

# PHONOTACTIC LANGUAGE RECOGNITION USING I-VECTORS AND PHONEME POSTERIogram COUNTS

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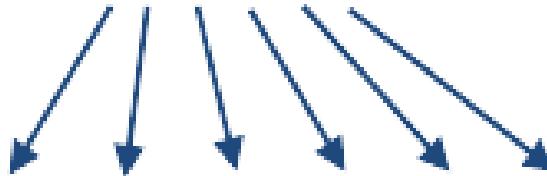
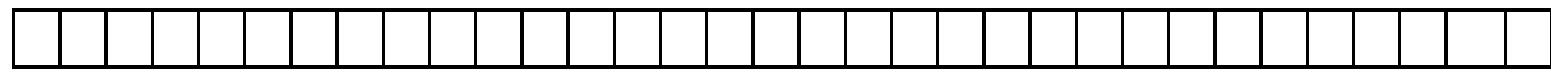
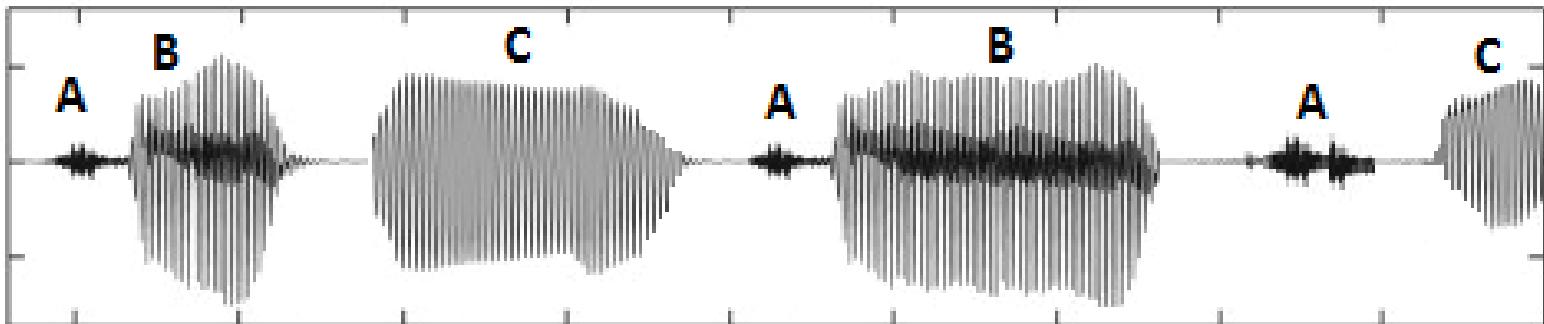
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# Introduction

- ◆ Current main approaches to LID
  - ◆ Acoustic-based: i-vectors, JFA, SVMs, or GMMs
  - ◆ Phonotactic
- ◆ Phonotactic systems:
  - ◆ PRLM, PPRLM: LMs created using different phonetic ASR
  - ◆ Lattice-based soft-counts: Created from phone lattices
    - ◆ Zero counts (i.e. data sparseness)
    - ◆ Limited by the number of phonemes and n-gram order
    - ◆ Dimensionality reduction: PCA or n-gram selection
  - ◆ Using i-vectors through Subspace Multinomial Models (SMMs)
    - ◆ We propose a new feature vector that performs better than soft-counts

