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Big picture

- Speech-to-text translation has many potential applications for low-resource languages.
- But is available for a tiny fraction of the world's spoken languages as most are zero or low resource.
- Recent work has found that neural encoder-decoder models can learn to directly translate foreign speech in high-resource scenarios.
- Will this work in settings where both data and computation are limited?
- Beyond translations, word-level precision/recall results indicate that models can be useful for keyword-spotting or topic-modeling.

Experimental Setup

- Fisher Spanish speech dataset: a multispeaker corpus of telephone calls in a variety of Spanish dialects recorded in realistic noise conditions.
- Crowdsourced English translations.
- Train models using as little as 20hrs of labeled data.
- Adapt model from state-of-the-art architecture of Weiss et al. [1].
- Simplified to fit on a single GPU: word-level decoder, no model replicas, reduced dim speech features.

Evaluation

We evaluate with several metrics:

- BLEU. Measures up to 4-gram precision.
- METEOR. Allows inexact matches in predictions.
- Word-level unigram precision.

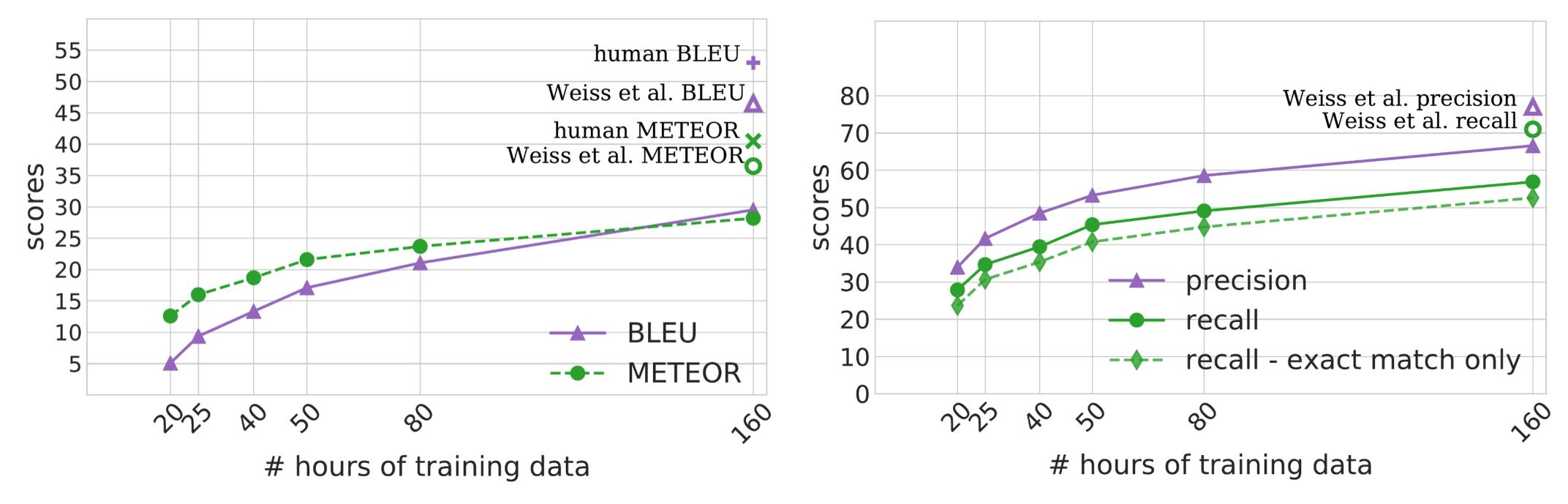


Figure 1: Fisher dev set BLEU, METEOR, precision and recall results.

