

Introduction to HTK Toolkit

Berlin Chen 2004

Reference:

- Steve Young et al. *The HTK Book*. Version 3.2, 2002.

Outline

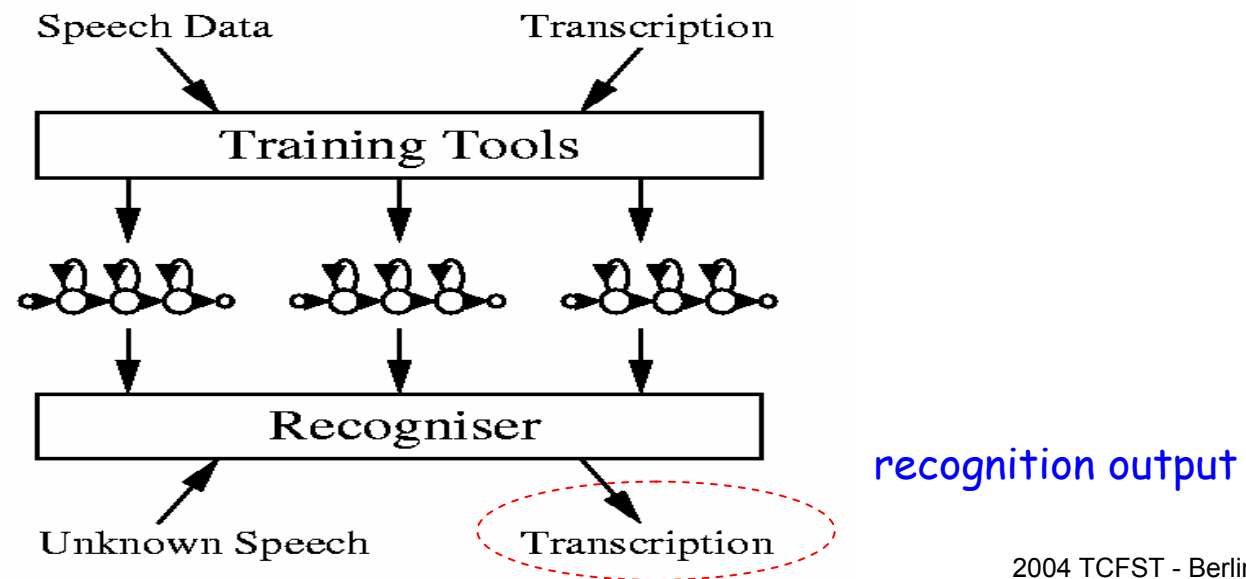
- An Overview of HTK
- HTK Processing Stages
- Data Preparation Tools
- Training Tools
- Testing Tools
- Analysis Tools
- Homework: Exercises on HTK

An Overview of HTK

- HTK: A toolkit for building Hidden Markov Models
- HMMs can be used to model any time series and the core of HTK is similarly general-purpose
- HTK is primarily designed for building HMM-based speech processing tools, in particular speech recognizers

An Overview of HTK (cont.)

- Two major processing stages involved in HTK
 - **Training Phase:** The training tools are used to estimate the parameters of a set of HMMs using training utterances and their associated transcriptions
 - **Recognition Phase:** Unknown utterances are transcribed using the HTK recognition tools

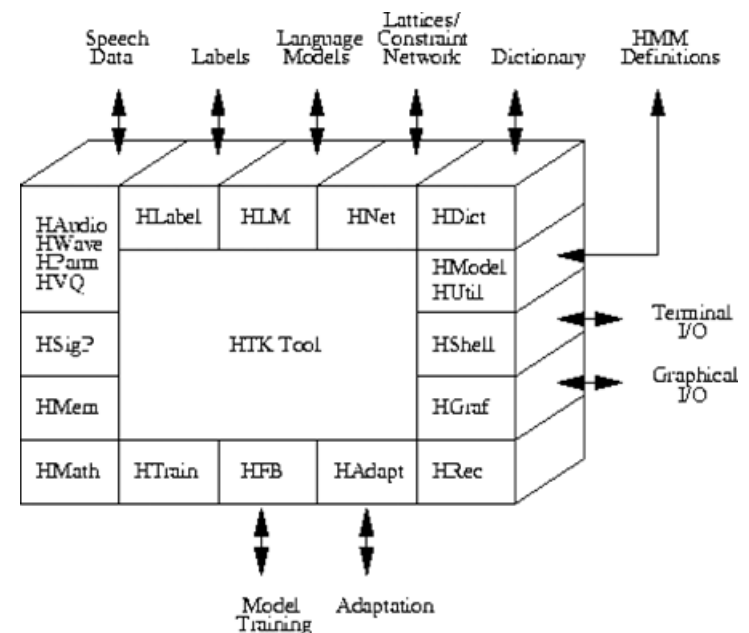


An Overview of HTK (cont.)

- HTK Software Architecture
 - Much of the functionality of HTK is built into the library modules
 - Ensure that every tool interfaces to the outside world in exactly the same way
- Generic Properties of an HTK Tools
 - HTK tools are designed to run with a traditional **command line style interface**

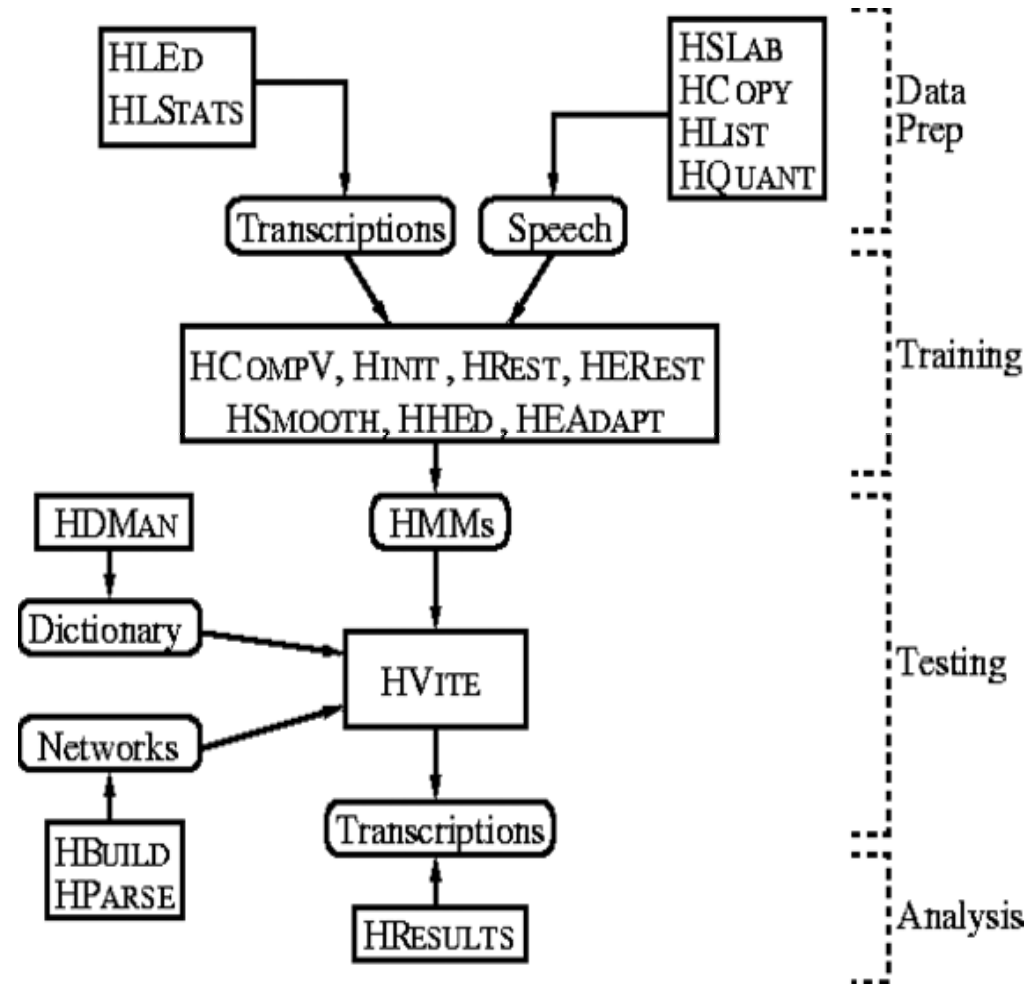
HFoo -T -C Config1 -f 34.3 -a -s myfile file1 file2

- The main use of configuration files is to control the detailed behavior of the library modules on which all HTK tools depend



HTK Processing Stages

- Data Preparation
- Training
- Testing/Recognition
- Analysis



Data Preparation Phase

- In order to build a set of HMMs for acoustic modeling, a set of speech data files and their associated transcriptions are required
 - Convert the speech data files into an appropriate **parametric format** (or the appropriate acoustic feature format)
 - Convert the associated transcriptions of the speech data files into an appropriate format which consists of the required **phone or word labels**
- **HSLAB**
 - Used both to record the speech and to manually annotate it with any required transcriptions if the speech needs to be recorded or its transcriptions need to be built or modified

Data Preparation Phase (cont.)