# 1. Compute Posterior Probabilities



A	0.70	0.82	0.90	0.65	0.20	0.05
B	0.20	0.16	0.08	0.30	0.70	0.57
C	0.10	0.02	0.02	0.05	0.10	0.38

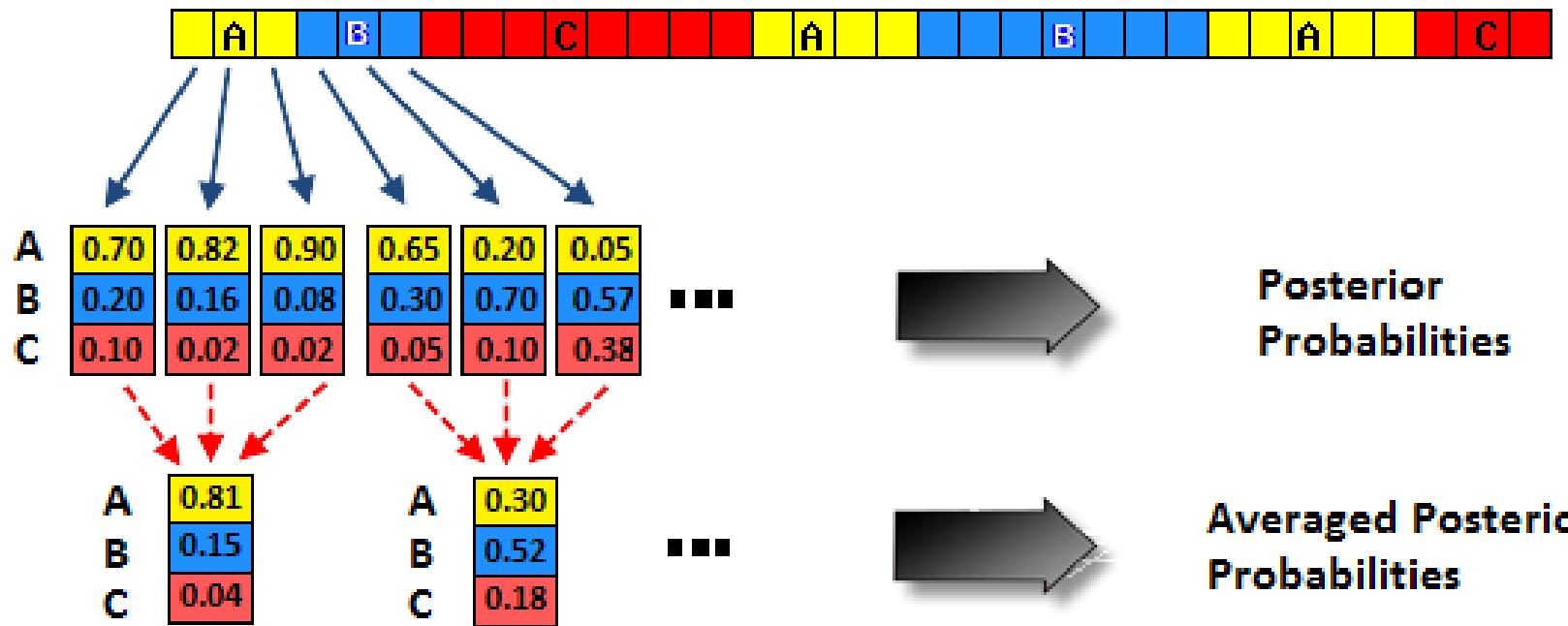
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Posterior  
Probabilities

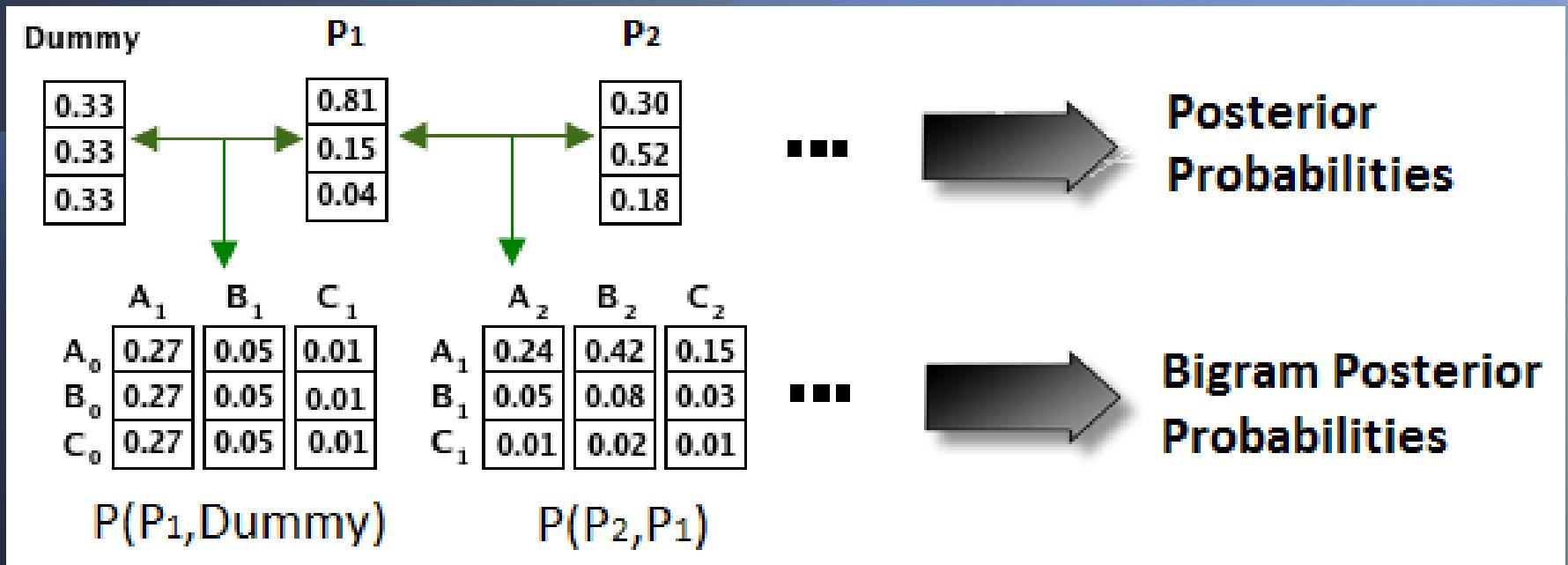
- In the example, we consider only three phonemes and bigrams. In our experiments, they were 33 and we used trigrams.

