

Motivation

- ➔ **Optimization of model topology not a common feature of most contemporary acoustic modeling systems**
- ➔ **Model topology/size decided heuristically and often uniformly (eg. 3-state triphones)**
- ➔ **Analysis indicates existence of optimal model size for larger acoustic units like syllables**
- ➔ **Information theoretic measures can be used to learn model size (number of states for HMMs)**

Current Approach

- ➔ **Most triphone systems have 3 states/model**
- ➔ **Assume transition probabilities encode duration information**
- ➔ **Syllable models with number of states proportional to average model duration**
- ➔ **Syllable models with equal number of states**
- ➔ **Bayesian model merging used for Markov models (eg. pronunciation modeling)**

Previous Experience



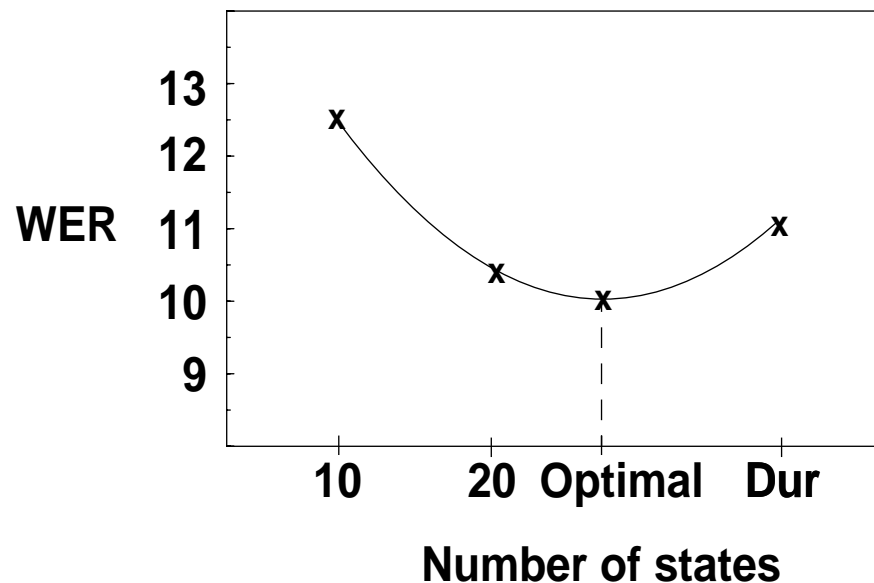
Sys1

- **No. States: median duration / 20 msec**
- **WER: 11.1% on OGI Alphadigits**



Sys2:

- **Max. states/syllable: 20 states**
- **WER: 10.4% on OGI Alphadigits**



Bhattacharyya Distance

- ➔ A separability measure between two Gaussian distributions

$$D = (M_2 - M_1)^T \cdot \left[\frac{\Sigma_1 + \Sigma_2}{2} \right]^{-1} \cdot (M_2 - M_1) + \frac{1}{2} \log \frac{\left| \frac{\Sigma_1 + \Sigma_2}{2} \right|}{\sqrt{|\Sigma_1| \cdot |\Sigma_2|}}$$

- ➔ Two terms represent separability due to class means and variances
- ➔ Used in phone clustering (Mak et. al., 1996)

