | visual | Auditory | visual | rused | Comb. | Other | | |
| ba-ba | ga-ga | 2 | 0 | 98 | 0 | 0 | | |
| ga-ga | ba-ba | 11 | 31 | 0 | 54 | 4 | | |
| pa-pa | ka-ka | 6 | 7 | 81 | 0 | 6 | | |
| ka-ka | pa-pa | 13 | 37 | 0 | 44 | 6 | | |
| | P. P. | | | | | | | |

 Visual quality and lip-speech synchronization do not explicitly reflect intelligibility.

Overview of lip-reading expert



Self-supervised Pre-training



Supervised Fine-tuning



- Self-supervised Pre-training uses the clustering class of hand-crafted audio feature or learned audio-visual feature as **pseudo labels**.
- Supervised Finetuning constructs a lip-reading experts with the pre-trained transformer encoder and a decoder and trains it with text annotation.

Audio encoder







- Local audio embedding crop a 0.2s audio segment whose centre is temporally aligned with an input image.
- Global audio embedding extract audio context features from an entire audio and then crop a feature which is temporally aligned with an input image.

Architecture





- Synthesis of talking face given a triplet of a pose image, an identity image and a speech.
- Penalize incorrect lip movements in synthesized image via a lip reading expert.
- Contrastive learning between audio embeddings and output features of the lip reading expert's encoder.

Contribution



- We tackle the **reading intelligibility** problem of speech-driven talking face generation by **leveraging a lip-reading expert**.
- To **enhance lip-speech synchronization**, we propose a novel crossmodal **contrastive learning** strategy, **assisted by a lip-reading expert**.
- We employ a transformer encoder trained synchronically with the lipreading expert to consider global temporal dependency across the entire audio utterance.
- We propose a new strategy to evaluate reading intelligibility and make the benchmark code publicly available.
- Extensive experiments show that our proposal achieve SOTA reading intelligibility and lip-speech synchronization.

Experiments



- Training dataset
 - LRS2 train set (29 hours)

Evaluation dataset

- LRS2 test set: continuous audio-visual speech recognition
- LRW test set: audio-visual word classification

Metrics

- Visual quality:
 - SSIM
 - PSNR
- Lip-speech synchronization:
 - LSE-C
 - LSE-D
- Reading intelligibility:
 - Word Error Rate on LRS2
 - Accuracy on LRW

Quantitative Result



| Method | LRW | | | | LRS2 | | | | |
|---|--------|--------|--------|-----------|--------|-----------------|--------|----------------------|----------------------------------|
| | PSNR ↑ | SSIM ↑ | LSE-C↑ | ACC (%) ↑ | PSNR ↑ | SSIM \uparrow | LSE-C↑ | WER ₁ (%) | $\operatorname{WER}_2\downarrow$ |
| Ground Truth | N.A. | 1.000 | 6.88 | 88.51 | N.A. | 1.000 | 8.25 | 23.82 | 40.9 |
| ATVGnet | 30.71 | 0.791 | 5.64 | 18.10 | 30.42 | 0.751 | 5.05 | 113.69 | 91.8 |
| Wav2Lip | 31.52 | 0.874 | 7.18 | 59.98 | 31.36 | 0.854 | 8.40 | 82.06 | 73.9 |
| Faceformer | 29.19 | 0.856 | 5.58 | 53.43 | 29.47 | 0.840 | 6.42 | 97.64 | 79.0 |
| PC-AVS* | 30.44 | 0.778 | 6.42 | - | 29.89 | 0.747 | 6.73 | - | - |
| SyncTalkFace* | 33.13 | 0.893 | 6.62 | - | 32.59 | 0.876 | 7.93 | - | - |
| $\overline{\text{TalkLip}}(\overline{l})$ | 31.24 | 0.867 | 6.44 | 79.78 | 31.38 | 0.849 | 7.58 | 45.74 | 55.7 |
| TalkLip $(l + c)$ | 31.52 | 0.867 | 6.51 | 83.17 | 31.14 | 0.850 | 7.76 | 38.00 | 49.2 |
| TalkLip (g) | 30.78 | 0.871 | 7.01 | 86.57 | 30.86 | 0.854 | 8.38 | 25.31 | 36.5 |
| TalkLip $(g + c)$ | 31.18 | 0.866 | 7.28 | 87.81 | 31.19 | 0.850 | 8.53 | 23.43 | 35.1 |
| $\bar{\mathbf{B}}$ ase w.o. $\bar{\mathcal{L}}_{lip}$ | 31.22 | 0.865 | 6.01 | 48.58 | 31.08 | 0.852 | 7.09 | 103.57 | 82.2 |
| Base w.o. $\mathcal{L}_{lip,gan}$ | 30.64 | 0.864 | 5.03 | 30.80 | 30.70 | 0.851 | 5.93 | 116.26 | 89.3 |

- g and l: global and local audio embedding
- c: Contrastive learning
- **Base** denotes **Talklip** (*l*)
- * indicates that results are scratched from another paper as these methods do not opensource their training scripts.

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Qualitative Result





Ablation on Audio Encoder





a) TalkLip (l+c)



b) TalkLip (g+c)



c) Ground Truth

Ablation on Contrastive Learning





a) TalkLip (*l*)



b) TalkLip (l+c)



c) Ground Truth

Audio Embedding Visualization





Talklip (l)

Talklip (l + c)

Talklip (g + c)

Demo















What's the Best Thing about the Royal Highland show

Conclusion



- A lip reading expert is efficient to improve reading intelligibility.
- The contrastive learning can boost not only lip-speech synchronization but also reading intelligibility.
- The transformer encoder can both improve reading intelligibility and lip-speech synchronization.
- Extensive experiments prove that our proposal achieve SOTA reading intelligibility and lip-speech synchronization.