Introduction to HTK Toolkit

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Reference:

- Steve Young et al. *The HTK Book*. Version 3.4, March 2006.

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

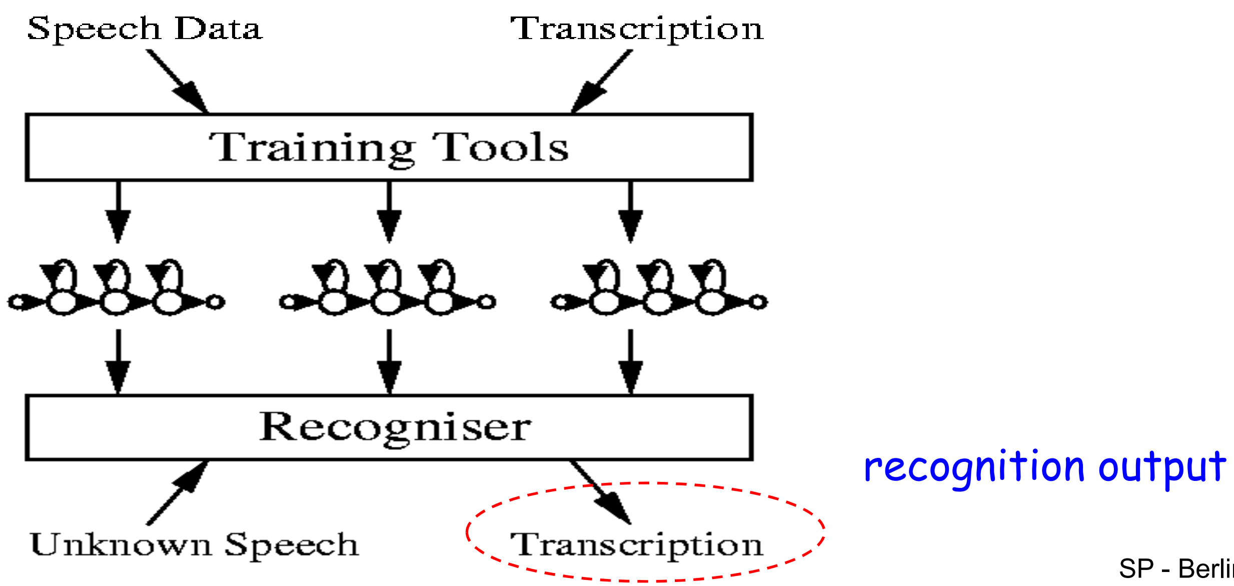
- HTK: A toolkit for building Hidden Markov Models

- 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

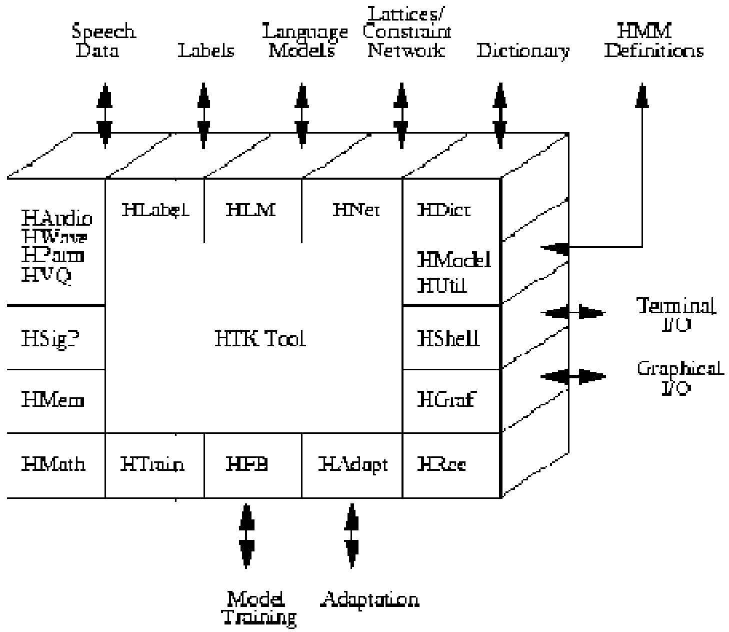


An Overview of HTK (cont.)

