Improved Surname Pronunciation Using Decision Trees



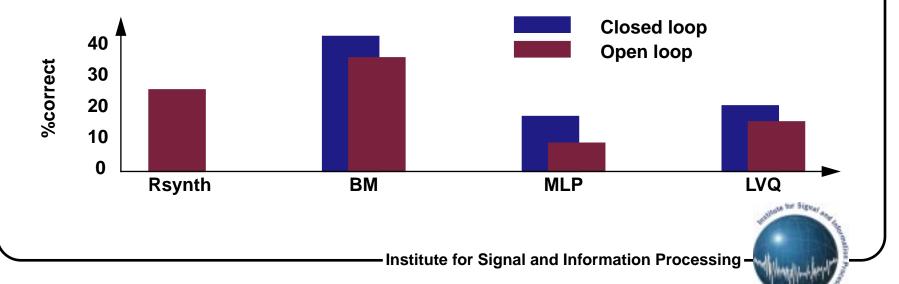


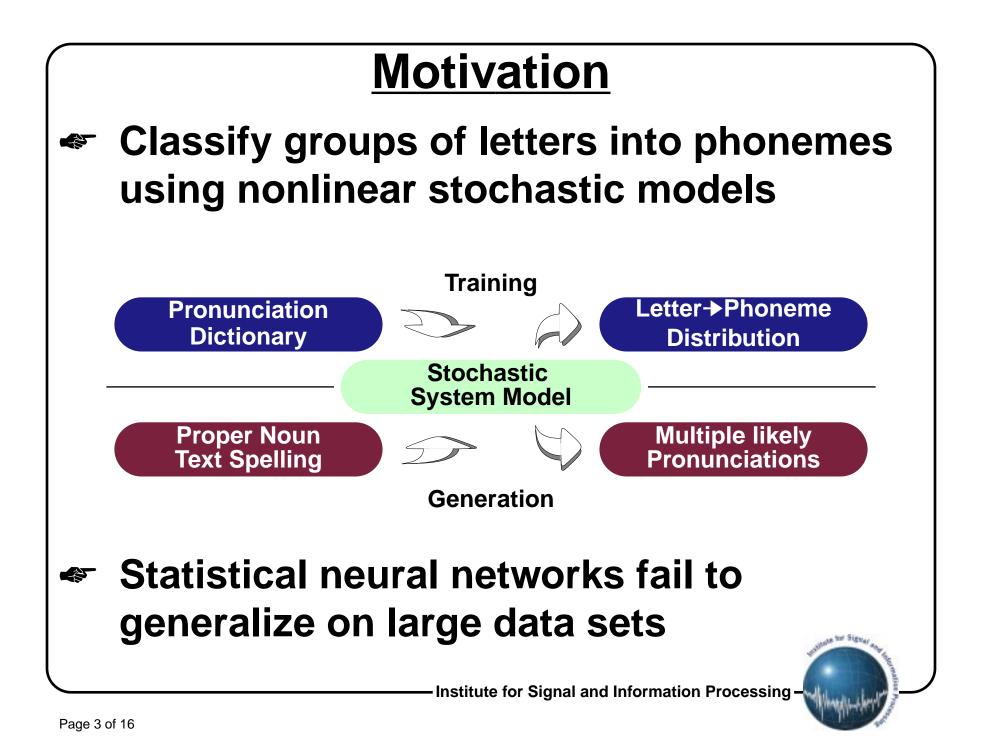
J. Ngan, N. Deshmukh, A. Ganapathiraju, and J. Picone Institute for Signal and Information Processing Mississippi State University {ngan, deshmukh, ganapath, picone}@isip.msstate.edu http://www.isip.msstate.edu



Introduction

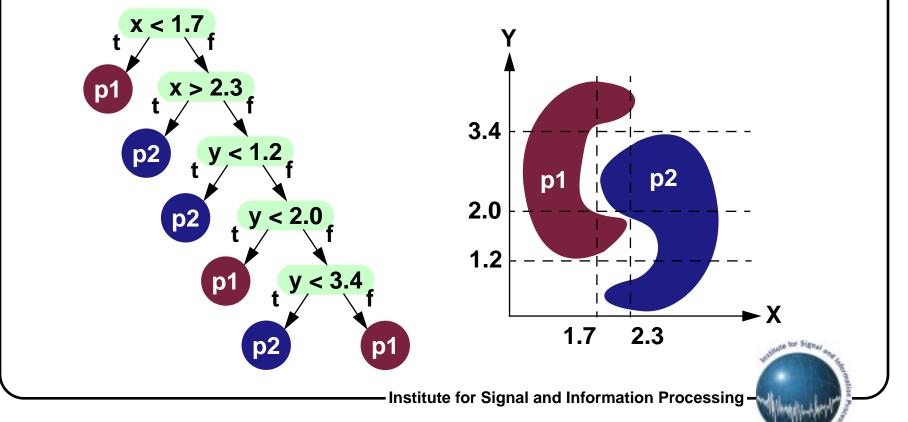
- Proper nouns and errors in LVCSR
- Pronunciation networks needed
- Direct letter-to-sound rules do not apply
- Rule-based systems are unsuitable,
 NN-based systems do not work