## 2. Compute Posterior Probs



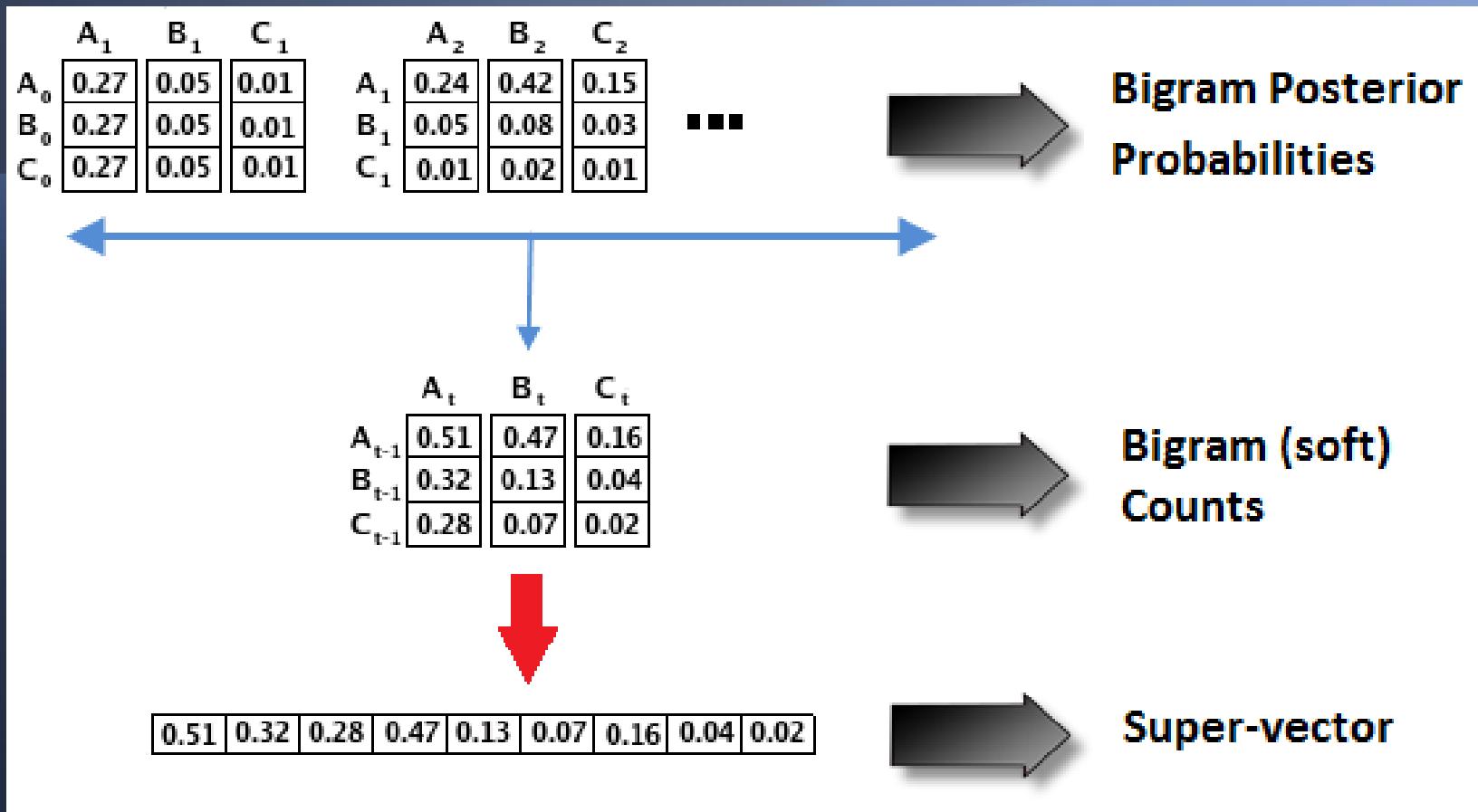
- Find the phoneme boundaries using Viterbi algorithm
  - Can be seen as incorporation of a-priori information
- Average the posterior probs over the phone boundaries
  - Smoothes the posterior probs and avoids the high-correlation of within-phoneme posteriors