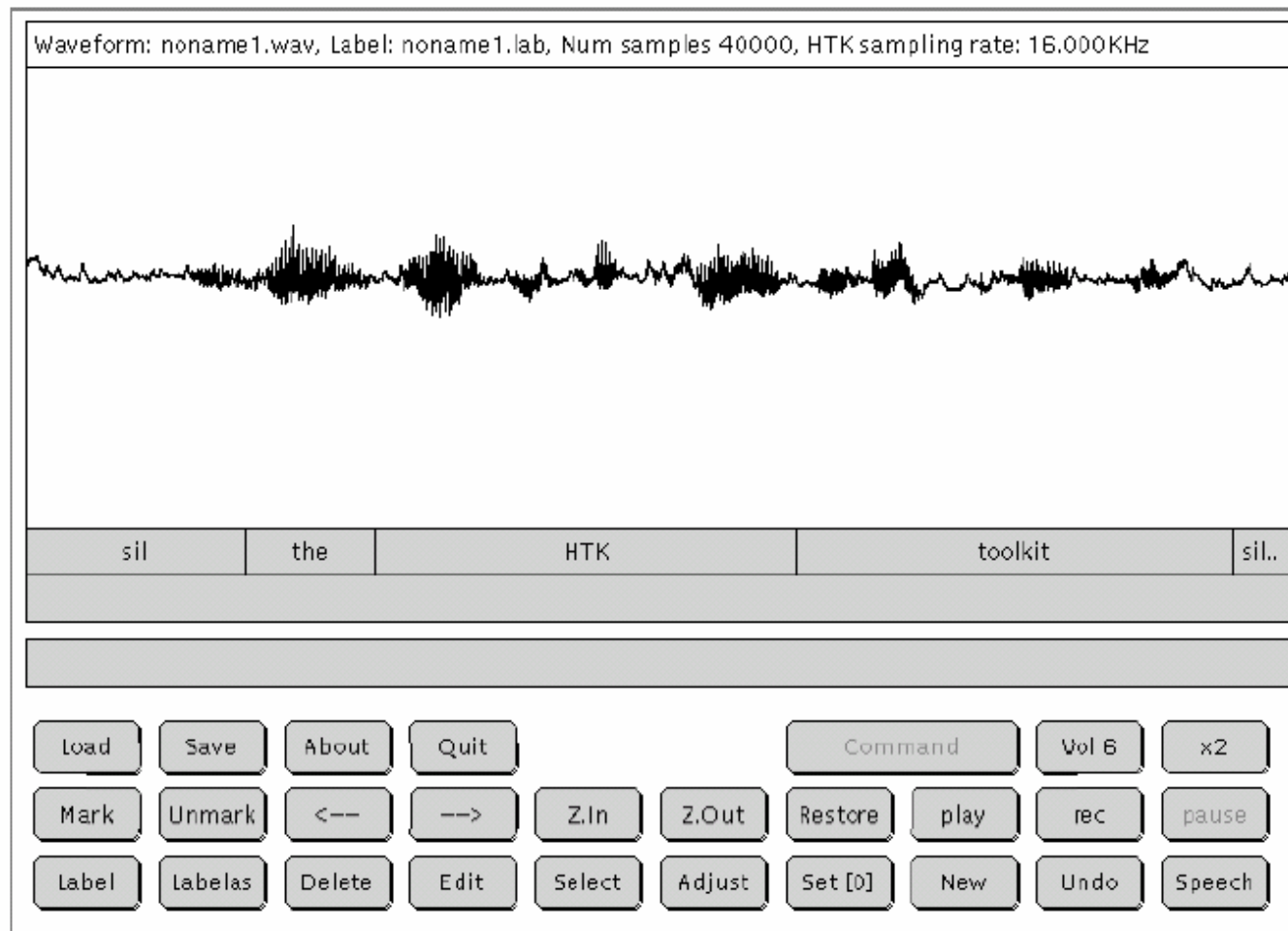
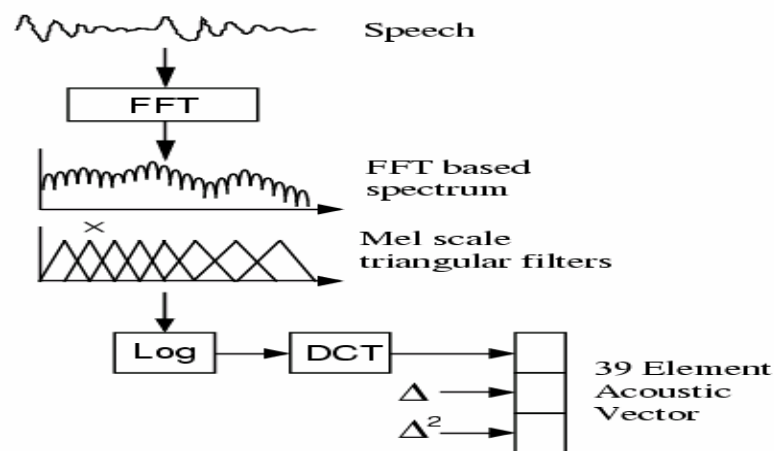


Fig. 14.1 HSLab display window

Data Preparation Phase (cont.)

- *HCOPY*

- Used to parameterize the speech waveforms to a variety of acoustic feature formats by setting the appropriate configuration variables



MFCC-based Front-End Processor.

LPC	linear prediction filter coefficients
LPCREFC	linear prediction reflection coefficients
LPCEPSTRA	LPC cepstral coefficients
LPDELCEP	LPC cepstra plus delta coefficients
MFCC	mel-frequency cepstral coefficients
MELSPEC	linear mel-filter bank channel outputs
DISCRETE	vector quantized data

Data Preparation Phase (cont.)

- *HLIST*
 - Used to check the contents of any speech file as well as the results of any conversions before processing large quantities of speech data
- *HLED*
 - A script-driven text editor used to make the required transformations to label files, for example, the generation of context-dependent label files
- *HLSTATS*
 - Used to gather and display statistical information for the label files
- *HQUANT*
 - Used to build a VQ codebook in preparation for build discrete probability HMM systems

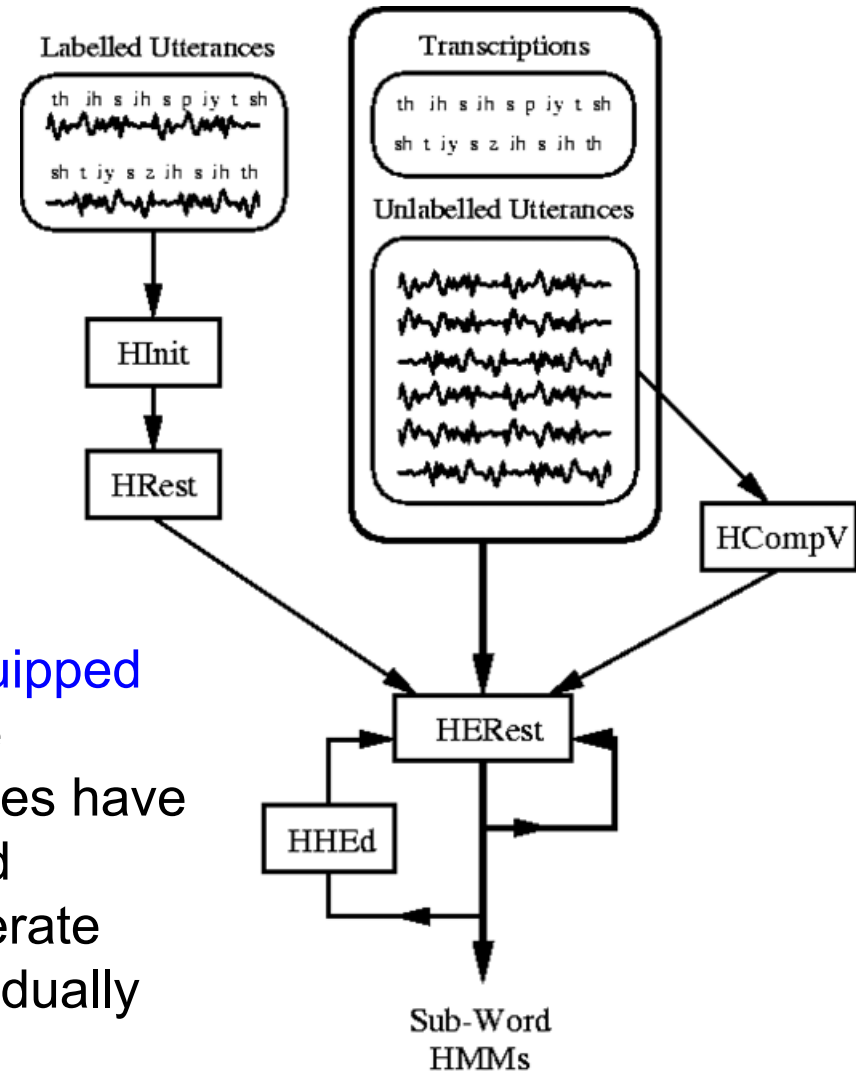
Training Phase

- Prototype HMMs
 - Define the topology required for each HMM by writing a prototype Definition
 - HTK allows HMMs to be built with any desired topology
 - HMM definitions stored as simple text files
 - All of the HMM parameters (the *means* and *variances* of Gaussian distributions) given in the prototype definition are ignored only with exception of the *transition* probability

```
~o <VecSize> 39 <MFCC_0_D_A>
~h "proto"
<BeginHMM>
  <NumStates> 5
  <State> 2
    <Mean> 39
      0.0 0.0 0.0 ...
    <Variance> 39
      1.0 1.0 1.0 ...
  <State> 3
    <Mean> 39
      0.0 0.0 0.0 ...
    <Variance> 39
      1.0 1.0 1.0 ...
  <State> 4
    <Mean> 39
      0.0 0.0 0.0 ...
    <Variance> 39
      1.0 1.0 1.0 ...
  <TransP> 5
    0.0 1.0 0.0 0.0 0.0
    0.0 0.6 0.4 0.0 0.0
    0.0 0.0 0.6 0.4 0.0
    0.0 0.0 0.0 0.7 0.3
    0.0 0.0 0.0 0.0 0.0
<EndHMM>
```

Training Phase (cont.)

- There are **two different versions for acoustic model training** which depend on whether the sub-word-level (e.g. the phone-level) boundary information exists in the transcription files or not



- If the training speech files are equipped the sub-word boundaries, i.e., the location of the sub-word boundaries have been marked, the tools **HINIT** and **HREST** can be used to train/generate each sub-word HMM model individually with all the speech training data

Training Phase (cont.)