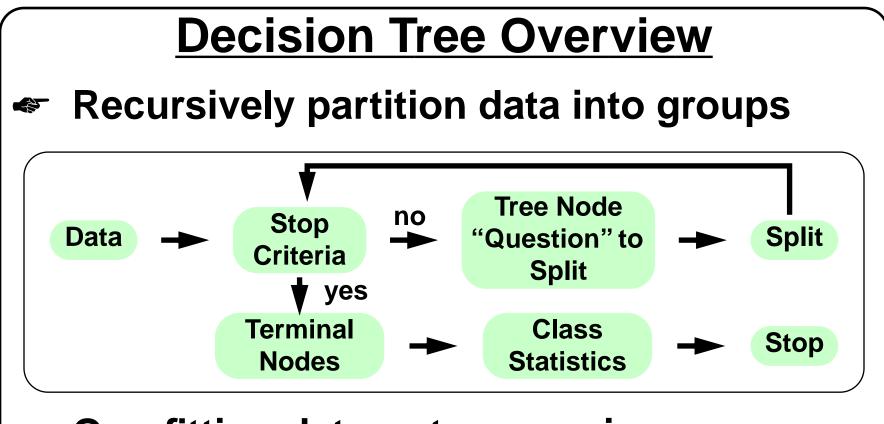




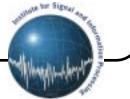
Decision Tree Classifiers

- Combine heuristics and statistics
- Ideally suited for nonlinear classification
- Classification based on data attributes





- Overfitting data tree pruning
- Information theoretic basis for splitting, stopping, and pruning
- Capture complex relationships



DT Splitting Criteria

- Used to design "questions" at each node
- Bayesian splitting
 - maximize a posteriori class likelihoods
- Information gain splitting
 - maximize the information gain
- Information gain ratio splitting

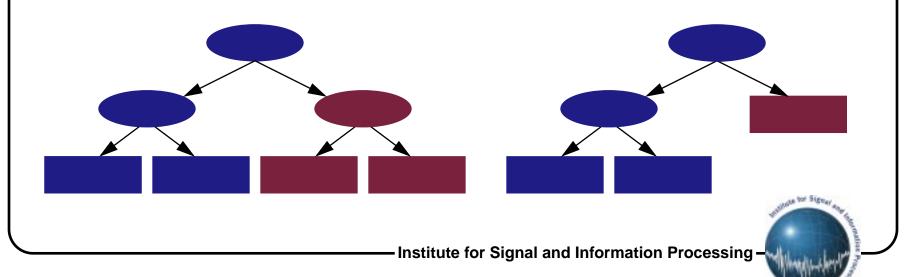
— maximize information gain normalized by the total information in split

Maximum entropy splitting



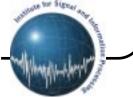
DT Pruning Criteria

- Collapse subtree into a terminal node if it results in lower predicted error
- Cost-complexity trade off tree size and error rate
- Pessimistic pruning statistical error estimates at node adjusted as per bias



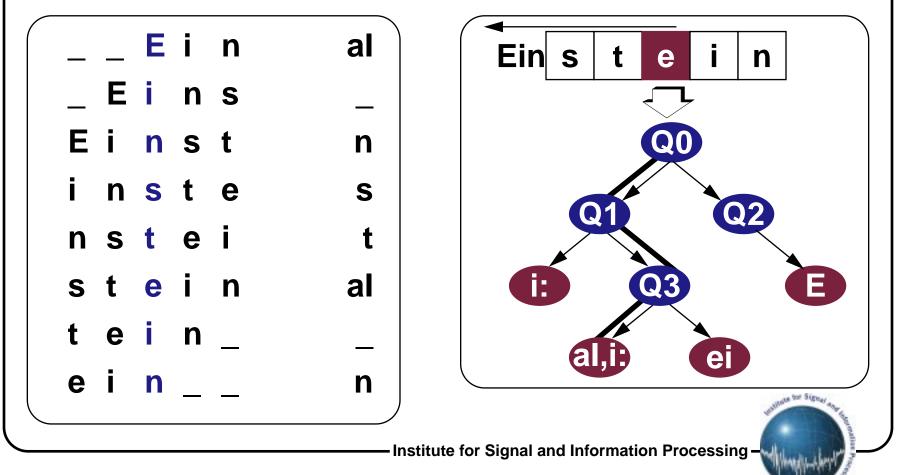
ISIP DT Toolkit

- Limitations of existing free software
 - small number of classes
 - number and values of attributes
 - restrictions on class labels
 - not amenable to text-based processing
- ISIP DT toolkit allows user-defined criteria for splitting, pruning, and stopping
- Data tagging allows selective attribute usage without reformatting