### 3. Create N-gram Posterior Probs



- Outer product with the posteriogram of the previous phones
- Assume that the frames of the averaged posteriogram are statistically independent,
  - Therefore we have joint probabilities for sequences of phonemes

## 4. N-gram Counts via N-gram Posterior Probs



- ◆ Sum up all matrices to obtain n-gram soft counts
- ◆ Obtain feature super-vector for creating next the phonotactic i-vectors using SMMs

# Subspace Multinomial Models

- ◆ Allows extraction of low-dimensional vectors of coordinates in total variability subspace (i.e. i-vectors)
- ◆ The log-likelihood of data  $D$  for a multinomial model with  $C$  discrete events is determined by

$$\log p(D) = \sum_{n=1}^N \sum_{c=1}^C \gamma_{nc} \log \varphi_{nc}$$

- ◆ Where  $\gamma_{nc}$  is the count for n-gram event  $c$  at utterance  $n$ , and  $\varphi_{nc}$  is the probability of a multinomial distribution defined by the subspace model

# i-vectors from SMM

$$\varphi_{nc} = \frac{\exp(m_c + t_c w_n)}{\sum_i^c \exp(m_i + t_i w_n)}$$

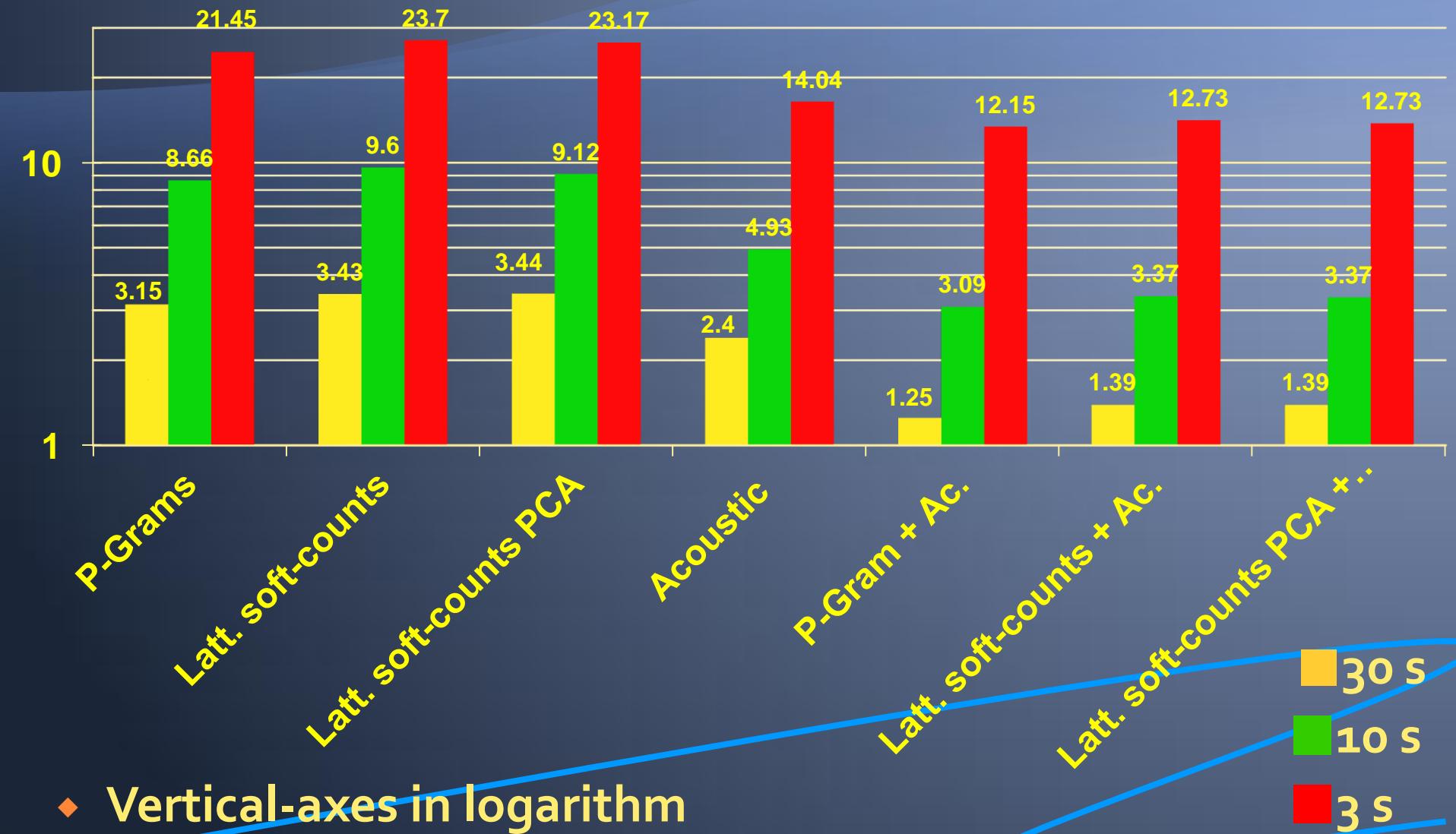
- Where  $t_c$  is the c-th row of subspace matrix T (Extractor), and  $w_n$  is the i-vector

- An i-vector for a single utterance is estimated numerically by maximizing the likelihood (ML)
- Matrix T is trained numerically using ML by iteratively optimizing T and re-estimating the i-vectors for all training utterances