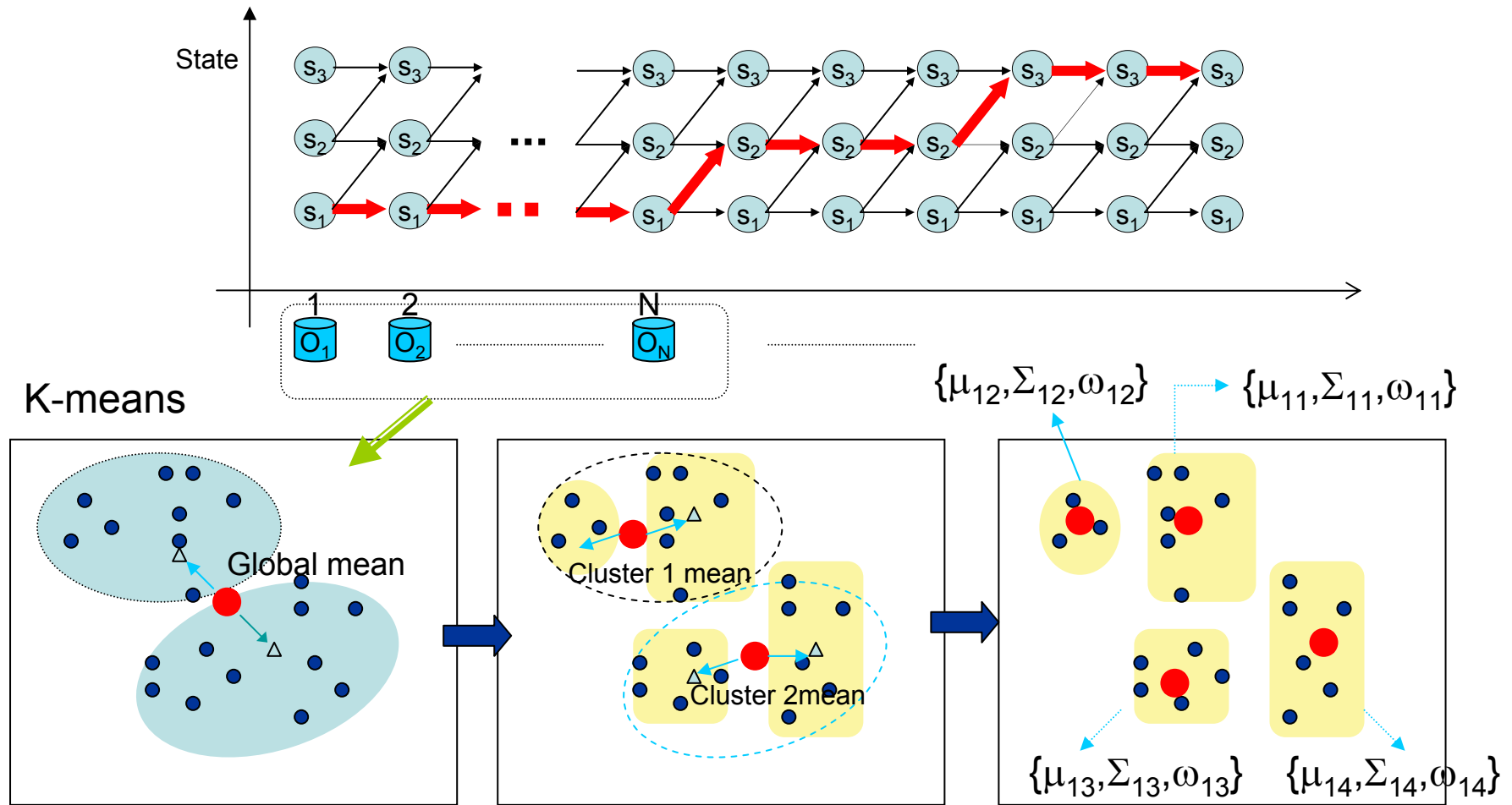
- *HINIT*

- Iteratively computes an initial set of parameter value using the *segmental k-means training* procedure
 - It reads in all of the bootstrap training data and cuts out all of the examples of a specific phone
 - On the first iteration cycle, the training data are uniformly segmented with respect to its model state sequence, and each model state matching with the corresponding data segments and then means and variances are estimated. *If mixture Gaussian models are being trained, then a modified form of k-means clustering is used*
 - *On the second and successive iteration cycles, the uniform segmentation is replaced by Viterbi alignment*

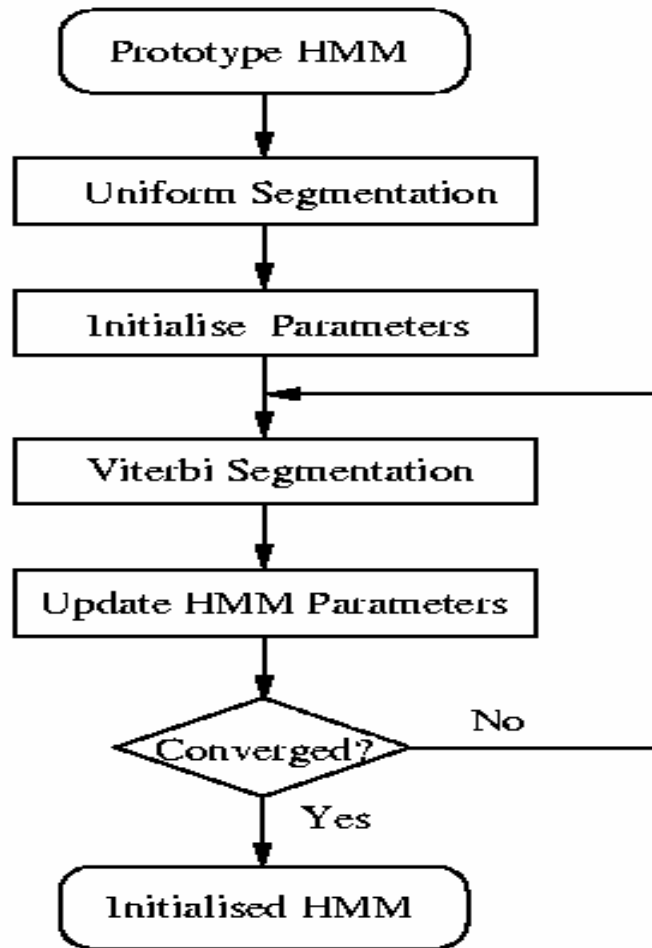
- *HREST*

- Used to further re-estimate the HMM parameters initially computed by *HINIT*
- *Baum-Welch* re-estimation procedure is used, instead of the segmental k-means training procedure for *HINIT*

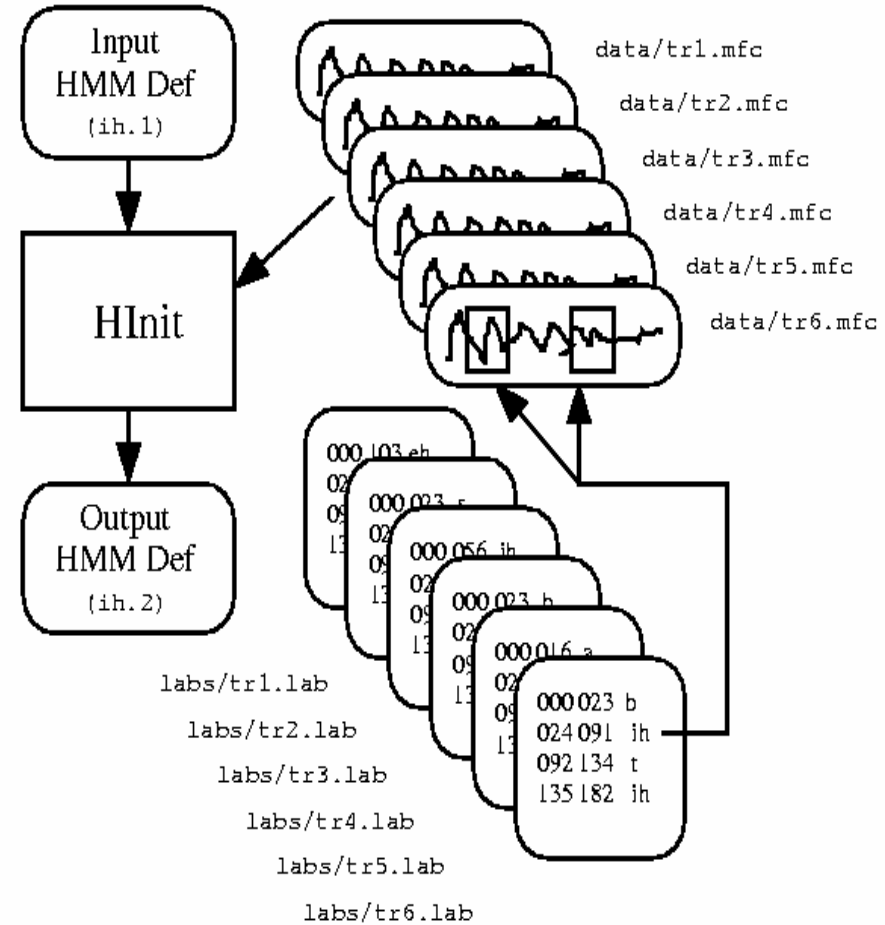
Training Phase (cont.)



Training Phase (cont.)

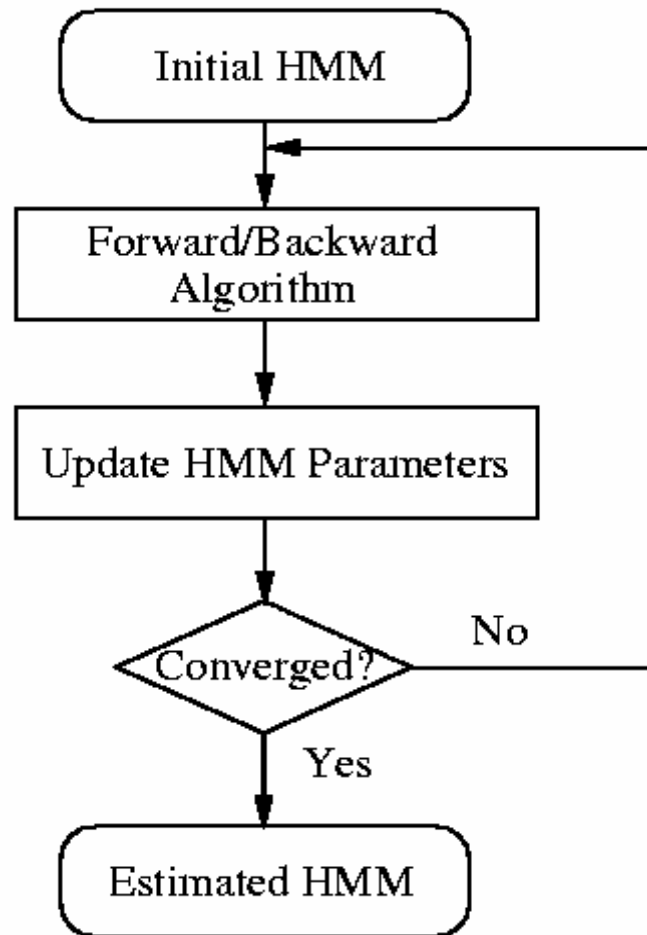


HInit Operation



File Processing in HInit

Training Phase (cont.)

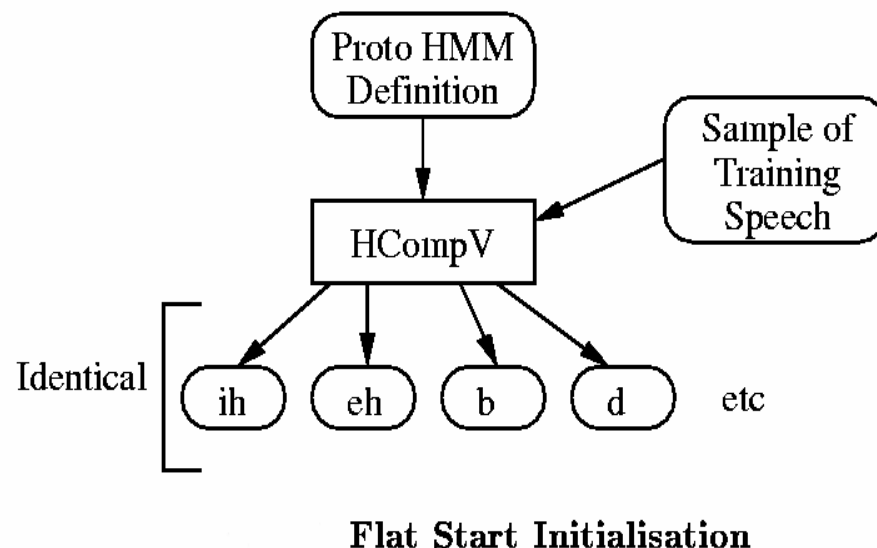


HRest Operation

Training Phase (cont.)

