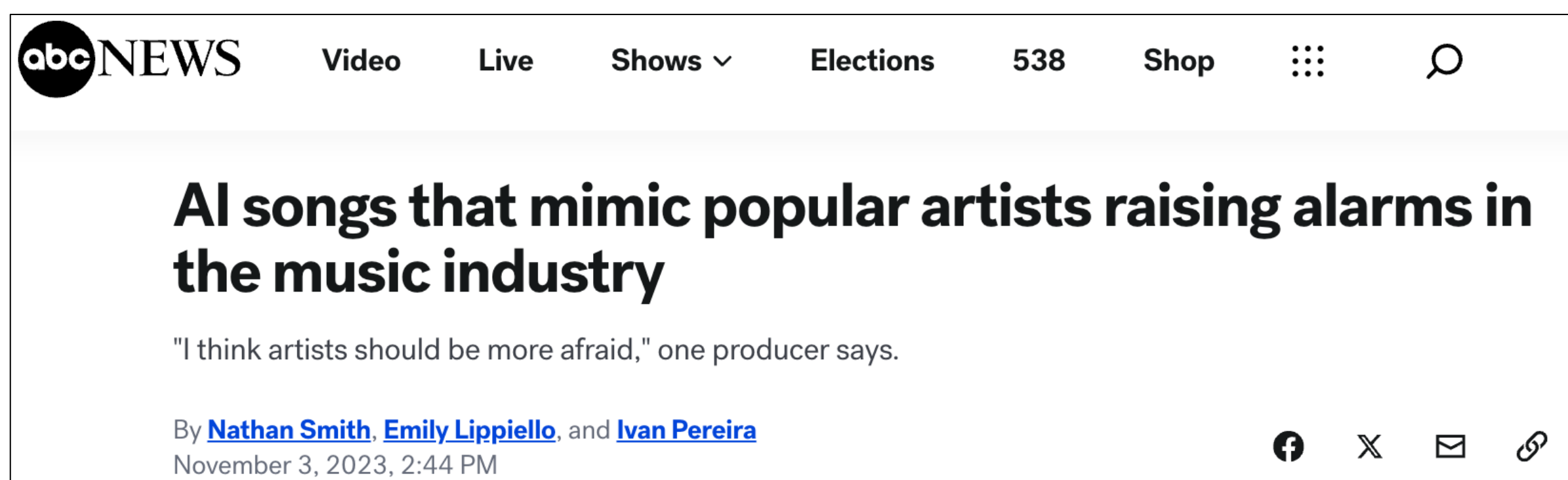


Background

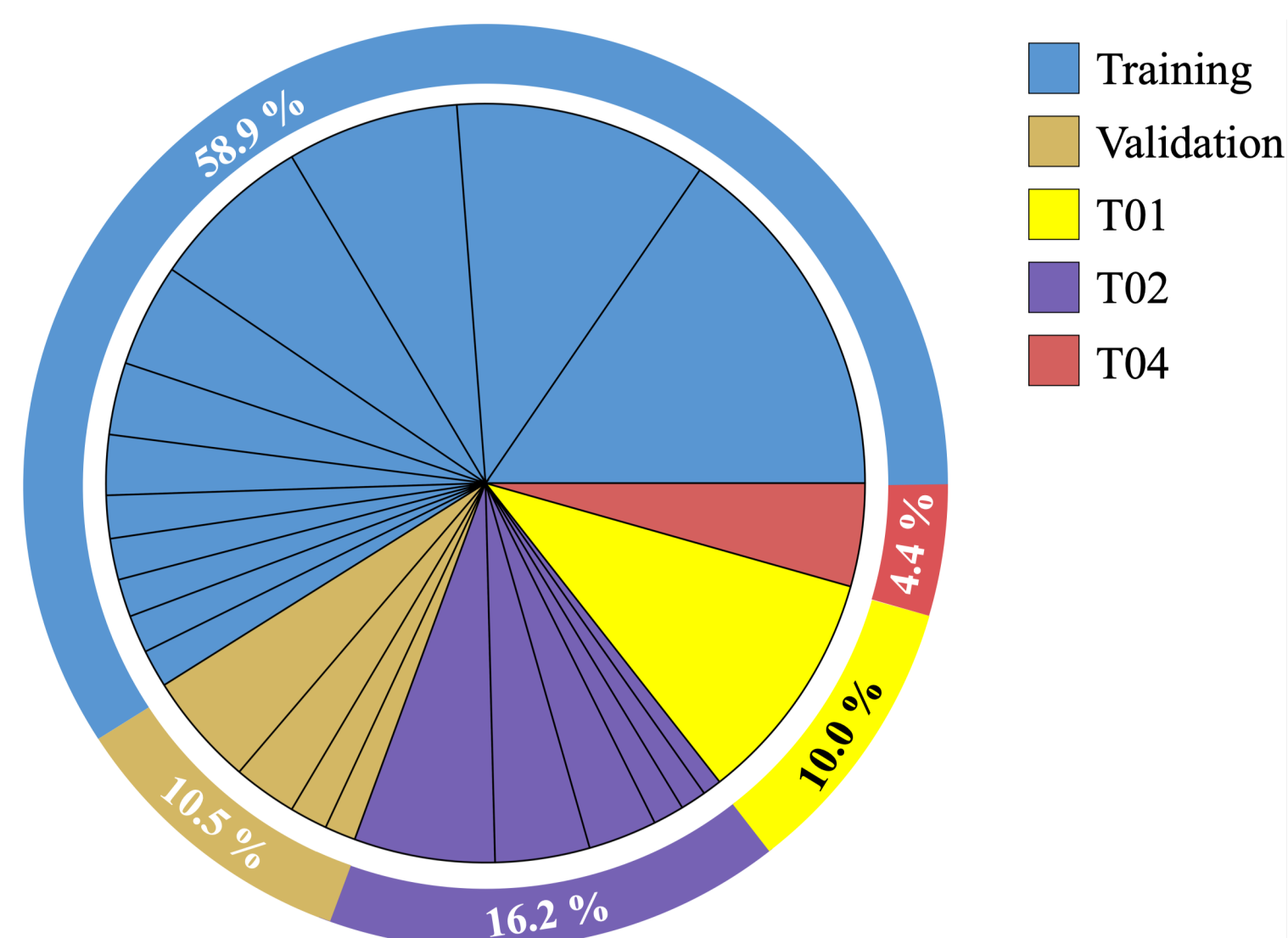
Singing Voice Deepfakes are raising public and industry concerns.



WildSVDD: Singing Voice Deepfake Detection in the Wild

Datasets collected from media platforms

Previous work: **SingFake** [1] proposed the novel task of SVDD, presented the SingFake dataset, and identified several challenges.



WildSVDD: An expanded SingFake with newly collected data

Participants can freely split the development set from the training set.

Test A: Unseen singers, similar to T02 in SingFake

Test B: Unseen musical context, same as T04 in SingFake

WildSVDD Baselines

AASIST [3] with various front-ends

Front-end	WildSVDD Test A		WildSVDD Test B	
	Mixtures	Vocals	Mixtures	Vocals
Raw Waveform	10.50	8.48	16.85	14.91
Spectrogram	27.93	20.55	30.97	24.41
Mel-Spectrogram	29.27	27.35	32.18	30.78
MFCC	17.78	19.14	22.92	23.31
LFCC	22.60	23.25	26.82	26.94
Wav2vec2 XLS-R	9.57	6.09	21.45	24.09

