Tandem modeling investigations

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Outline

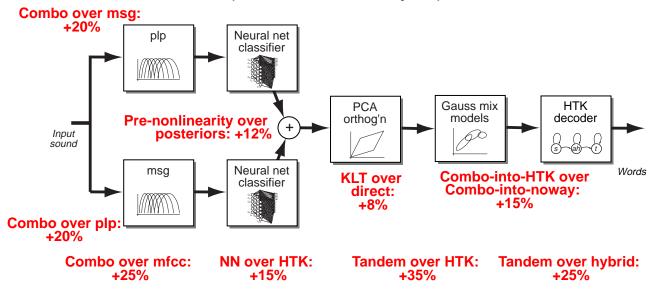
- What makes Tandem successful?
- Can we make Tandem better?
- 3 Does Tandem work with LVCSR tricks?





What makes Tandem work?

(with Manuel Reyes)

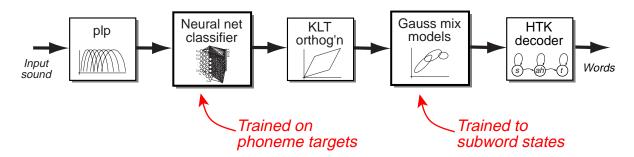


Tandem combo over HTK mfcc baseline: +53%

- Model diversity?
 - try a phone-based GMM model
 - try training the NN model to HTK state labels
- Discriminative network training?
 - (try posteriors from GMM & Bayes)



Phone vs. word models



- Try a phone-based HTK model (instead of whole-word models)
- Try training NN model to subword-state labels
 - 181 net outputs; reduce to 40 in KLT
- Results (Aurora2k, HTK-baseline WER ratio):

System	test A: matched	test B: var noise	test C: var chan
Tandem PLP baseline	63.5%	701.3%	59.5%
Phone-based HTK sys	63.6%	72.5%	61.5%
Subword-based NN sys	63.1%	62.8%	55.1%

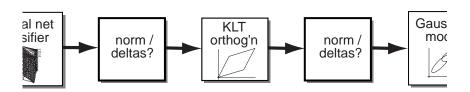
- Diversity doesn't help
 - subword units may be good for NN



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Enhancements to Tandem-Aurora

More tandem-feature-domain processing:



Results (HTK baseline WER ratio):

System	test A: matched	test B: var noise	test C: var chan
PLP: Tandem baseline	63.5%	701.3%	59.5%
PLP: norm - KLT	72.6%	71.2%	63.6%
PLP: KLT - norm	57.8%	58.8%	51.3%
PLP: KLT - delta	59.0%	60.2%	52.9%
PLP: KLT - delta - norm	58.1%	59.9%	48.9%
PLP: delta - KLT - norm	54.7%	53.6%	46.9%

- delta-KLT-norm: 80% Tdm baseline WER



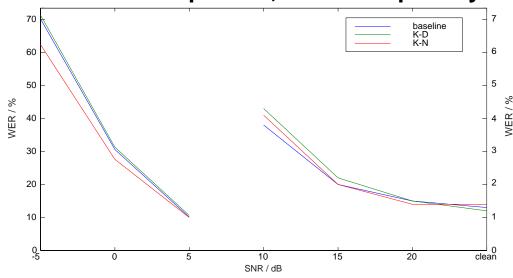
Best effort Tandem system

Deltas & norms help PLP: try on combo (PLP+MSG) system:

System	test A: matched	test B: var noise	test C: var chan
PLP+MSG: baseline	51.1%	52.0%	45.6%
PLP+MSG: dlt-KLT-nrm	50.9%	50.5%	43.6%
PLP+MSG: KLT-nrm	48.3%	49.5%	39.4%

- deltas hurt for MSG: features too sluggish?

• Deltas help clean, norms help noisy:



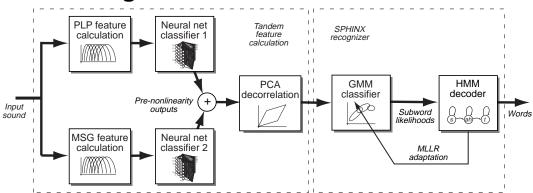
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Tandem for LVCSR: the SPINE task

(with Rita Singh/CMU & Sunil Sivadas/OGI)

- Noisy spontaneous speech, ~5000 word vocab
- Recognition:



- same tandem features
- NN training from Broadcast News boot + iterate
- GMM-HMM has context-dependence, MLLR



SPINE-Tandem results

Evaluation WER results:

Features (dimensions)	CI system	CD system	CD + MLLR
MFCC + d + dd (39)	69.5%	35.1%	33.5%
Tandem features (56)	47.6%	35.7%	32.8%

- much better for CI systems
- differences evaporate with CD, MLLR

Not quite fair:

- CD senones optimized for MFCC
- worth 2-3% absolute?

• Not unexpected:

- NN confounds CD variants
- Tandem 'space' very nonlinear bad for MLLR

Any hope?

- more training data / train CD classes / ...