- 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

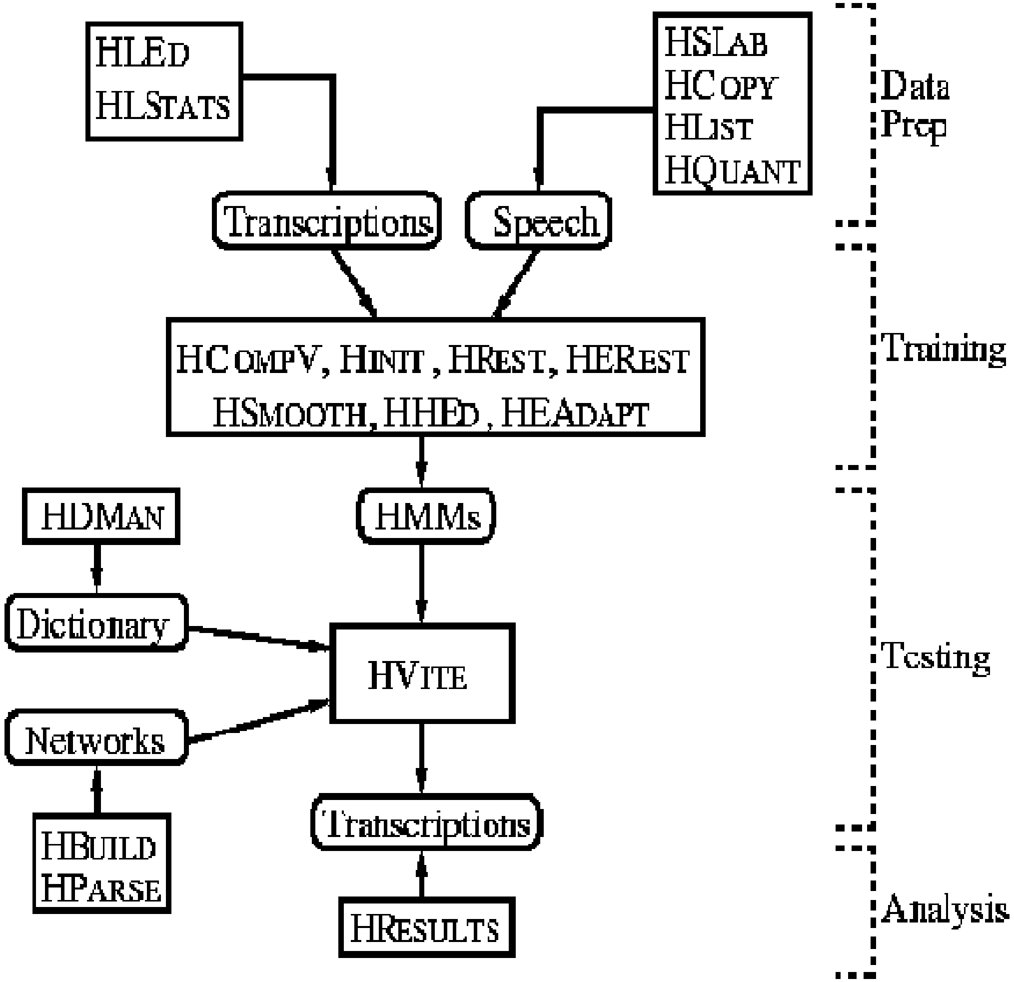
HFoo -T -C Config1 -f 34.3 -a -s myfile file1 file2



    - 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.)

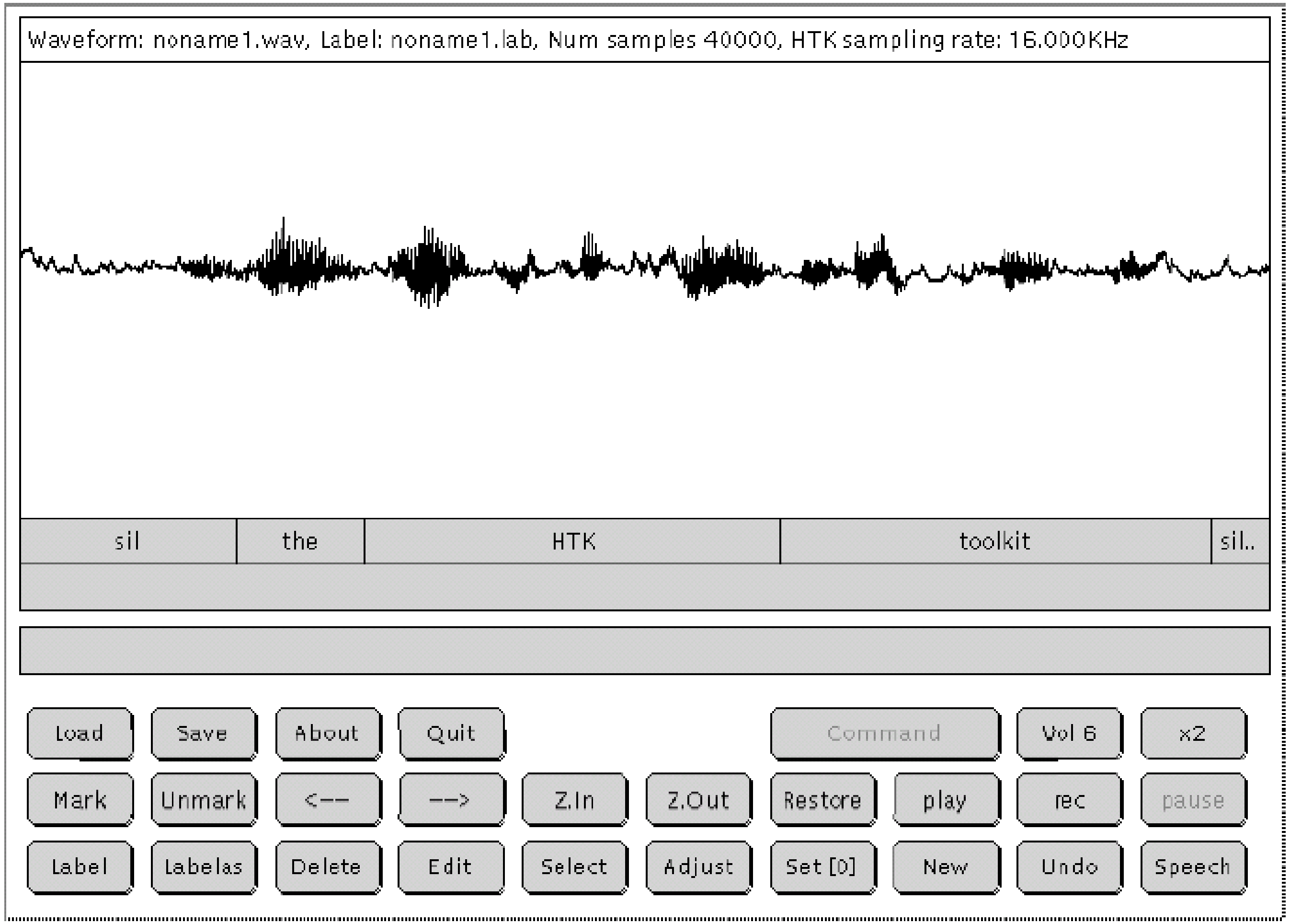
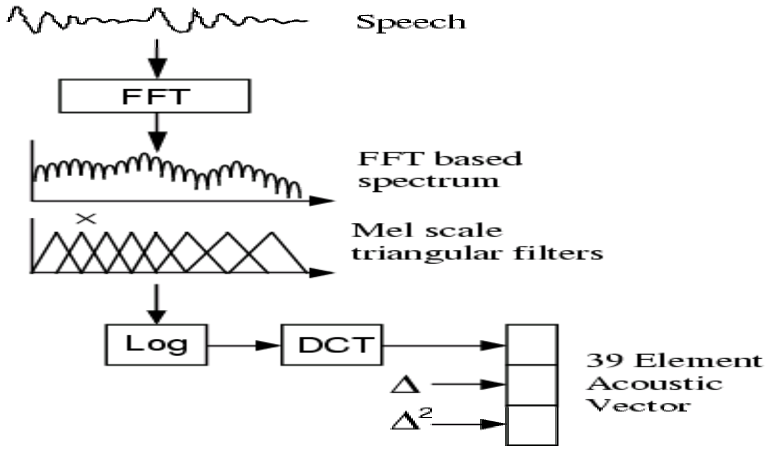