- Then we use these i-vectors as feature input for training a discriminative LID classifier
  - Multiclass logistic regression

# Experimental Setup

- ◆ NIST LRE 2009 database
  - ◆ 50K segments for training (~119h), 38K segments for dev (~153h) and 31K sentences for test (~125h)
  - ◆ 23 languages, test on 3, 10, and 30 s conditions
  - ◆ Results given using  $C_{avg}$  metric
- ◆ Acoustic i-vector system
  - ◆ 7 MFCC + 49 SDCs, CMN/CVN, 2048 Gaussians -> i-vectors of 400 dimensions
- ◆ Comparisons with:
  - ◆ Lattice-based soft-counts with i-vectors (size 600)
  - ◆ Lattice-based soft-counts with PCA (reduction to 1000 dimensions)
- ◆ Fusion: Multiclass logistic regression
  - ◆ Acoustic and Phonotactic

# Results on NIST LRE 2009



# Conclusions and Future Work

- ◆ Advantages of the new features
  - Avoid data sparseness (i.e. robustness)
  - Results outperforms a similar system based on lattice soft-counts with i-vectors
    - 8,16% relative on 30 s condition
  - Fusion with acoustic i-vectors are also better
    - 10% relative on 30 s condition
- ◆ Future Work: Apply discriminative n-gram selection techniques to reduce the vector size
  - Avoids low frequency n-gram counts
  - Allows using high n-gram orders

...Thanks for your  
attention...

Comments or Questions?