Acknowledgments



Resources

Website

Paper

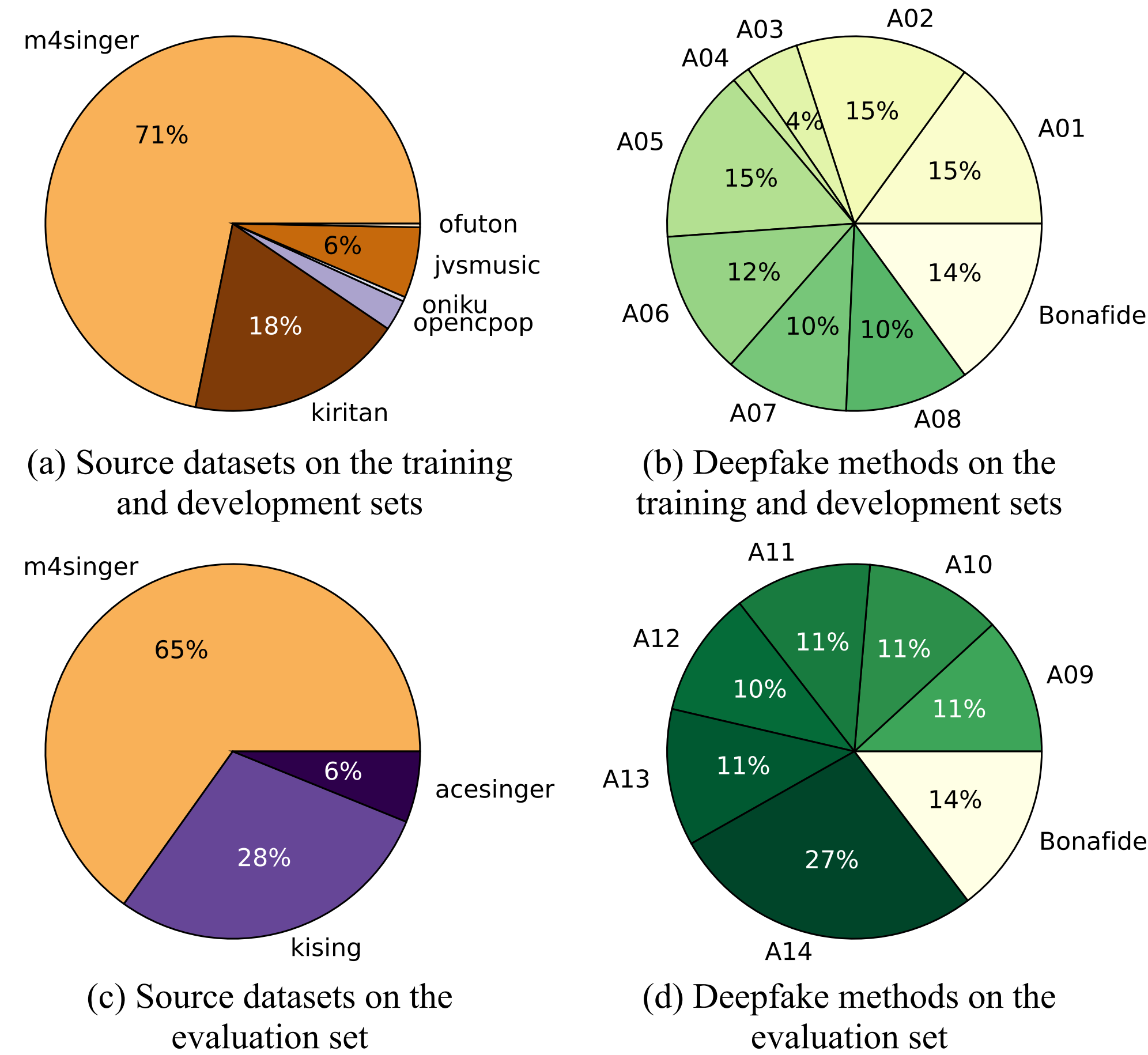


CtrSVDD: Controlled Singing Voice Deepfake Detection

CtrSVDD Dataset [2]

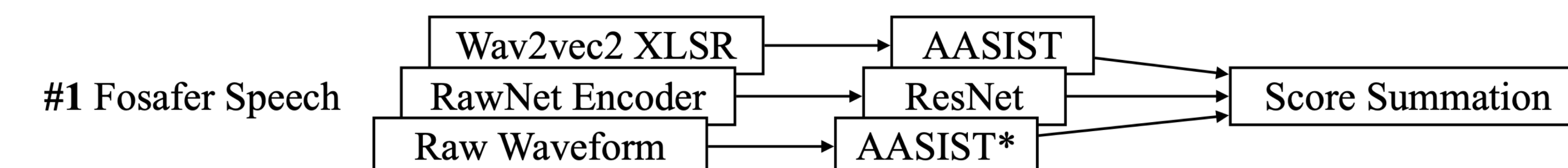
47.64 hours of bonafide vocals from open-source singing datasets
260.34 hours of deepfake vocals using 14 synthesis methods