Fig. 14.1 HSLab display window

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

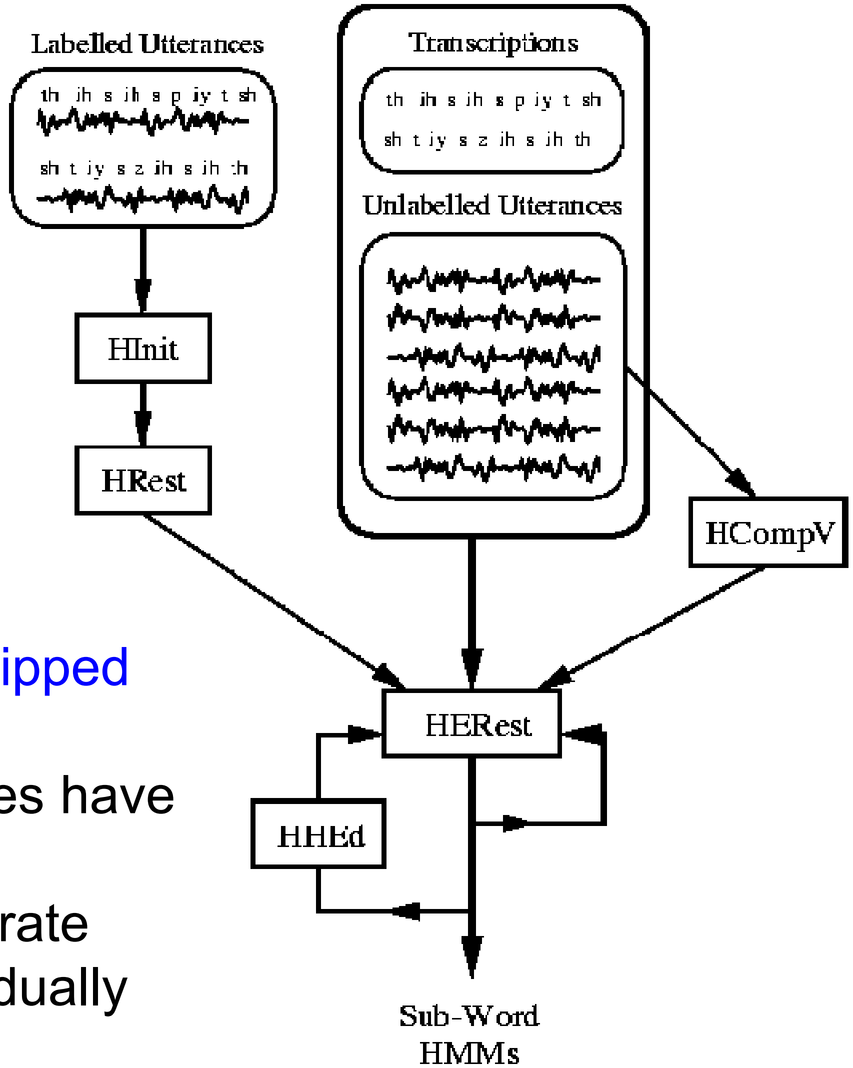
- Prototype HMMs
  - Define the topology required for each HMM by writing a prototype Definition
  - HTK allows HMMs to be built with any desired topology
  - HMM definitions stored as simple text files
  - All of the HMM parameters (the *means* and *variances* of Gaussian distributions) given in the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