- On the other hand, if the training speech files are not equipped the sub-word-level boundary information, a so-called *flat-start* training scheme can be used
 - In this case all of the phone models are initialized to be identical and have state means and variances equal to the global speech mean and variance. The tool *HCOMPV* can be used for this
- *HCOMPV*
 - Used to calculate the global mean and variance of a set of training data

Training Phase (cont.)

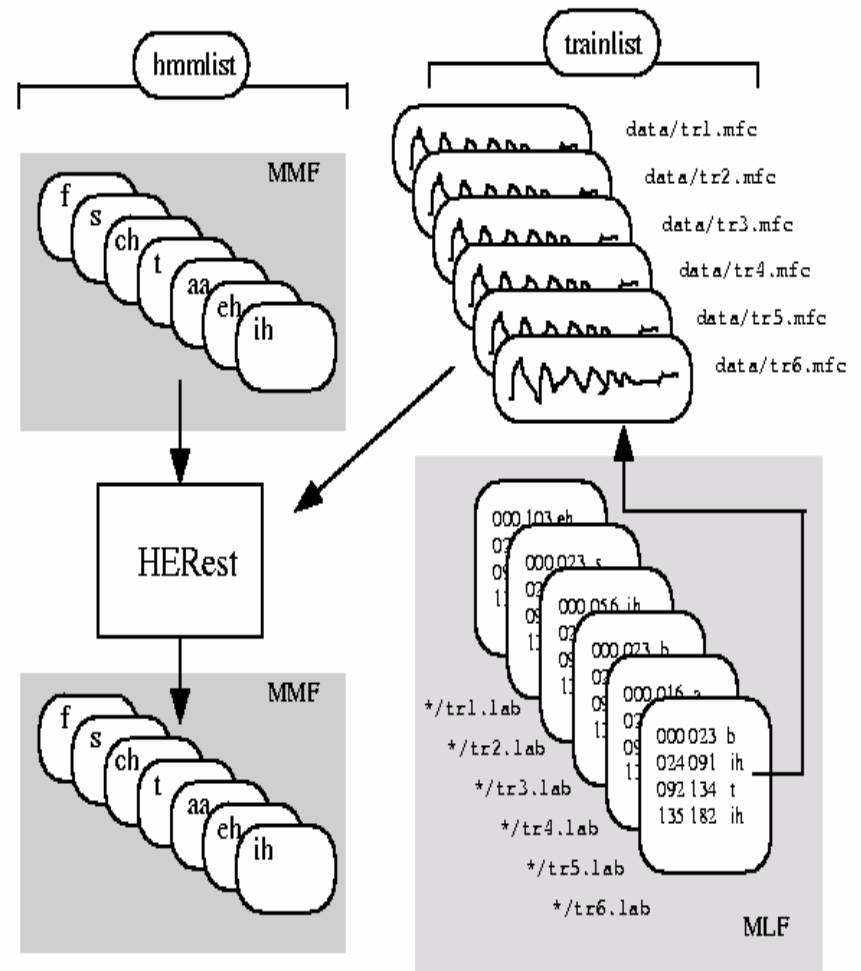


- Once the initial parameter set of HMMs has been created by either one of the two versions mentioned above, the tool *HEREST* is further used to perform *embedded training* on the whole set of the HMMs simultaneously using the entire training set

Training Phase (cont.)

- **HEREST**

- Performs a single *Baum-Welch* re-estimation of the whole set of the HMMs simultaneously
 - For each training utterance, the corresponding phone models are concatenated and the forward-backward algorithm is used to accumulate the statistics of state occupation, means, variances, etc. for each HMM in the sequence
 - When all of the training utterances has been processed, the accumulated statistics are used to re-estimate the HMM parameters
- *HEREST* is the *core* HTK training tool



File Processing in HERest

Training Phase (cont.)

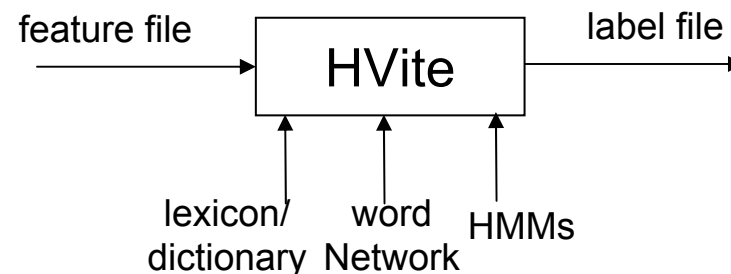
- Model Refinement

- The philosophy of system construction in HTK is that HMMs should be refined incrementally
- **CI to CD:** A typical progression is to start with a simple set of single Gaussian context-independent phone models and then iteratively refine them by expanding them to include context-dependency and use multiple mixture component Gaussian distributions



- **Tying:** The tool *HHED* is a HMM definition editor which will clone models into context-dependent sets, apply a variety of parameter tyings and increase the number of mixture components in specified distributions
- **Adaptation:** To improve performance for specific speakers the tools *HEADAPT* and *HVITE* can be used to adapt HMMs to better model the characteristics of particular speakers using a small amount of training or adaptation data

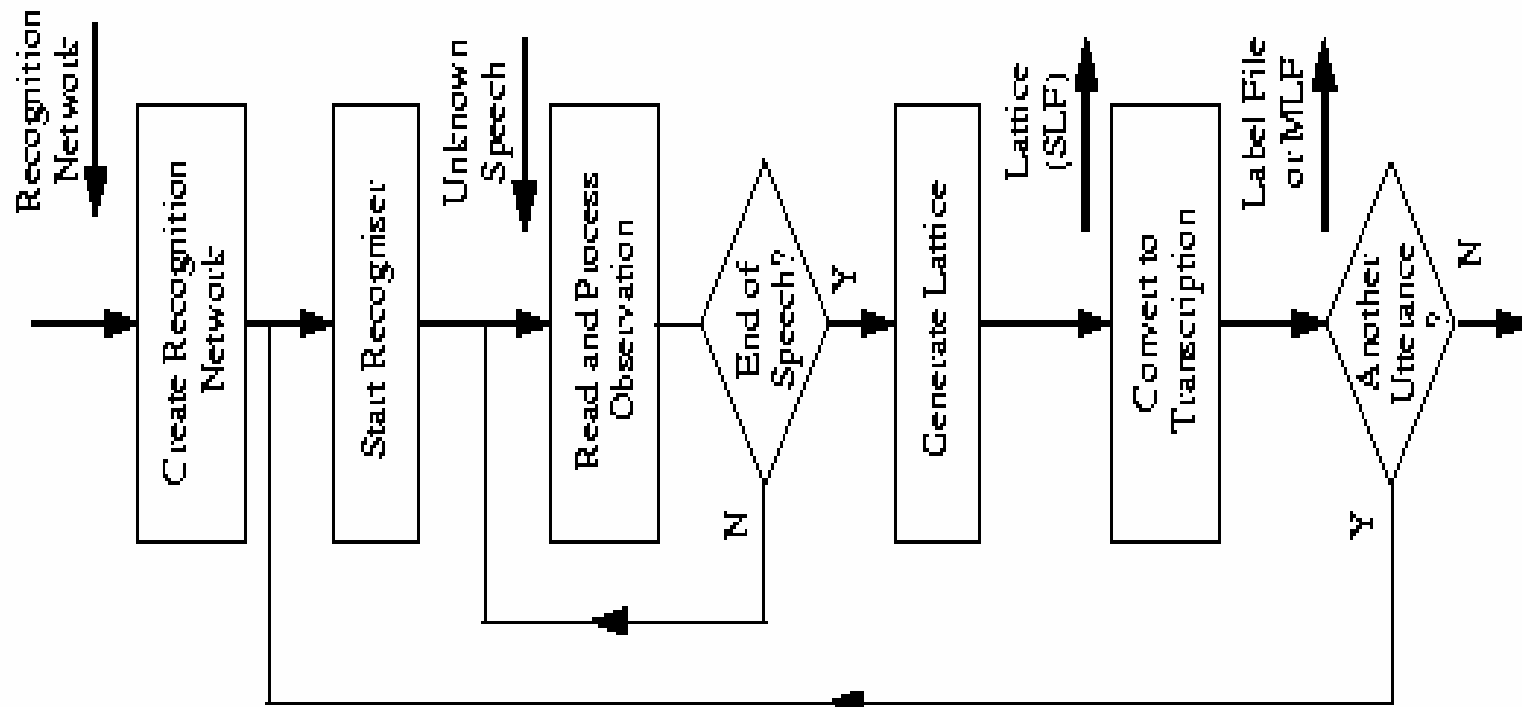
Recognition Phase



- *HVITE*

- Performs Viterbi-based speech recognition
- Takes a network describing the allowable word sequences, a dictionary defining how each word is pronounced and a set of HMMs as inputs
- Supports cross-word triphones, also can run with multiple tokens to generate lattices containing multiple hypotheses
- Also can be configured to rescore lattices and perform forced alignments
- The word networks needed to drive *HVITE* are usually either simple word loops in which any word can follow any other word or they are directed graphs representing a finite-state task grammar
 - *HBUILD* and *HPARSE* are supplied to create the word networks

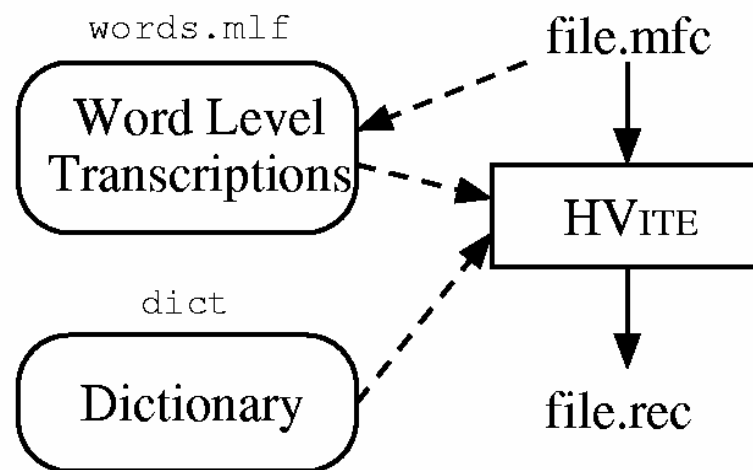
Recognition Phase (cont.)



Recognition Processing

Recognition Phase (cont.)