Kullback-Leibler Distance

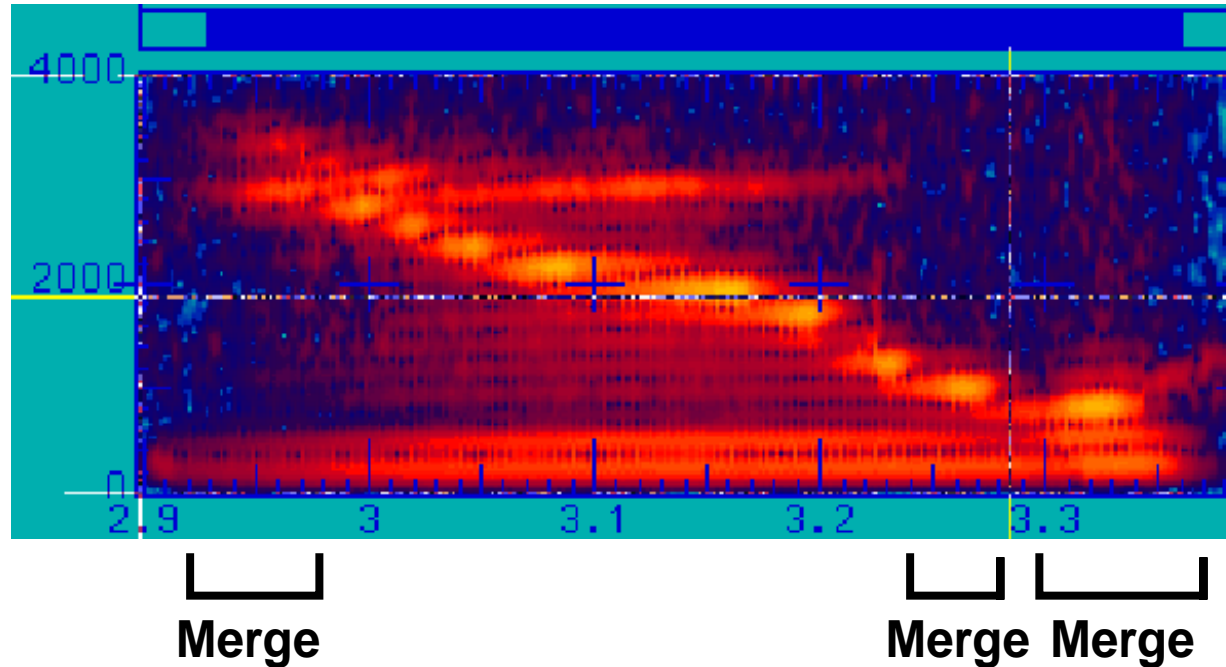
- ➡ **Divergence between two Gaussian distributions**

$$KL2(A;B) = \frac{\sigma_A^2}{\sigma_B} + \frac{\sigma_B^2}{\sigma_A} + (\mu_A - \mu_B)^2 \cdot \left(\frac{1}{\sigma_A^2} + \frac{1}{\sigma_B^2} \right)$$

- ➡ **Deviates from traditional definition of divergence to make it symmetric**
- ➡ **Successfully used for speaker change detection (Seigler et. al. 1997)**