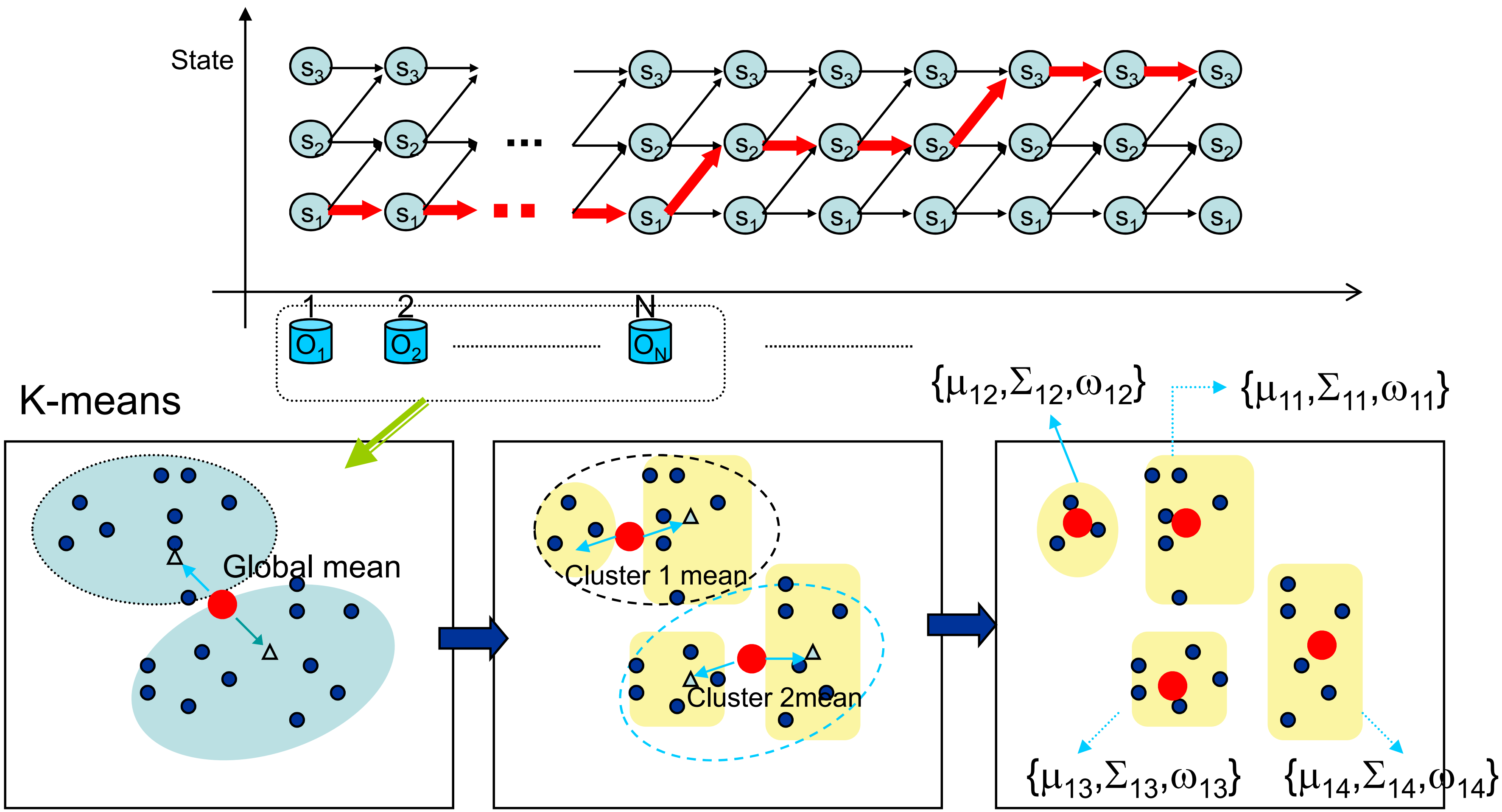


Training Phase (cont.)

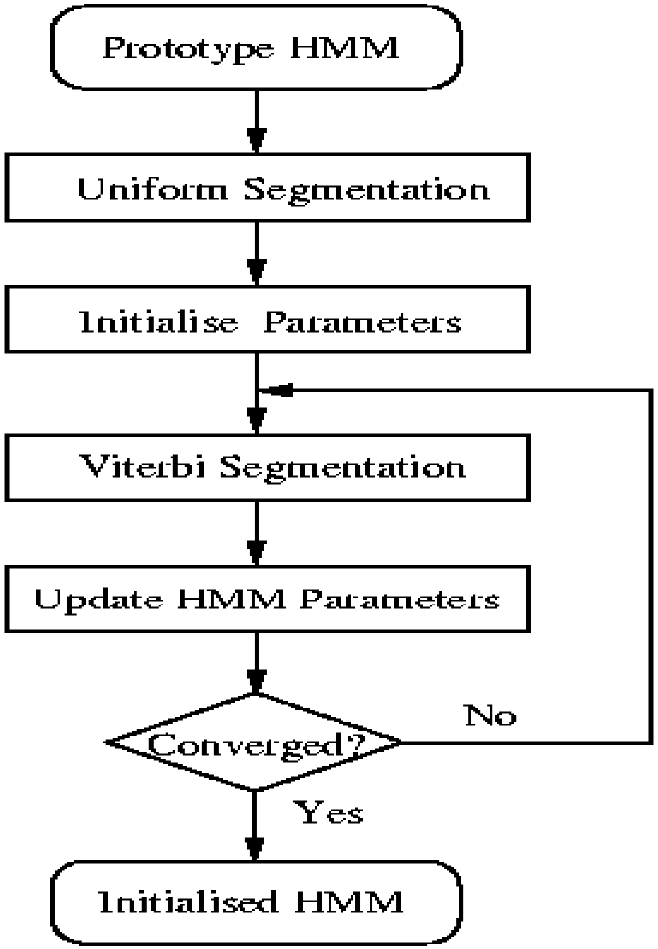
- *HINIT*
  - Iteratively computes an initial set of parameter value using the *segmental k-means training* procedure
    - It reads in all of the bootstrap training data and cuts out all of the examples of a specific phone
    - On the first iteration cycle, the training data are uniformly segmented with respective 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*