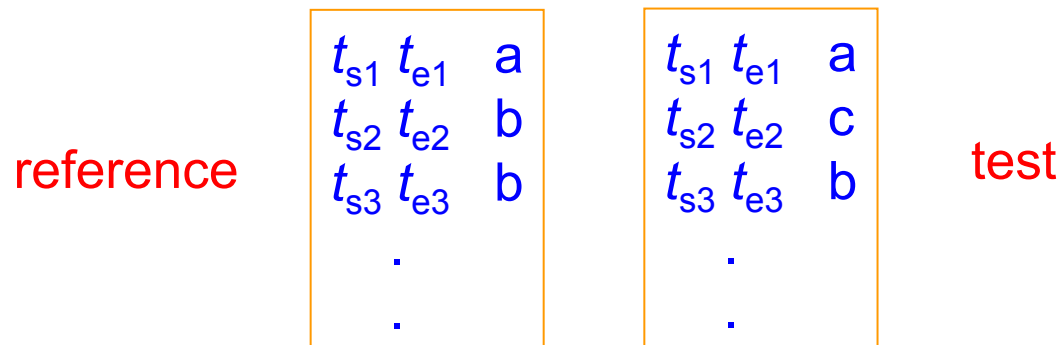
- Generating Forced Alignment
 - HVite computes a new network for each input utterance using the word level transcriptions and a dictionary
 - By default the output transcription will just contain the words and their boundaries. One of the main uses of forced alignment, however, is to determine the actual pronunciations used in the utterances used to train the HMM system



Forced Alignment

Analysis Phase

- The final stage of the HTK Toolkit is the analysis stage
 - When the HMM-based recognizer has been built, it is necessary to evaluate its performance by comparing the recognition results with the correct reference transcriptions. An analysis tool called *HRESULTS* is used for this purpose
- *HRESULTS*
 - Performs the comparison of recognition results and correct reference transcriptions by using dynamic programming to align them
 - The assessment criteria of *HRESULTS* are compatible with those used by the US *National Institute of Standards and Technology (NIST)*



A Tutorial Example

- A Voice-operated interface for phone dialing

Dial three three two six five four

Dial nine zero four one oh nine

Phone Woodland

Call Steve Young

regular expression

– \$digit = ONE | TWO | THREE | FOUR | FIVE |

SIX | SEVEN | EIGHT | NINE | OH | ZERO;

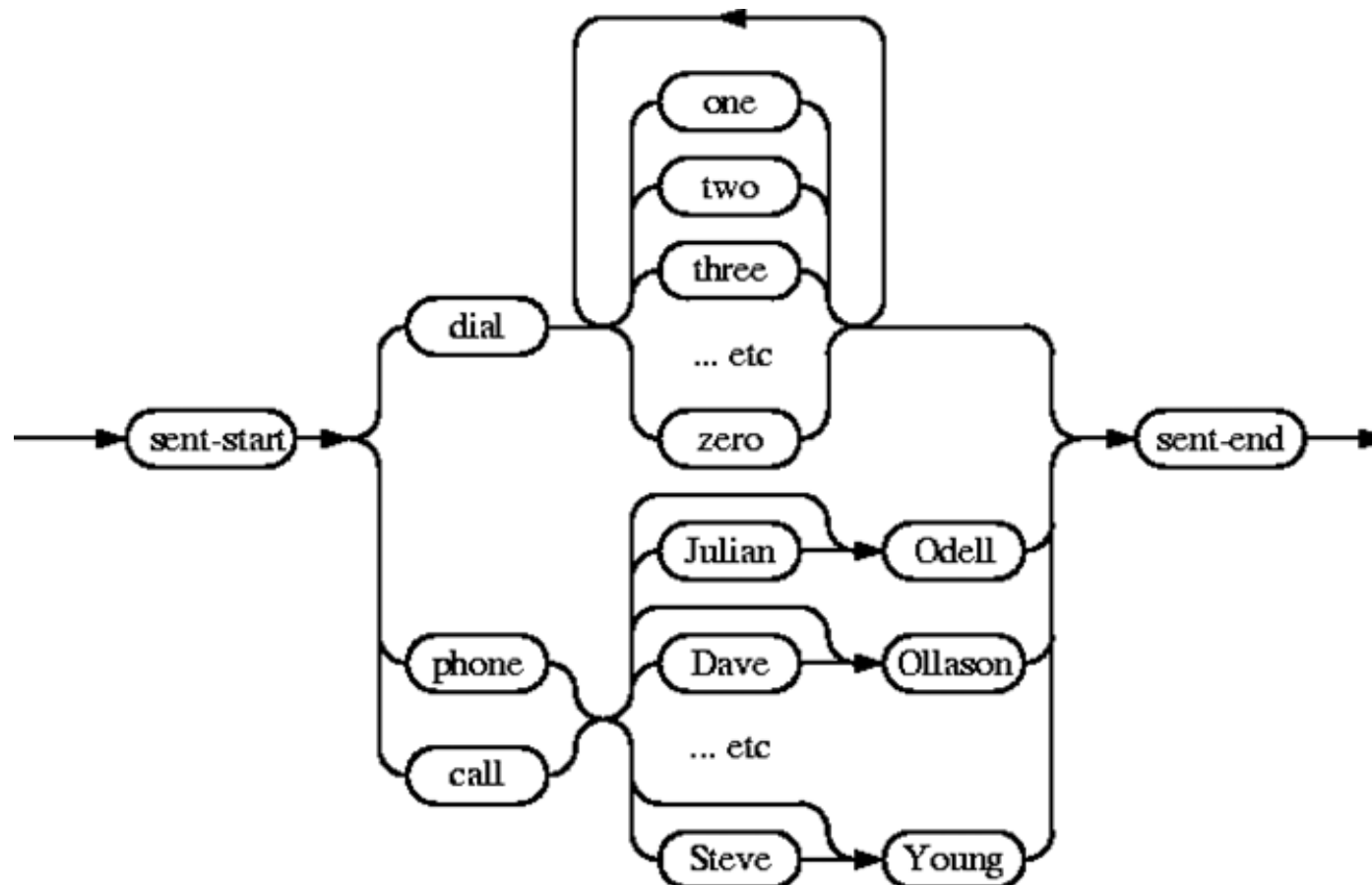
\$name = [JOOP] JANSEN | [JULIAN] ODELL | [DAVE]

OLLASON | [PHIL] WOODLAND | [STEVE] YOUNG;

(SENT-START (DIAL <\$digit> | (PHONE|CALL) \$name) SENT-END)

Grammar for Voice Dialing

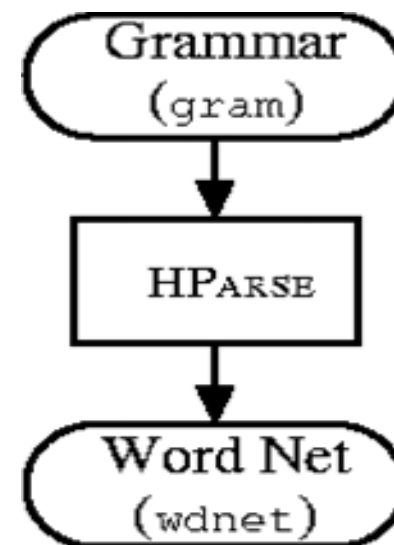
- Grammar for Phone Dialing



Network

- The above high level representation of a task grammar is provided for user convenience
- The HTK recognizer actually requires a word network to be defined using a low level notation called *HTK Standard Lattice Format* (SLF) in which each word instance and each word-to-word transition is listed explicitly

HParse gram wdnnet



Dictionary

- A dictionary with a few entries

A		ah sp
A		ax sp
A		ey sp
CALL		k ao l sp
DIAL		d ay ax l sp
EIGHT		ey t sp
PHONE		f ow n sp
SENT-END	[]	sil
SENT-START	[]	sil
SEVEN		s eh v n sp
TO		t ax so
TO		t uw sp
ZERO		z ia r ow sp

- Function words such as A and TO have multiple pronunciations
The entries
- For SENTSTART and SENTEND have a silence model sil as
their pronunciations and null output symbols

Transcription

- To train a set of HMMs, every file of training data must have an associated phone level transcription
- Master Label File (MLF)

```
#!MLF!#
"/S0001.lab"
ONE
VALIDATED
ACTS
OF
SCHOOL
DISTRICTS
-
"/S0002.lab"
TWO
OTHER
CASES
ALSO
WERE
UNDER
ADVISEMENT
-
"/S0003.lab"
BOTH
FIGURES
(etc.)
```

Coding The Data

- Configuration (Config)