Merging Phenomenon

Syllable _y_uw

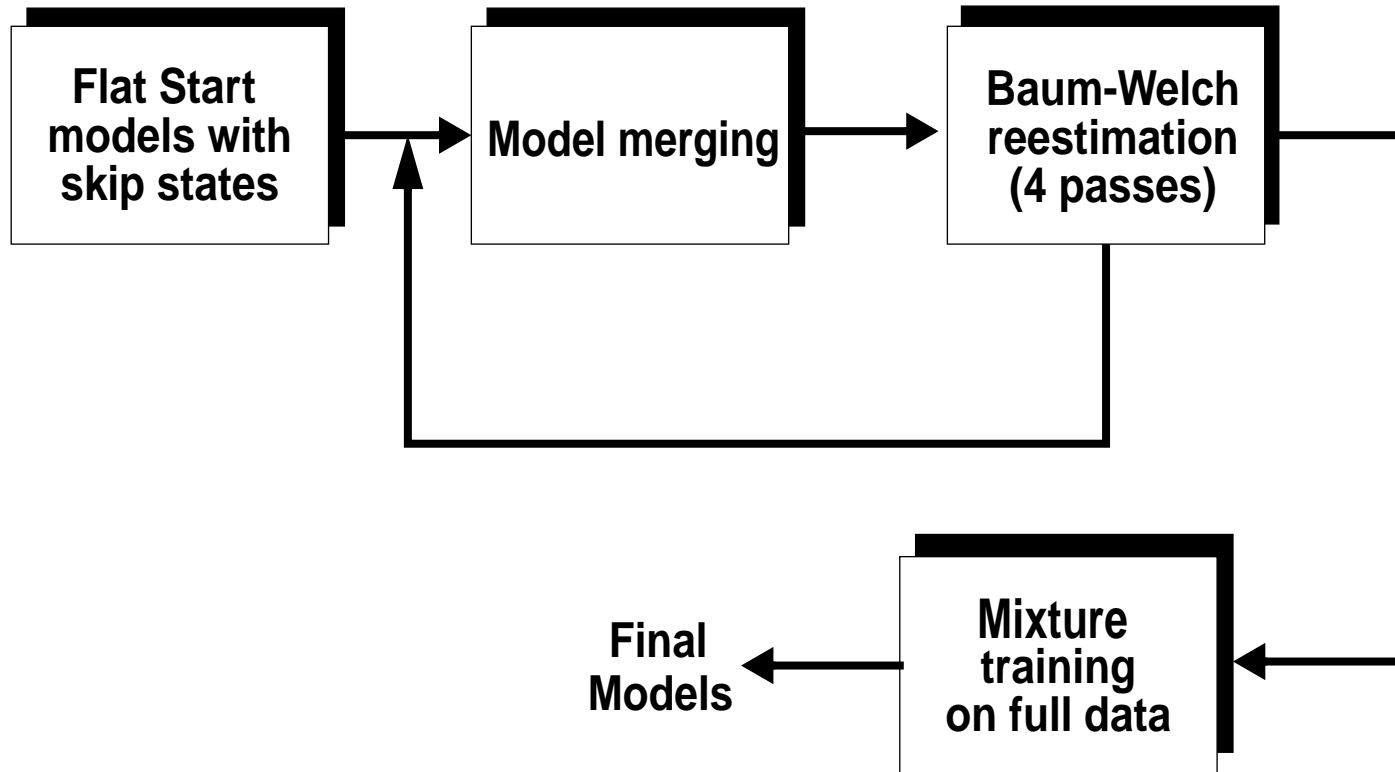


- ➡ Ideally states best used to model transitional phenomenon
- ➡ Merge states around stable spectral regions

OGI Alphadigits

- ➔ **Telephone database collected digitally using a T1 interface to the telephone network**
- ➔ **3000 subjects in the corpus**
- ➔ **19 or 29 alphanumeric strings per speaker**
- ➔ **Each utterance averages about six words in length (“8 H A 8 B H”, “8 W R W 8 E”)**
- ➔ **1102 unique prompting strings**
- ➔ **Balanced phonetic context of bigrams**