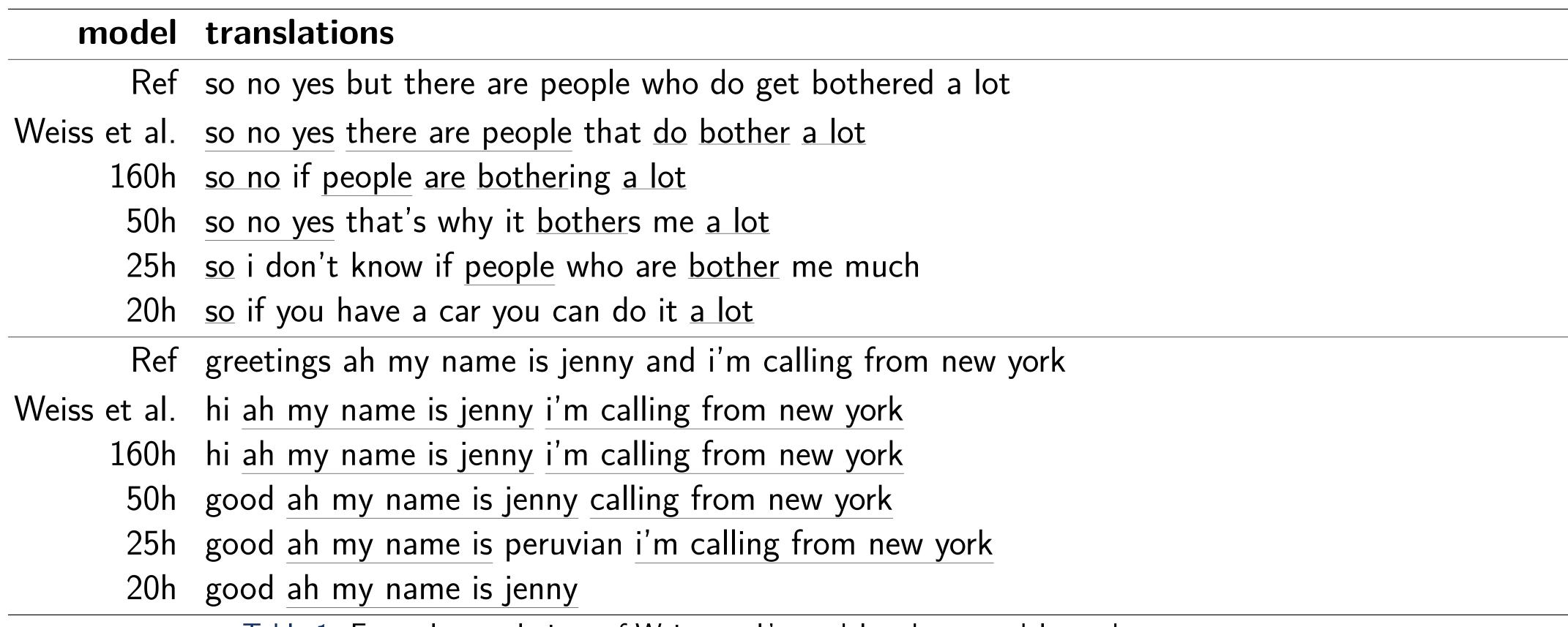


Table 1: Example translations of Weiss et al.'s model and our models on dev set utterances.

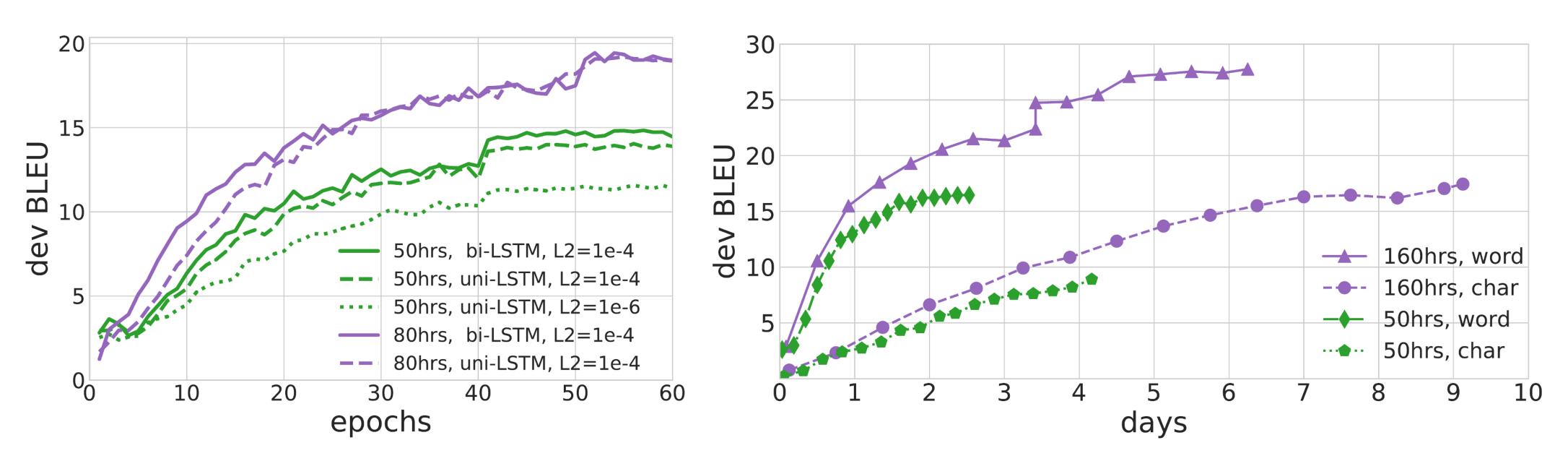


Figure 2: Left: uni-directional vs. bidirectional encoders, L2 loss penalties. Right: word vs char level model training speed

Takeaways

- With only 50hrs of labeled speech data, models achieve high precision and recall—around 50%.
- Models can be trained on a single GPU (Titan X equivalent), and converge in ~3 days.
- Regularization parameters are critical to model performance.

Future work

- Use sub-word unit modeling to strike balance between speed of word-level decoder, and generalization capacity of character-level decoder.
- Build and evaluate models for cross-lingual keyword-spotting.

Preview of the improved scores:

model	\mathbf{BLEU}
Weiss et al.	47.3
20h word	5
20h bpe+noise*	10.8
50h word	17.1
50h bpe+noise*	23.3

Table 2: BLEU scores before and after training improvements. *Yet to be published.

References

[1] "Sequence-to-sequence models can directly transcribe foreign speech", R. J. Weiss, J. Chorowski, N. Jaitly, Y. Wu, and Z. Chen. In Proc. Interspeech, 2017.

Acknowledgments

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