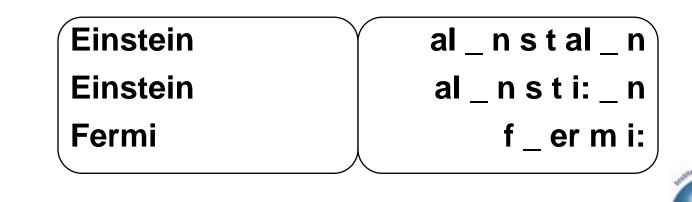
Surname Pronunciations

Train DT on name-pronunciation pairs
 Sliding window of letter context



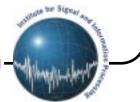
Pronunciation Dictionary

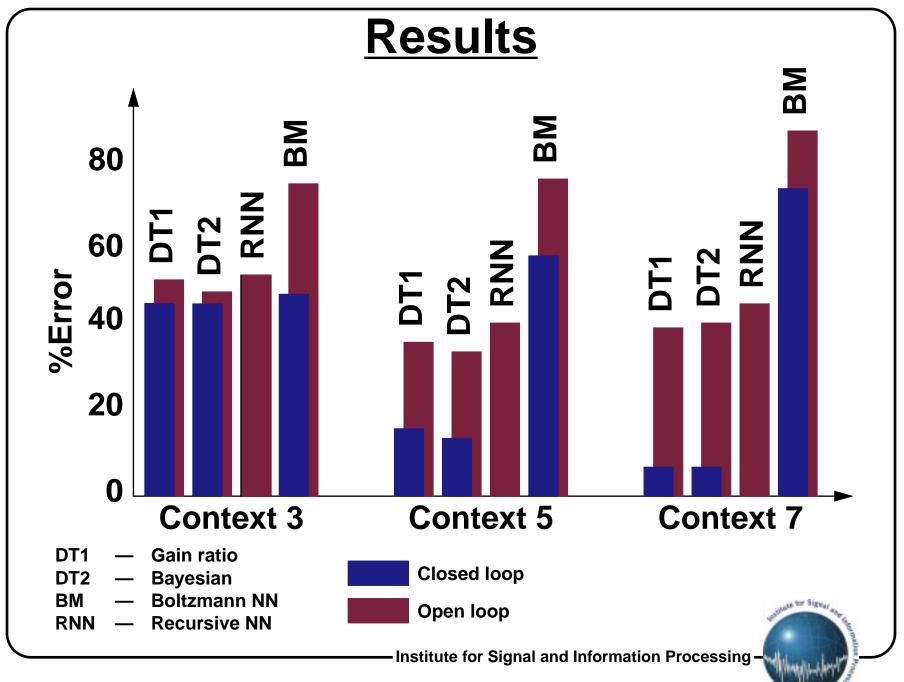
- 18494 surnames, 25648 pronunciations
- Worldbet phonetic convention to handle multilingual data
- Hand-transcribed pronunciations
- Automatic letter-to-phoneme alignment
- Public domain resource



Experiments

- Training set of 15000 names (3494 names held out for testing)
- Three such partitions for cross-validation of results
- Context lengths of 3, 5 and 7
- Binary univariate tree, single (1-best) pronunciation output
- Bayesian and gain ratio splitting
- Pessimistic pruning





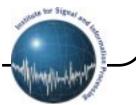
<u>Analysis</u>

- DTs try to model letter context-to-sound relationship
- As context size increases
 - closed loop performance improves
 - open loop performance decreases due to overfitting
- Extremely confusible classification space
- Performance of both splitting criteria is comparable
- Pruning only helps marginally



General English

- Used best surname-trained DT
- 8000 word subset from Switchboard lexicon as test data
- Word error rate 83%!!!
- Pruning makes no difference
- Letter-to-sound mapping for names is radically different from general English
- Issue can a single tree be trained for both types of data?



Conclusions

- Decision tree systems perform better 38% error
- Generation of multiple pronunciations will further improve error rate
- Highly nonlinear letter-to-phone maps need more data for effective training
- Future plans extend DT application to general English
- Public domain resources DT toolkit, pronunciation dictionary

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References

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