Methodology

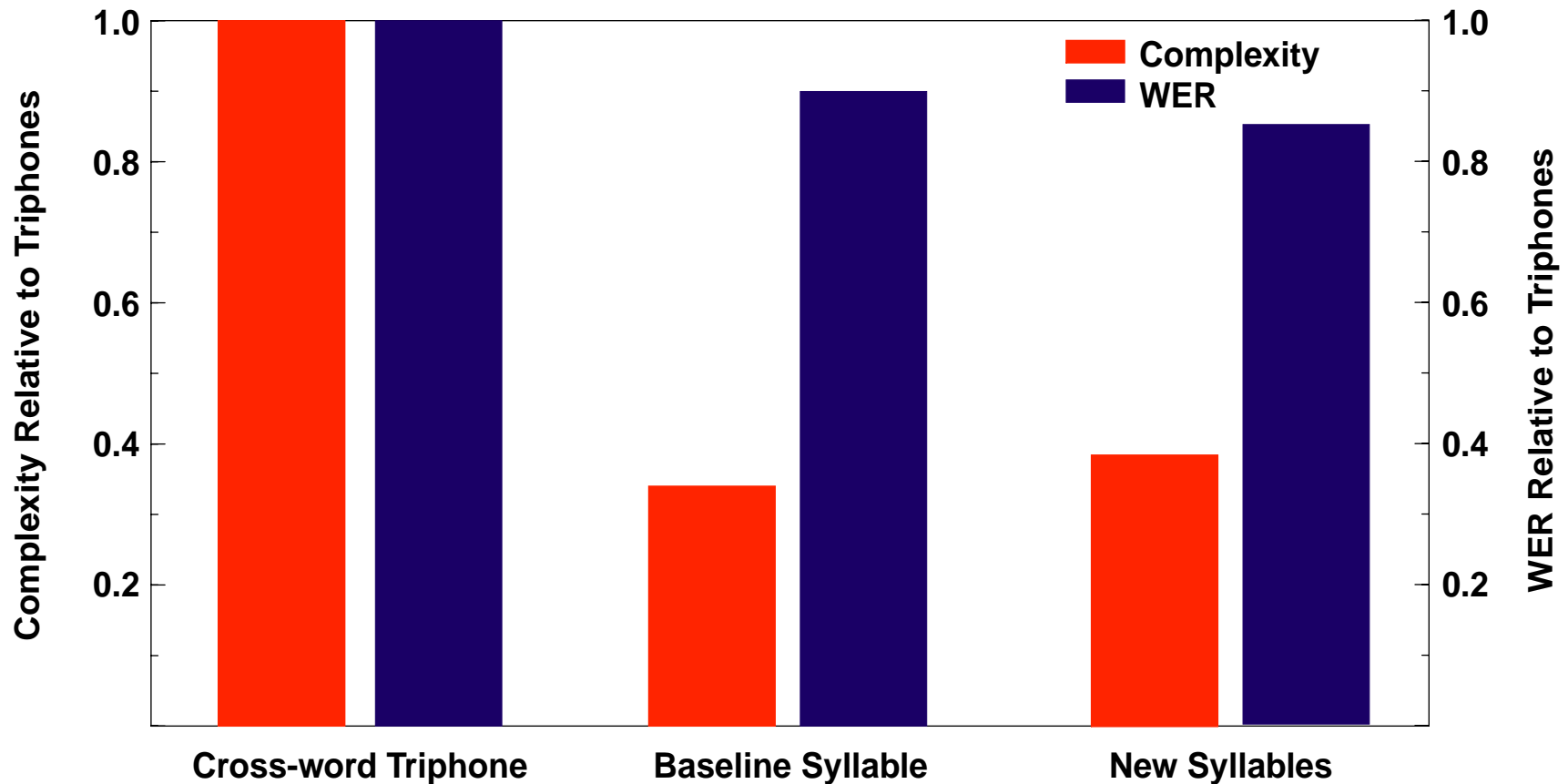


- ➡ **Minimum probability of error used for Bhattacharyya distance is 0.2**
- ➡ **Minimum distance used for KL measure is 0.1**

Effect on Model Topology

- ➡ **Initial models have number of states proportional to model duration (in frames)**
- ➡ **Before: models averaged 28 states/model**
- ➡ **After:**
 - Pass 1: average of 21 states/model**
 - Pass 2: average of 19 states/model**
- ➡ **Analysis shows that merges in the “stable” spectral portion of most models**

Results



➡ **Using Bhattacharyya distance, WER reduced from 10.4% to 9.9% in a syllable system**

➡ **Models 13% longer on an average in the new system (compensated skip states)**

Analysis

- ➡ **KL2 and Bhattacharyya distance metrics consistently give similar merges**
- ➡ **Convergence of some models slower than others, attributable to transitional formant structure**
- ➡ **Similar error modalities as previous syllable system**
- ➡ **Incorporation of skip states in models warrants higher word insertion penalty**

Summary

- ➡ **Information theoretic measures for optimal model size — Bhattacharyya Distance and KL2 Distance**
- ➡ **Significant improvement in performance: reduced WER from 10.4% to 9.9% (alphadigits)**
- ➡ **Need to determine impact of model merging on syllable-based SWB systems**
- ➡ **Explore approaches like BIC and MDL for more general topology selection**