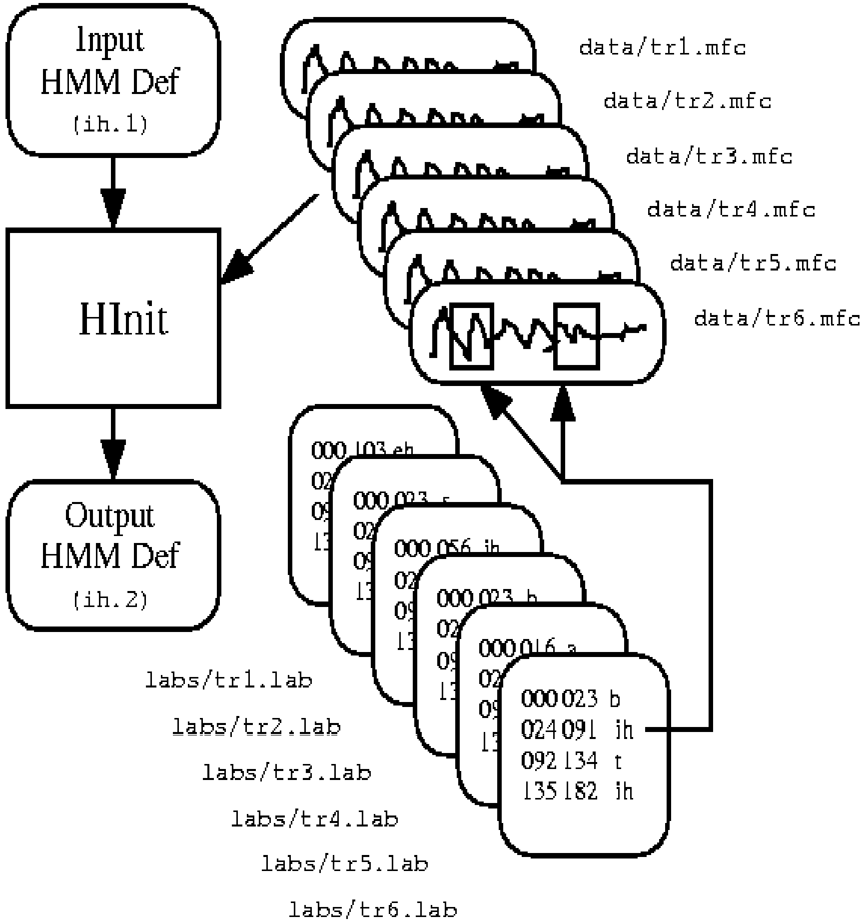
Training Phase (cont.)



Training Phase (cont.)

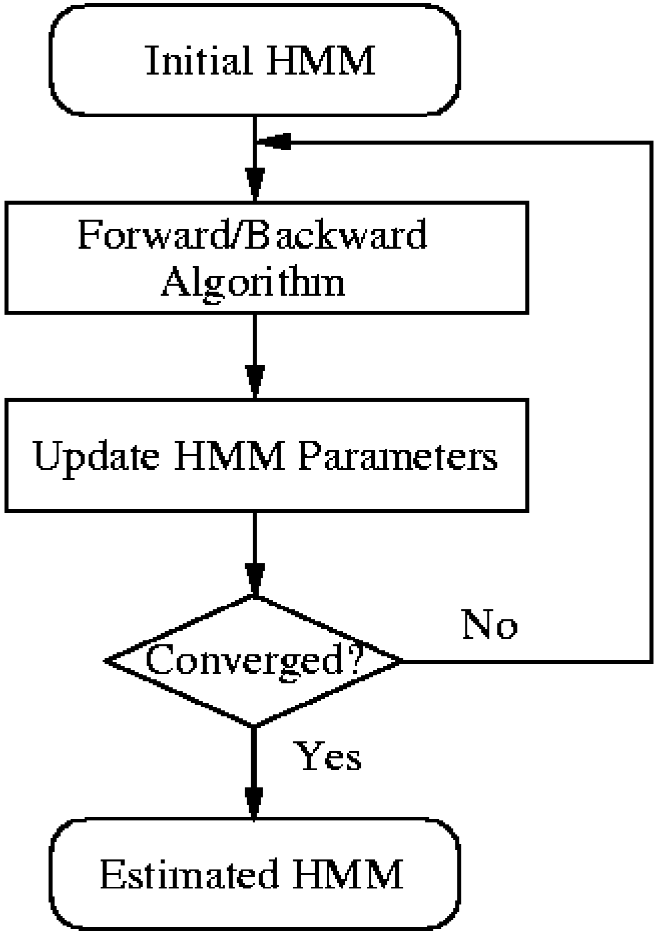


HInit Operation



File Processing in HInit

Training Phase (cont.)



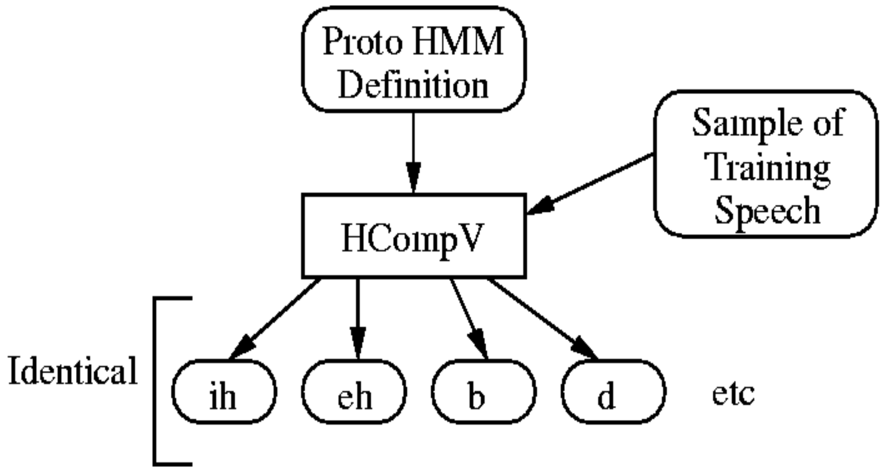
HRest Operation

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.)