Overview of source datasets and deepfake methods distribution [2]

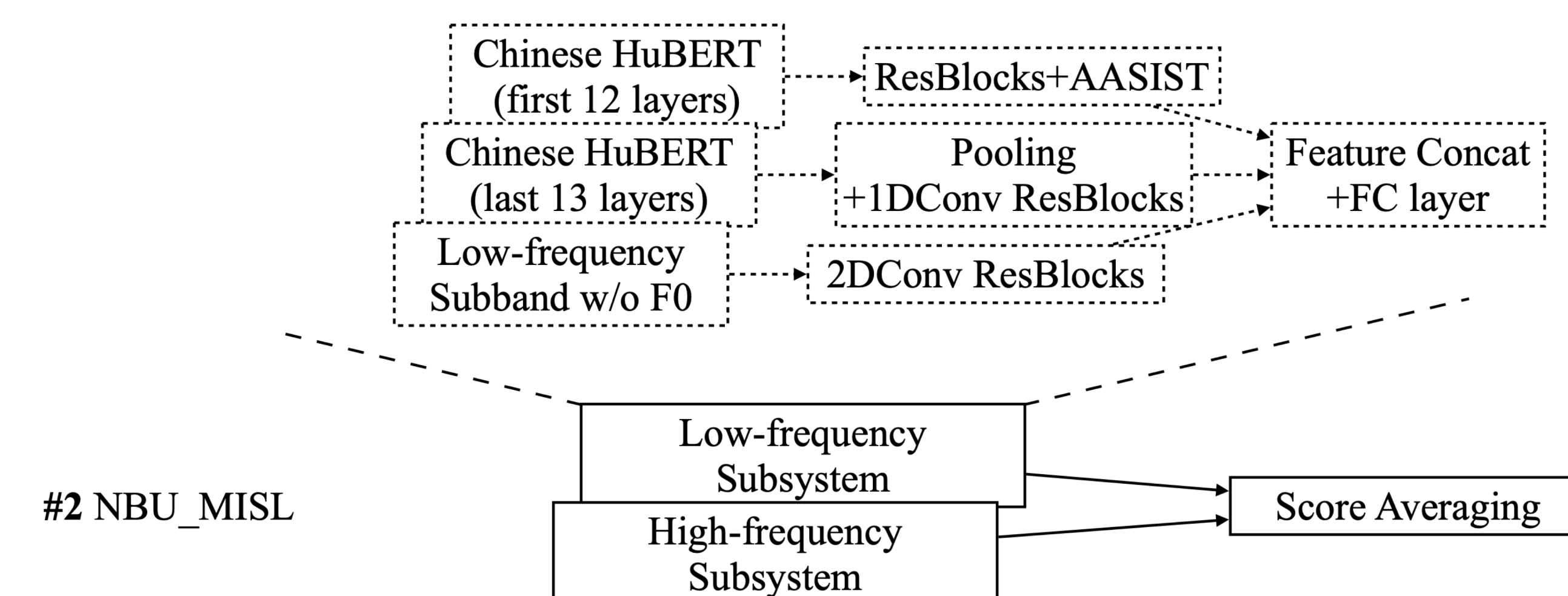


Winning solutions:

