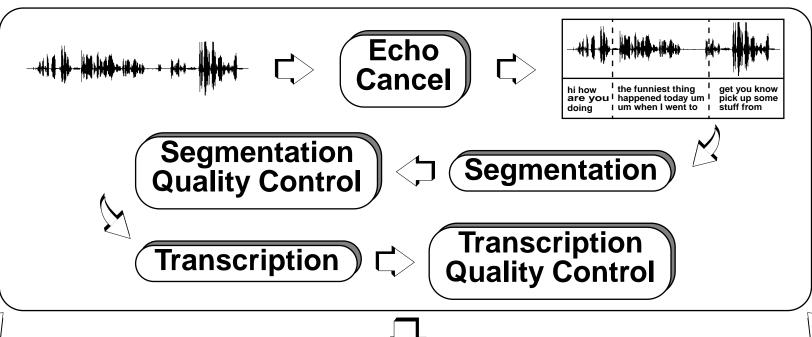
#### Introduction to SWITCHBOARD

- Challenging and popular LVCSR benchmark
- Spontaneous telephone conversations
- 240 hours, 2430 conversations, 3+ million words, 500+ speakers (male and female)
- Low bandwidth, channel noise, echo
- Speaking rates, dialects, coarticulation, speaking styles, accents, dysfluencies
- Poor quality acoustic models, large mismatch

### **Motivation**

- Reduce acoustic model mismatch
- Segmentation and transcription must capture both acoustic and linguistic properties
- Automatic (energy-based) segmentation unnatural breakpoints
- Linguistic structure-based segmentation corrupted acoustic context
- Dysfluencies make transcription difficult (Current LDC transcription WER ~ 8%)

# **Approach**



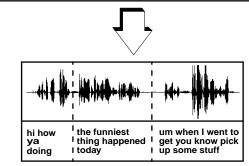




Cross Validation

Word Alignment

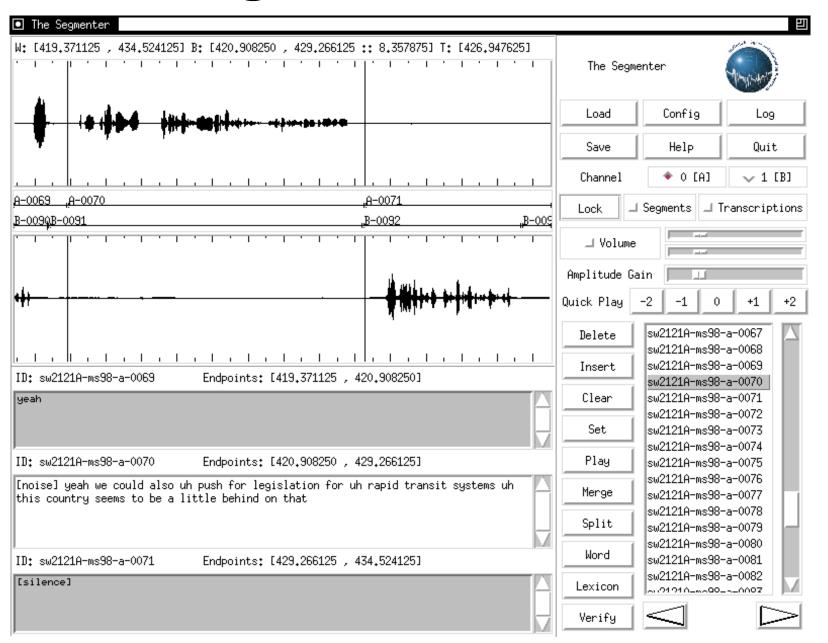
Recognition Experiments



## **Guidelines**

- Segment boundaries with at least 1 sec of silence between speech
- Segment along phrase / sentence / train-ofthought boundaries
- Merge utterances split at counterintuitive points (e.g. middle of sentence)
- Limit maximum utterance duration to 15 sec
- Fix transcriptions taking into account dysfluencies and capitalization issues

# **Segmentation Tool**

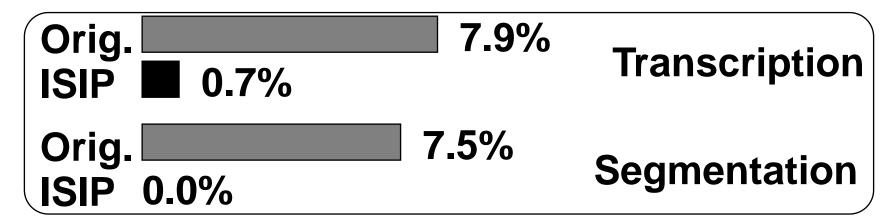


#### **Issues and Concerns**

- Large number of dysfluencies (pauses, laughter, partially pronounced words etc.)
- Affirmative statements (yes/no) and pause fillers (um/hmm) cover ~ 30% of utterances
- Marking boundaries near noise or echo
- Consistency in capitalization ("I" vs "i") and handling proper nouns
- Marking asides, background noise / music and background speech

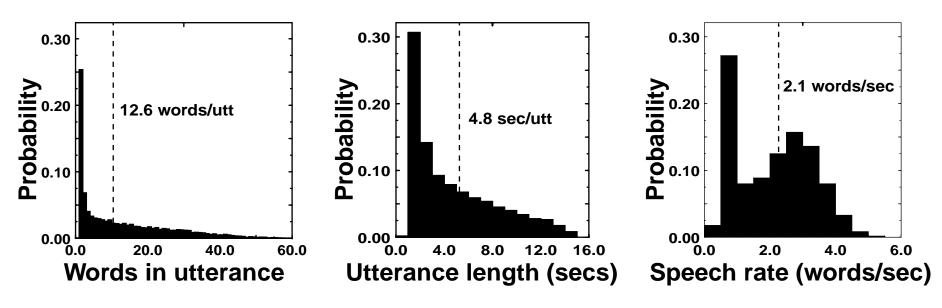
#### **Cross-Validation**

- All validators segment / transcribe the same conversation
- Adjudicated reference transcription
- Word alignment review will further reduce error rate



**Cross-Validation Word Error Rates** 

### The New SWITCHBOARD



- Segmentation and transcription rate 20xRT
- Monosyllabic words constitute 53% of data on WS'97 subset (down from 67%)
- Lexicon updates partial words, laughter words, alternate pronunciations

# Effect on Recognition

- Adapt existing acoustic models to resegmented speech data
- 20 hours training data (27500 utterances) including silence
- Word-internal triphone system to bootstrap seed models (HTK)
- 4 passes of re-estimation
- Lattice rescoring on WS'97 dev test set

### Results

Error Rate	ISIP	WS'97
Total WER	47.9%	49.8%
Correct words	55.8%	53.1%
Substitutions	31.6%	32.2%
Deletions	12.6%	14.8%
Insertions	3.7%	2.9%

- 63% of total errors on monosyllabic words (down from 71%)
- Reduction in substitution and deletion errors

# <u>Analysis</u>

- 1.9% absolute improvement in WER
- Monosyllabic words are the principal factor in error analysis
- Performance improvement attributed to better modeling of monosyllabic words
- Acoustically "complete" transcriptions (no partial words at utterance boundaries) help in improved acoustic modeling
- Longer utterance transcriptions facilitate LM application

### **Conclusions**

- Uniformity and accuracy are critical for the quality of training segmentation and transcriptions
- Segmentation at natural boundaries allows better acoustic modeling
- Dysfluencies pose significant challenges to accurate transcription
- Acoustic models trained on corrected SWB data will result in major improvements in WER (e.g. 2% absolute improvement from adapting models)