Coding parameters

TARGETKIND = MFCC_0

TARGETRATE = 100000.0 10ms

SAVECOMPRESSED = T

SAVEWITHCRC = T

WINDOWSIZE = 250000.0 25ms

USEHAMMING = T

PREEMCOEF = 0.97

NUMCHANS = 26

CEPLIFTER = 22

NUMCEPS = 12

ENORMALISE = F

in 100 nanosecond unit

10ms

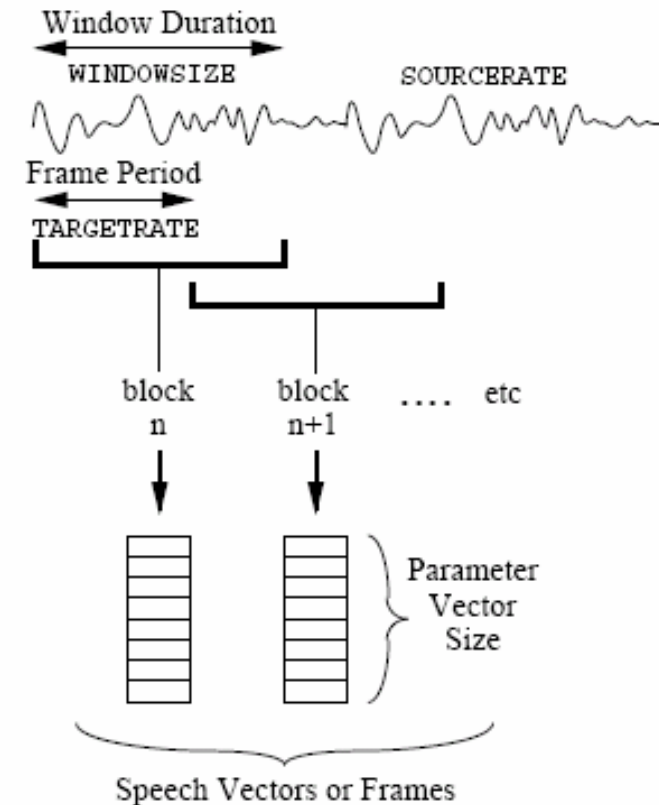
25ms

Pre-emphasis filter coefficient

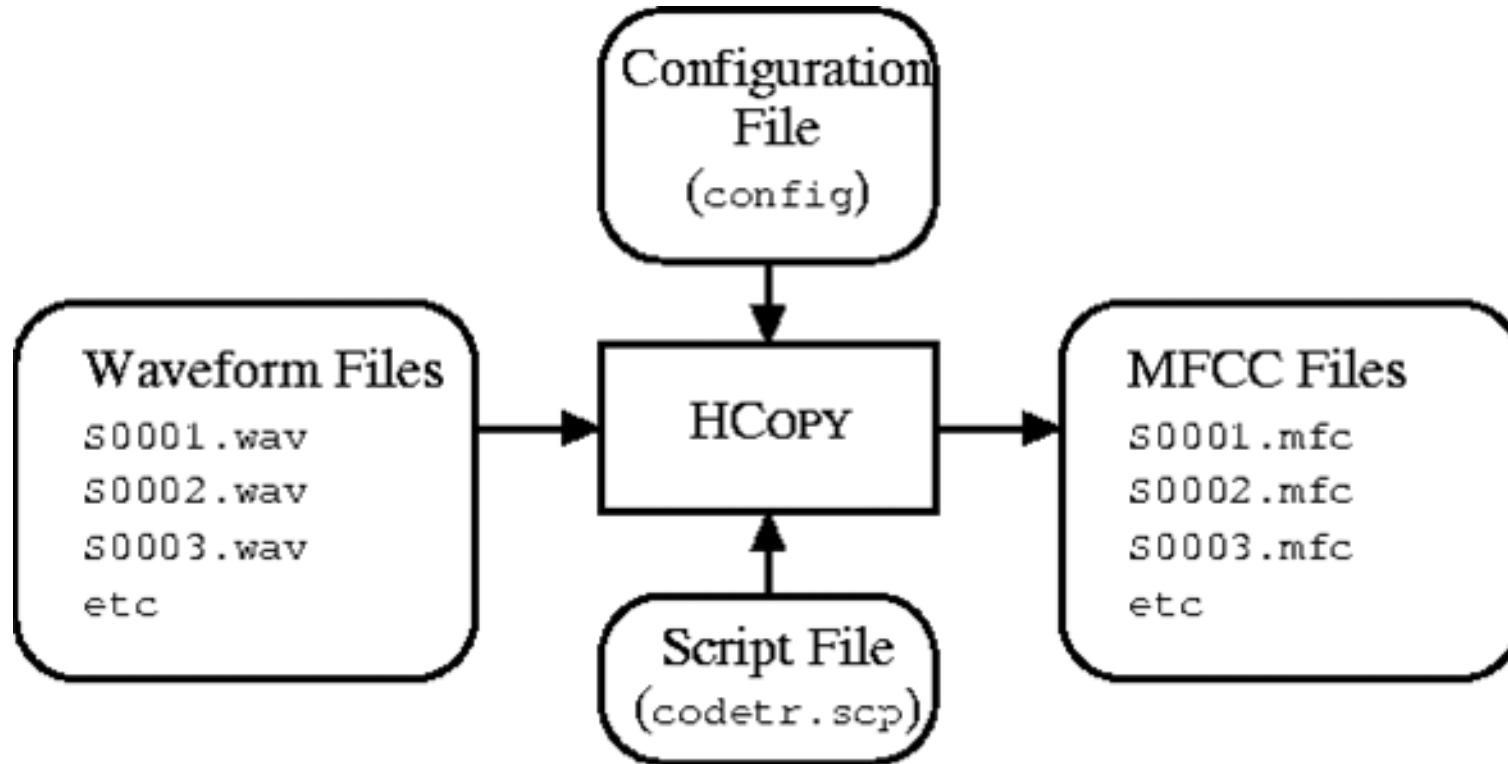
Filter bank numbers

Cepstral Liftering Setting

Number of output cepstral coefficients

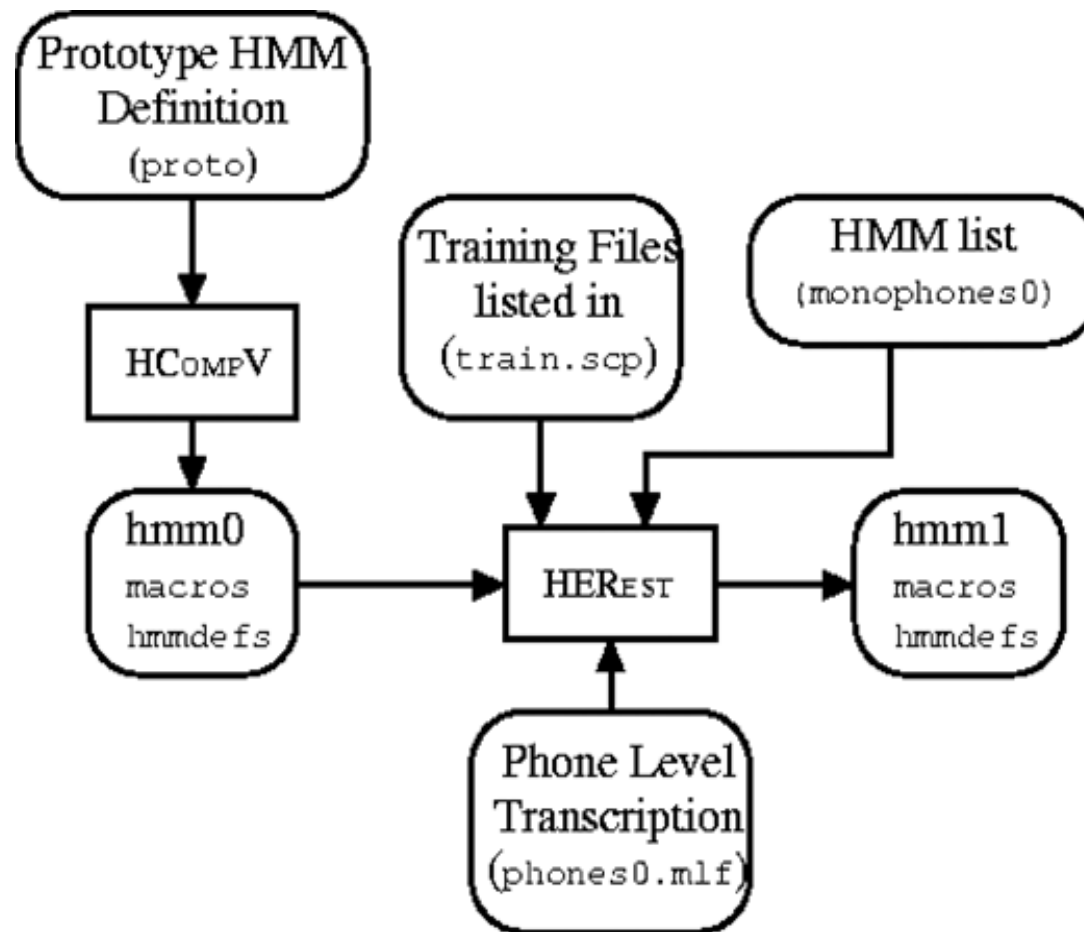


Coding The Data (cont.)

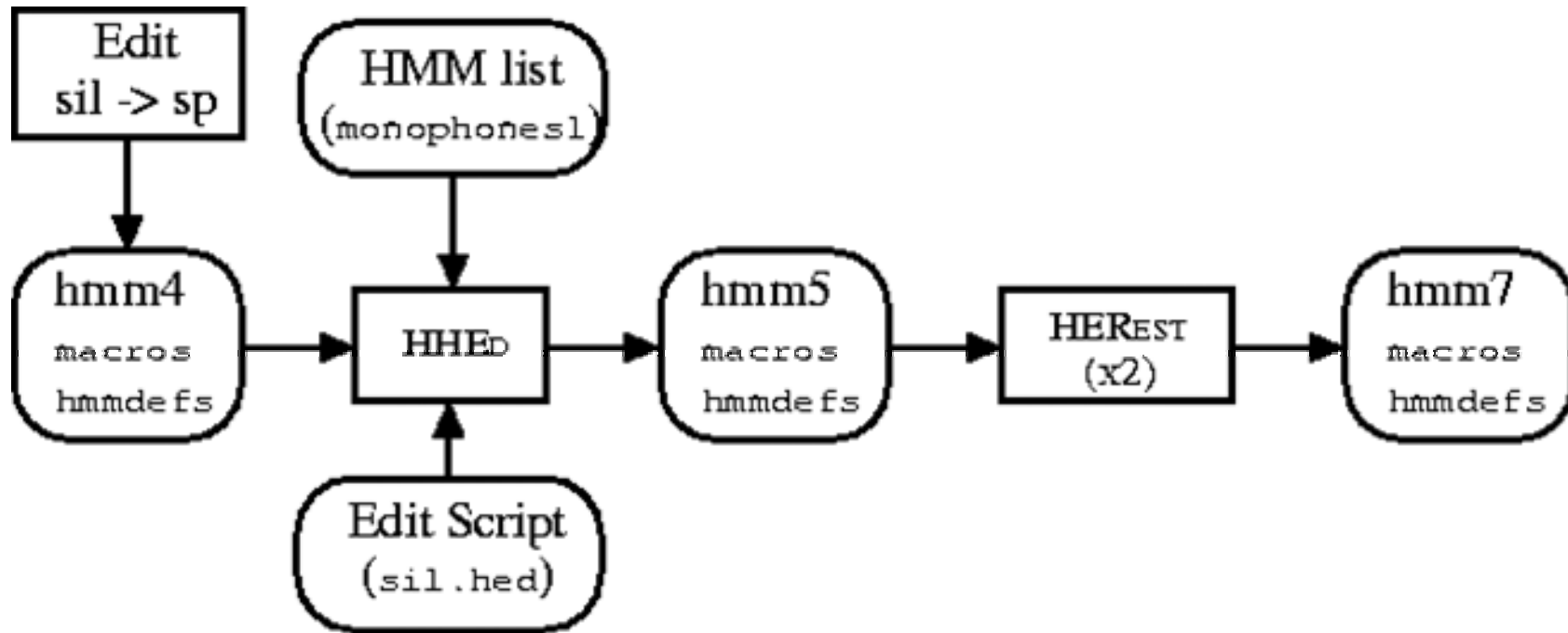


HCopy -T 1 -C config -S codetr.scp

Training



Tee Model



Recognition

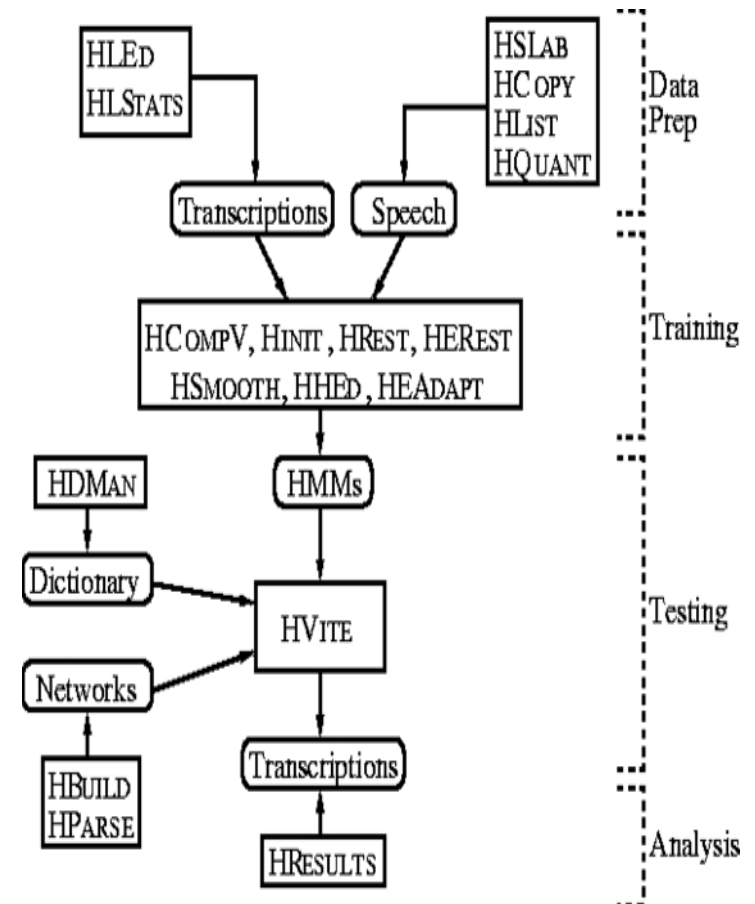
- HVite -T 1 -S test.scp -H hmmset -i results -w wdnet dict hmmlist
- HResults -I refs wlist results