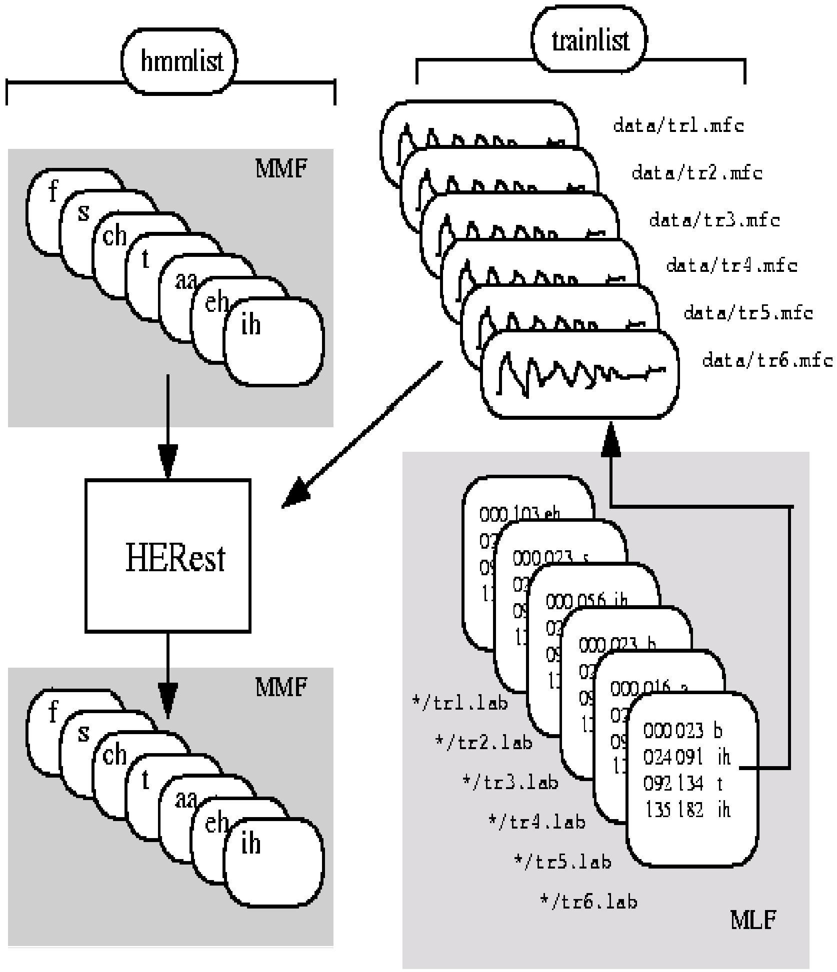


Flat Start Initialisation

- 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.)

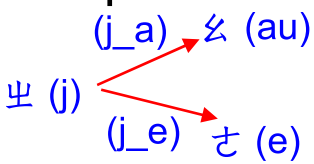
- *HEREST*
  - Performs a single *Baum-Welch* re-estimation of the whole set of the HMMs simultaneously
    - For each training utterance, the corresponding phone models are concatenated and the forward-backward algorithm is used to accumulate the statistics of state occupation, means, variances, etc. 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.)

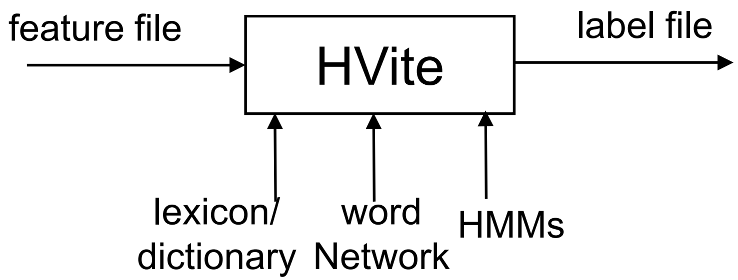
- Model Refinement
  - The philosophy of system construction in HTK is that HMMs should be refined incrementally
  - **CI to CD:** A typical progression is to start with a simple set of single Gaussian context-independent phone models and then iteratively refine them by expanding them to include context-dependency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