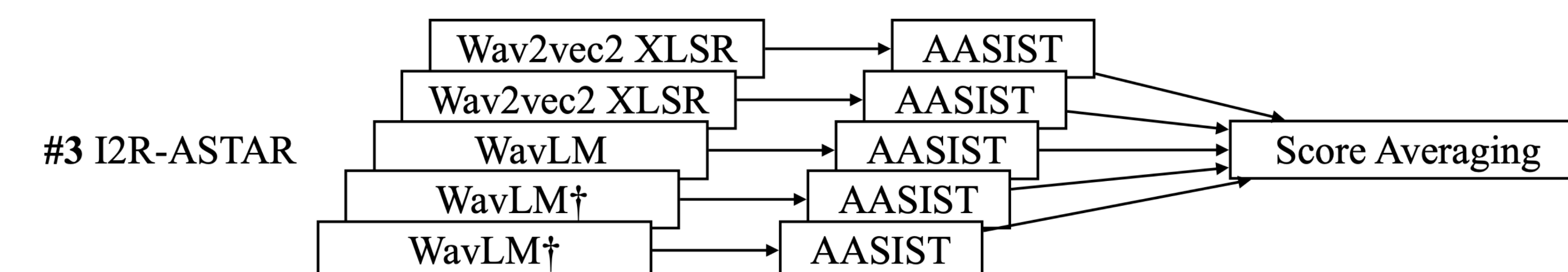
Illustration of the top-4 ranked system submissions for the CtrSVDD track



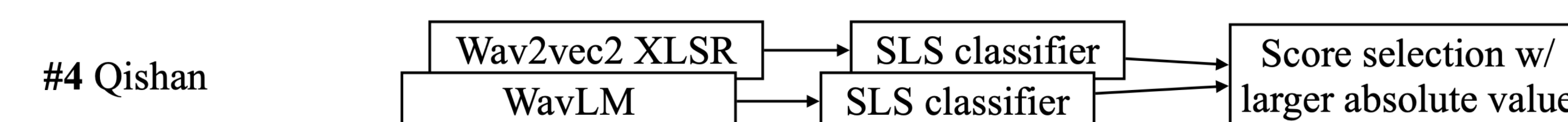
Data note: No data augmentation was used. Additional datasets were incorporated.



Data note: Augmented with HiFi-GAN vocoded audio. No additional datasets were incorporated.



Data note: Augmented with RawBoost variations. No additional datasets were incorporated.



Data note: No data augmentation was used. No additional datasets were incorporated.

Challenge results: Overview of the top-8 ranked submission results

Team Name	Results (w/o ACESinger)		Results (overall)		Per-Attack EER					Per-Dataset EER		ACESinger (A14)
	EER (%)	Rank	EER (%)	Rank	A09	A10	A11	A12	A13	KiSing	M4Singer	
Fosafer Speech	1.65	1	4.32	1	0.23	0.06	0.37	4.19	0.07	2.66	1.69	49.67
NBU_MISL	2.00	2	8.41	19	0.13	0.11	0.94	5.17	0.10	8.98	2.07	50.02
I2R-ASTAR	2.22	3	4.86	3	0.65	0.51	2.49	4.57	0.64	6.01	2.16	50.02
Qishan	2.32	4	4.45	2	1.02	0.69	2.54	4.42	0.76	2.82	2.32	50.05
Breast waves	2.73	5	5.38	5	1.50	0.76	2.03	6.14	0.88	3.56	2.84	50.44
MediaForensics beyond	2.75	6	5.83	8	0.56	0.38	3.90	4.45	1.02	10.56	2.56	49.91
Star	2.99	7	5.68	7	0.45	0.26	4.56	4.37	0.85	9.12	2.85	49.53
Star	3.31	8	5.21	4	1.64	0.19	1.11	7.30	0.23	1.79	3.51	49.70

REFERENCES

[1] Zang, Y., Zhang, Y., Heydari, M., & Duan, Z. (2024). Singfake: Singing voice deepfake detection. In *Proc. IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP)* (pp. 12156-12160).

[2] Zang, Y., Shi, J., Zhang, Y., Yamamoto, R., Han, J., Tang, Y., ... & Duan, Z. (2024). CtrSVDD: A Benchmark Dataset and Baseline Analysis for Controlled Singing Voice Deepfake Detection. *Proc. Interspeech* (pp. 4783-4787).

[3] Jung, J. W., Heo, H. S., Tak, H., Shim, H. J., Chung, J. S., Lee, B. J., ... & Evans, N. (2022). AASIST: Audio anti-spoofing using integrated spectro-temporal graph attention networks. In *Proc. IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP)* (pp. 6367-6371).