$$\text{Percent Correct} = \frac{N - D - S}{N} \times 100\%$$

$$\text{Percent Accuracy} = \frac{N - D - S - I}{N} \times 100\%$$

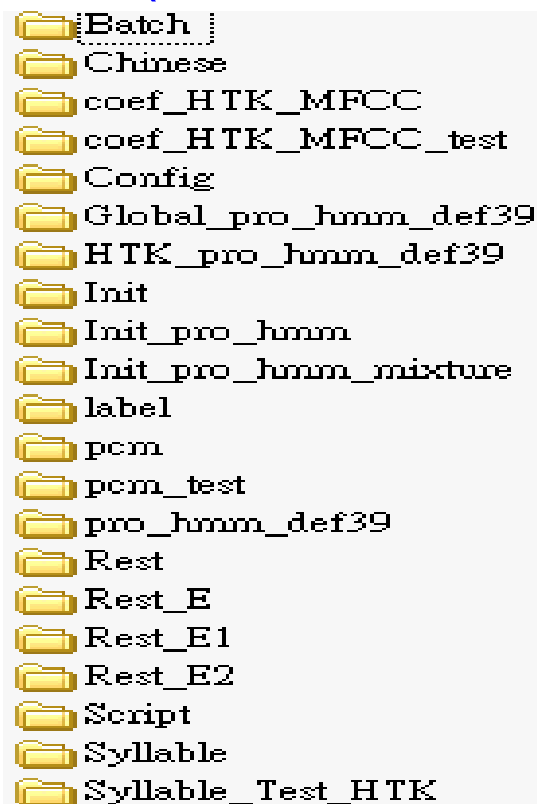
Homework 3: Exercises on HTK

- Practice the use of HTK
- Five Major Steps
 - Environment Setup
 - Data Preparation
 - [HCopy](#)
 - Training
 - [HHed](#), [HCompV](#), [HErest](#)
 - Or [Hinit](#), [HHed](#), [HRest](#), [HERest](#)
 - Testing/Recognition
 - [HVite](#)
 - Analysis
 - [HResults](#)



Experimental Environment Setup

- Download the HTK toolkit and install it
- Copy zipped file of this exercise to a directory name “HTK_Tutorial”, and unzipped the file
- Ensure the following subdirectories have been established (If not, make the subdirectories !)



```
Batch
Chinese
coef_HTK_MFCC
coef_HTK_MFCC_test
Config
Global_pro_hmm_def39
HTK_pro_hmm_def39
Init
Init_pro_hmm
Init_pro_hmm_mixture
label
pcm
pcm_test
pro_hmm_def39
Rest
Rest_E
Rest_E1
Rest_E2
Script
Syllable
Syllable_Test_HTK
```

Step01_HCopy_Train.bat

- Function:
 - Generate MFCC feature files for the training speech utterances
- Command

HCOPY -T 00001 -C ..\config\HCOPY.fig -S ..\script\HCopy_Train.scp

Level of trace information

specify the detailed configuration for feature extraction

specify the pcm and coefficient files and their respective directories

user defined wave format
file header (set to 0 here)
in accordance with sampling rate
Z(zero mean), E(Energy), D(delta)
A(Delta Delta)

Hamming window
Pre-emphasis
filter bank no
liftering setting
Cepstral coefficient no

Intel PC byte Order

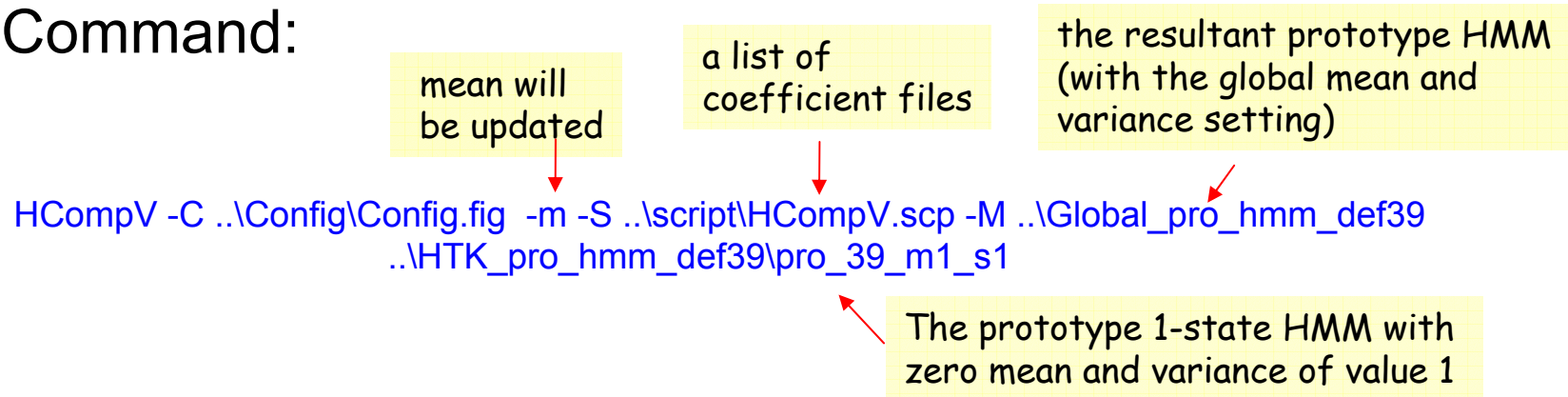
```
#Coding parameters
SOURCEFORMAT=ALEN } 2 bytes per
HEADERSIZE=0       } sample
SOURCERATE=625    1e7/16000
TARGETKIND=MFCC_Z_E_D_A
TARGETRATE=100000.0 10e-3 *1e7
#frameshift 10ms
SAVECOMPRESSED=F
SAVEWITHCRC=F
WINDOWSIZE=320000.0
# framesize = 32ms 32e-3 *1e7
USEHAMMING=T
PREEMCOEF=0.97
NUMCHANS=26
CEPLIFTER=22
NUMCEPS=12
ENORMALIZE=T
NATURALREADORDER=TRUE
NATURALWRITEORDER=TRUE
```

```
../pcm/ag1-n3-001.pcm    ../coef_HTK_MFCC/ag1-n3-001.cof
../pcm/ag1-n3-002.pcm    ../coef_HTK_MFCC/ag1-n3-002.cof
../pcm/ag1-n3-003.pcm    ../coef_HTK_MFCC/ag1-n3-003.cof
../pcm/ag1-n3-004.pcm    ../coef_HTK_MFCC/ag1-n3-004.cof
../pcm/ag1-n3-005.pcm    ../coef_HTK_MFCC/ag1-n3-005.cof
../pcm/ag1-n3-006.pcm    ../coef_HTK_MFCC/ag1-n3-006.cof
../pcm/ag1-n3-007.pcm    ../coef_HTK_MFCC/ag1-n3-007.cof
../pcm/ag1-n3-008.pcm    ../coef_HTK_MFCC/ag1-n3-008.cof
../pcm/ag1-n3-009.pcm    ../coef_HTK_MFCC/ag1-n3-009.cof
../pcm/ag1-n3-010.pcm    ../coef_HTK_MFCC/ag1-n3-010.cof
../pcm/ag1-n3-011.pcm    ../coef_HTK_MFCC/ag1-n3-011.cof
../pcm/ag1-n3-012.pcm    ../coef_HTK_MFCC/ag1-n3-012.cof
../pcm/ag1-n3-013.pcm    ../coef_HTK_MFCC/ag1-n3-013.cof
../pcm/ag1-n3-014.pcm    ../coef_HTK_MFCC/ag1-n3-014.cof
../pcm/ag1-n3-015.pcm    ../coef_HTK_MFCC/ag1-n3-015.cof
../pcm/ag1-n3-016.pcm    ../coef_HTK_MFCC/ag1-n3-016.cof
../pcm/ag1-n3-017.pcm    ../coef_HTK_MFCC/ag1-n3-017.cof
../pcm/ag1-n3-018.pcm    ../coef_HTK_MFCC/ag1-n3-018.cof
../pcm/ag1-n3-019.pcm    ../coef_HTK_MFCC/ag1-n3-019.cof
../pcm/ag1-n3-020.pcm    ../coef_HTK_MFCC/ag1-n3-020.cof
```

Step02_HCompv_S1.bat

- Function:
 - Calculate the global mean and variance of the training data
 - Also set the prototype HMM

- Command:



- Similar for the batch instructions

Step02_HCompv_S2.bat
Step02_HCompv_S3.bat
Step02_HCompv_S4.bat

Generate prototype HMMs with
different state numbers

Step02_HCompv_S1.bat (count.)