right-context-dependent modeling

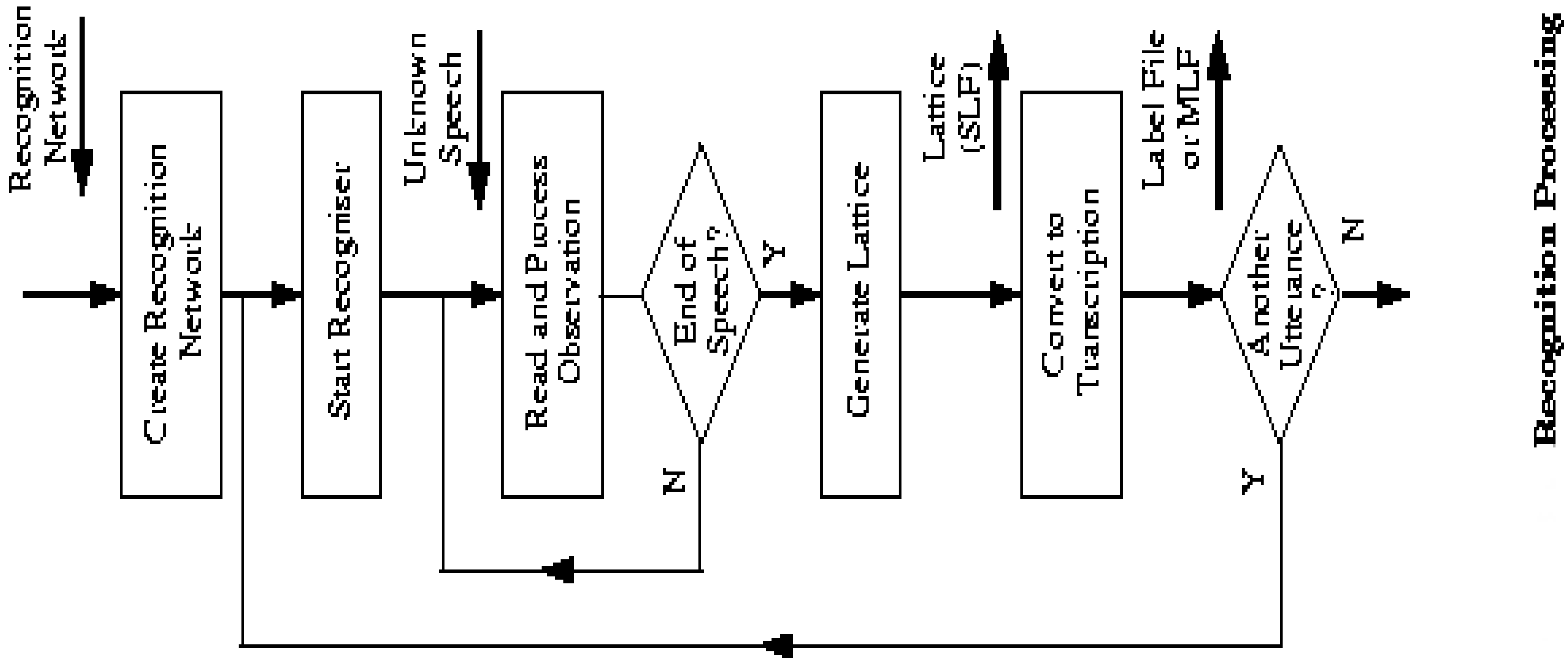
  - **Tying:** The tool *HHED* is a HMM definition editor which will clone models into context-dependent sets, apply a variety of parameter tyings and increase the number of mixture components in specified distributions
  - **Adaptation:** To improve performance for specific speakers the tools *HEADAPT* 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



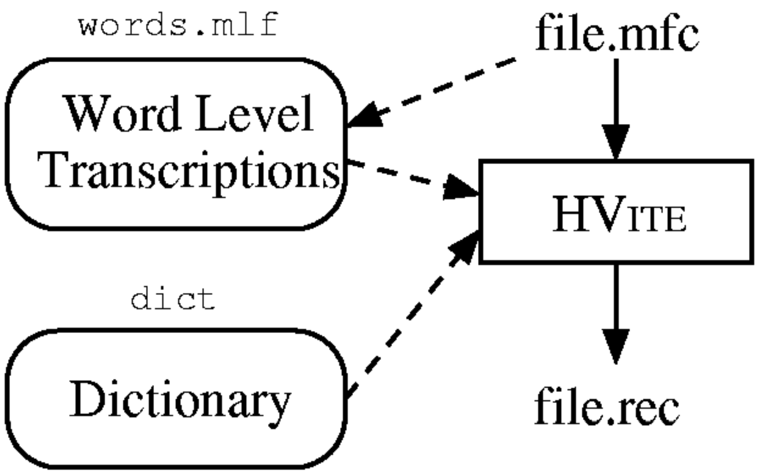
- *HVITE*
  - Performs Viterbi-based speech recognition
  - Takes a network describing the allowable word sequences, a dictionary defining how each word is pronounced and a set of HMMs as inputs
  - Supports cross-word triphones, also can run with multiple tokens to generate lattices containing multiple 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 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*)

reference				test		
t_{s1}	t_{e1}	a		t_{s1}	t_{e1}	a
t_{s2}	t_{e2}	b		t_{s2}	t_{e2}	c
t_{s3}	t_{e3}	b		t_{s3}	t_{e3}	b
.				.		
.				.		