- Note! You should manually edit the resultant prototype HMMs in the directory “Global_pro_hmm_def39” to remove the row

~h “prot_39_m1_sX”

- Remove the name tags, because these proto HMMs will be used as the prototypes for all the INITIALs, FINALs, and silence models

```
~0
<STREAMINFO> 1 39
<VECSIZE> 39<NULLD><MFCC_E_D_A_Z>
~h "pro 39 m1 s1"
<BEGINHMM>
<NUMSTATES> 3
<STATE> 2
<MEAN> 39
 1.839634e-008 1.182030e-008 -2.191493e-009 -3.174569e-009 -2.986750e-009 2.4
<VARIANCE> 39
 5.612354e+001 5.309968e+001 6.036213e+001 4.772774e+001 5.011119e+001 4.7409
<GCONST> 9.671951e+001
<TRANSP> 3
 0.000000e+000 1.000000e+000 0.000000e+000
 0.000000e+000 7.000000e-001 3.000000e-001
 0.000000e+000 0.000000e+000 0.000000e+000
<ENDHMM>
```

remove this row
for all proto HMMs



```
~0
<STREAMINFO> 1 39
<VECSIZE> 39<NULLD><MFCC_E_D_A_Z>
<BEGINHMM>
<NUMSTATES> 3
<STATE> 2
<MEAN> 39
 1.839634e-008 1.182030e-008 -2.191493e-009 -3.174569e-009 -2.986750e-009 2
<VARIANCE> 39
 5.612354e+001 5.309968e+001 6.036213e+001 4.772774e+001 5.011119e+001 4.74
<GCONST> 9.671951e+001
<TRANSP> 3
 0.000000e+000 1.000000e+000 0.000000e+000
 0.000000e+000 7.000000e-001 3.000000e-001
 0.000000e+000 0.000000e+000 0.000000e+000
<ENDHMM>
```

Step03_CopyProHMM.bat

- Function
 - Copy the prototype HMMs, which have global mean and variances setting, to the corresponding acoustic models as the prototype HMMs for the subsequent training process
- Content of the bath file

```
copy ..\Global_pro_hmm_def39\pro_39_ml_s2 ..\Init_pro_hmm\sic_e
copy ..\Global_pro_hmm_def39\pro_39_ml_s2 ..\Init_pro_hmm\sic_a
copy ..\Global_pro_hmm_def39\pro_39_ml_s2 ..\Init_pro_hmm\sic_o
copy ..\Global_pro_hmm_def39\pro_39_ml_s2 ..\Init_pro_hmm\sic_i
copy ..\Global_pro_hmm_def39\pro_39_ml_s2 ..\Init_pro_hmm\sic_u
copy ..\Global_pro_hmm_def39\pro_39_ml_s2 ..\Init_pro_hmm\sic_iu
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\u
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ua
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\uai
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\uan
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\uang
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\uei
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\uen
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ueng
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\uo
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ a
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ ai
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ an
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ ang
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ au
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ e
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ ei
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ en
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ eng
copy ..\Global_pro_hmm_def39\pro_39_ml_s4 ..\Init_pro_hmm\ er
```


Step04_HHed_ModelMixSplit.bat

- Function:
 - Split the single Gaussian distribution of each HMM state into n mixture of Gaussian distributions, while the mixture number is set with respect to size of the training data for each model

- Command:

```

HHed -C ..\Config\ConfigHHed.fig -d ..\nit_pro_hmm -M ..\nit_pro_hmm_mixture
..\Script\HedCmd.scp ..\Script\rcdmodel_sil
    
```

dir of the proto HMMs

dir of the resultant HMMs

HHed configuration

HMM model list

List of the models to be trained

mixture splitting command
the resultant mixture number

The states of a specific model to be processed

HHed configuration

```

MU 8 { sic_a.state[2-3].mix }
MU 4 { sic_o.state[2-3].mix }
MU 16 { sic_e.state[2-3].mix }
MU 32 { sic_i.state[2-3].mix }
MU 16 { sic_u.state[2-3].mix }
MU 16 { sic_iu.state[2-3].mix }
MU 16 { b_u.state[2-4].mix }
MU 16 { b_a.state[2-4].mix }
MU 8 { b_e.state[2-4].mix }
MU 8 { b_ee.state[2-4].mix }
MU 8 { b_i.state[2-4].mix }
MU 8 { b_o.state[2-4].mix }
MU 16 { chi_i.state[2-4].mix }
MU 8 { chi_iu.state[2-4].mix }
MU 8 { ch_a.state[2-4].mix }
MU 8 { ch_e.state[2-4].mix }
MU 4 { ch_empty.state[2-4].mix }
MU 4 { ch_o.state[2-4].mix }
MU 8 { ch_u.state[2-4].mix }
    
```

```

a
ai
an
ang
au
b_a
b_e
b_ee
b_i
b_o
b_u
ch_a
ch_e
ch_empty
ch_o
ch_u
chi_i
chi_iu
d_a
d_e
d_ee
d_i
d_o
d_u
e
    
```

Step05_HERest_Train.bat

- Function:
 - Perform HMM model training
 - Baum-Whelch (EM) training performed over each training utterance using the composite model

- Commands:

```
HERest -T 00001 -t 100 -v 0.000000001 -C ..\Config\Config.fig -L ..\label -X rec -d ..\Init_pro_hmm_mixture  
-s statics -M ..\Rest_E -S ..\script\HERest.scp ..\Script\rcdmodel_sil
```

List of the coefficient files of the training data

List of the models to be trained

```
HERest -T 00001 -t 100 -v 0.000000001 -C ..\Config\Config.fig -L ..\label -X rec -d ..\Rest_E  
-s statics -M ..\Rest_E -S ..\script\HERest.scp ..\Script\rcdmodel_sil
```

.....

Pruning threshold
of the forward-backward procedures

cut-off value of the variance

- You can repeat the above command multiple times, e.g., 30 time, to achieve a better set of HMM models

Step05_HERest_Train.bat (cont.)

A label file of a training utterance

```
0 1100000 sil
1100000 2800000 b_o
2800000 3600000 o
3600000 4800000 l_u
4800000 6500000 uan
6500000 7300000 f_a
7300000 8800000 en
8800000 10200000 j_e
10200000 11400000 eng
11400000 15900000 sil
```

Boundary information of the segments of HMM models (will not be used for HERest)

List of the models to be trained

```
a
ai
an
ang
au
b_a
b_e
b_ee
b_i
b_o
b_u
ch_a
ch_e
ch_empty
ch_o
ch_u
chi_i
chi_iu
d_a
d_e
d_ee
d_i
d_o
d_u
e
```

Step06_HCopyTest.bat

- Function:
 - Generate MFCC feature files for the testing speech utterances
- Command

```
HCOPY -T 00001 -C ..\Config\Config.fig -S ..\script\HCopy_Test.scp
```

The detailed explanation can be referred to:

[Step01_HCopy_Train.bat](#)

Step07_HVite_Recognition.bat

- Function:
 - Perform free-syllable decoding on the testing utterances

- Command

```
HVite -C ..\Config\Config.fig -T 1 -X ..\script\netparsed -o SW  
-w ..\script\SYL_WORD_NET.netparsed -d ..\Rest_E -l ..\Syllable_Test_HTK  
-S ..\script\HVite_Test.scp ..\script\SYLLABLE_DIC ..\script\rcdmodel_sil
```

The extension file name for the search/recognition network

Set the output label files format: no score information, and no word information

A list of the testing utterances

The search/recognition network generated by HParse command

A list to lookup the constituent INITIAL/FINAL models for the composite syllable models

Dir to load the HMM models

Dir to save the output label files

Step07_HVite_Recognition.bat (cont.)

The search/recognition network
before performing HParse command

```
(  
j_empty-empt1 |  
ch_empty-empt1 |  
sh_empty-empt1 |  
r_empty-empt1 |  
. .  
. .  
ji_iu-iue |  
k_u-uo |  
sil > |  
)
```

a composite syllable model

Regular expression

or

loop

A list to lookup the constituent
INITIAL/FINAL models for the
composite syllable models

```
j_empty-empt1 j_empty empt1  
ch_empty-empt1 ch_empty empt1  
. .  
. .  
sic_a-a sic_a a  
j_a-a j_a a  
ch_a-a ch_a a  
sh_a-a sh_a a  
tz_a-a tz_a a  
. .  
sil sil
```

HParse SYL_WORD_NET SYL_WORD_NET.netparsed

```
VERSION=1.0  
N=407 L=12 13  
I=0 W=!NULL  
I=1 W=!NULL  
I=2 W=j_empty-empt1  
I=3 W=!NULL  
I=4 W=ch_empty-empt1  
I=5 W=sh_empty-empt1  
I=6 W=r_empty-empt1  
I=7 W=tz_empty-empt2  
I=8 W=ts_empty-empt2  
I=9 W=s_empty-empt2  
I=10 W=sic_a-a  
I=11 W=j_a-a
```

The search/recognition network
generated by HParse command

Step08_HResults_Test.bat

- Function:
 - Analyze the recognition performance

- Command The extension file name for the label files ignore the the silence label "sil"

```
HResults -C ..\Config\Config.fig -T 00020 -X rec -e ??? sil -L ..\Syllable  
-S ..\script\Hresults_rec600.scp ..\script\SYLLABLE_DIC
```

A list of the label files generated by the recognition process

```
Syllable_Test_HTK\bruce-b07-001.rec  
Syllable_Test_HTK\bruce-b07-002.rec  
Syllable_Test_HTK\bruce-b07-003.rec  
Syllable_Test_HTK\bruce-b07-004.rec  
Syllable_Test_HTK\bruce-b07-005.rec  
Syllable_Test_HTK\bruce-b07-006.rec  
Syllable_Test_HTK\bruce-b07-007.rec  
Syllable_Test_HTK\bruce-b07-008.rec  
Syllable_Test_HTK\bruce-b07-009.rec  
Syllable_Test_HTK\bruce-b07-010.rec  
Syllable_Test_HTK\bruce-b07-011.rec  
Syllable_Test_HTK\bruce-b07-012.rec  
Syllable_Test_HTK\bruce-b07-013.rec  
Syllable_Test_HTK\bruce-b07-014.rec  
Syllable_Test_HTK\bruce-b07-015.rec  
Syllable_Test_HTK\bruce-b07-016.rec  
Syllable_Test_HTK\bruce-b07-017.rec  
Syllable_Test_HTK\bruce-b07-018.rec  
Syllable_Test_HTK\bruce-b07-019.rec  
Syllable_Test_HTK\bruce-b07-020.rec
```

Dir lookup the reference label files

```
=====  
Date: Sat Nov 22 10:02:13 2003  
Ref : ../Syllable  
Rec : ..\Syllable_Test_HTK\bruce-b07-001.rec  
      : ..\Syllable_Test_HTK\bruce-b07-002.rec  
      : ..\Syllable_Test_HTK\bruce-b07-003.rec  
      : ..\Syllable_Test_HTK\bruce-b07-004.rec  
      : ..\Syllable_Test_HTK\bruce-b07-005.rec  
-----  
Overall Results  
SENT: %Correct=11.33 [H=68, S=532, N=6001]  
WORD: %Corr=69.49, Acc=64.16 [H=2685, D=47, S=1132, I=206, N=3864]  
=====
```

Step09_BatchMFCC_Def39.bat

- Also, you can train the HMM models in another way

Hinit  (HHEd)  HRest  HERest

- For detailed information, please referred to the previous slides or the HTK manual
- You can compare the recognition performance by running
 Step02~Step05
or Step09 alone