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

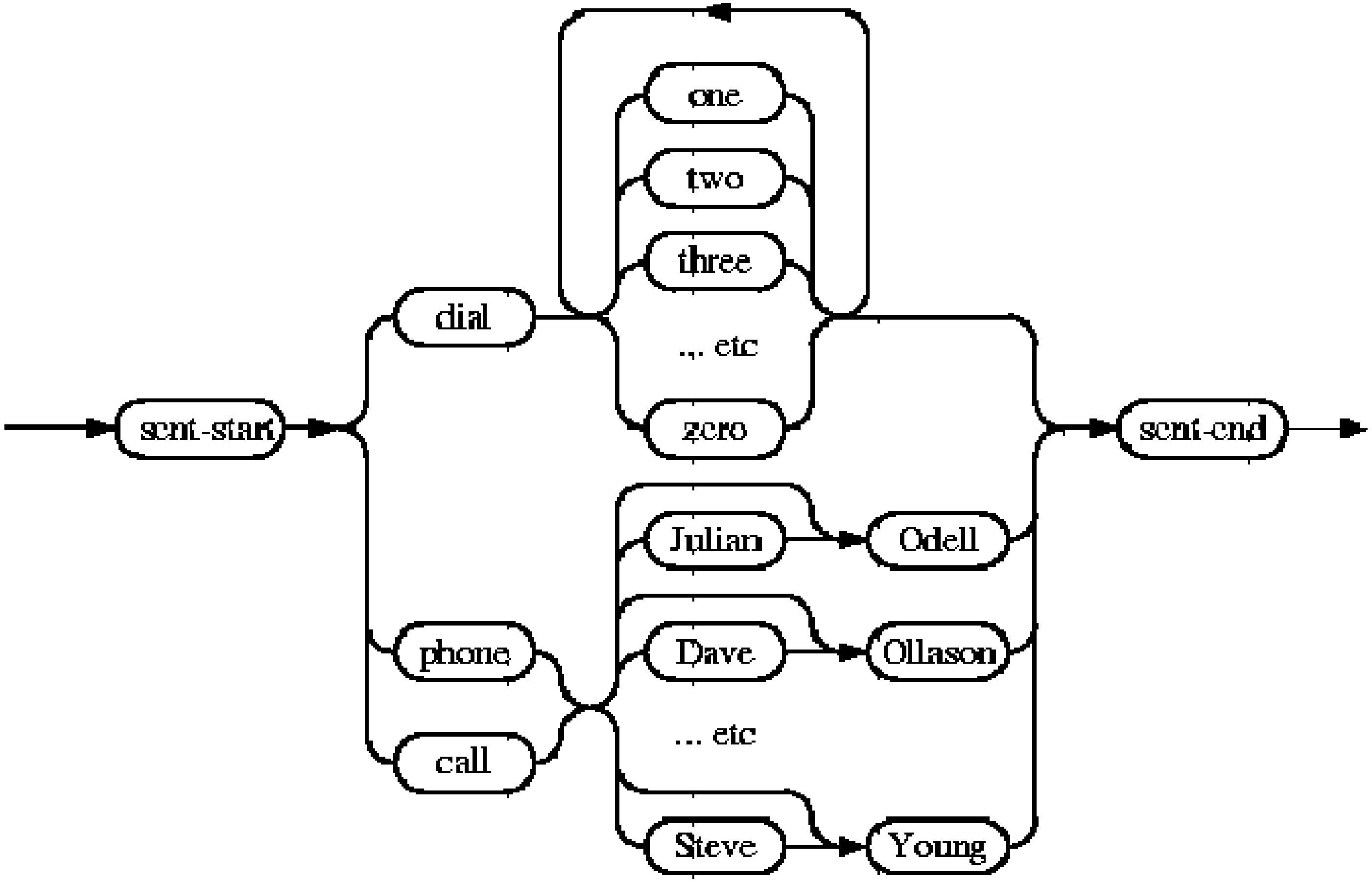
  – $digit = ONE | TWO | THREE | FOUR | FIVE |
    SIX | SEVEN | EIGHT | NINE | OH | ZERO;

    $name = [JOOP] JANSEN | [JULIAN] ODELL | [DAVE]
    OLLASON | [PHIL] WOODLAND | [STEVE] YOUNG;

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

Grammar for Voice Dialing

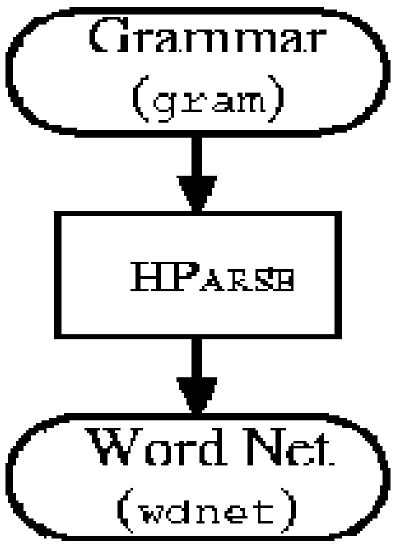
- Grammar for Phone Dialing



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 wdnet



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 sp
TO                    t uw sp
ZERO                  z ia r ow sp
```

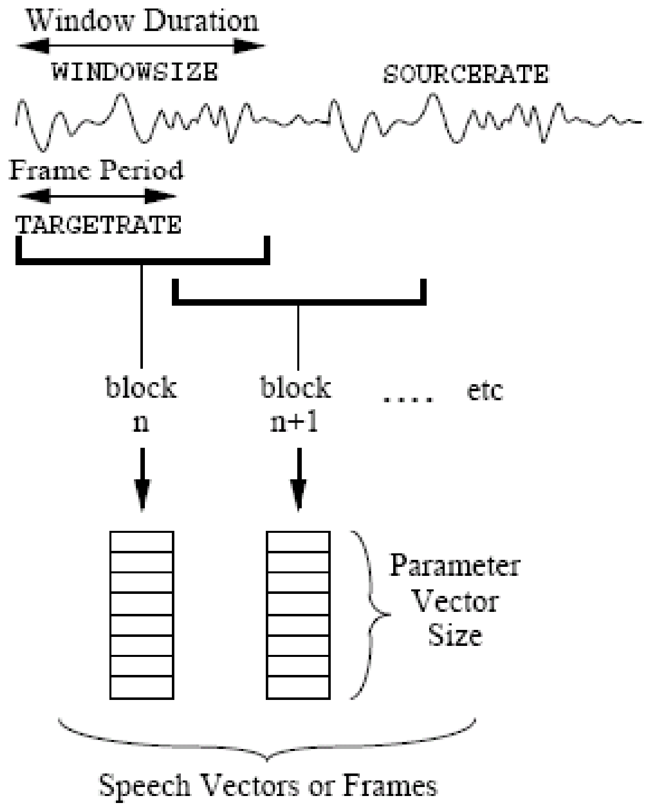
  – Function words such as A and TO have multiple pronunciations The entries
  – For SENTSTART and SENTEND have a silence model sil as their pronunciations and null output symbols

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)

in 100 nanosecond unit

```
# Coding parameters
TARGETKIND = MFCC_0
TARGETRATE = 100000.0
SAVECOMPRESSED = T
SAVEWITHCRC = T
WINDOWSIZE = 250000.0
USEHAMMING = T
PREEMCOEF = 0.97
NUMCHANS = 26
CEPLIFTER = 22
NUMCEPS = 12
ENORMALISE = F
```

TARGETRATE: 10ms

WINDOWSIZE: 25ms

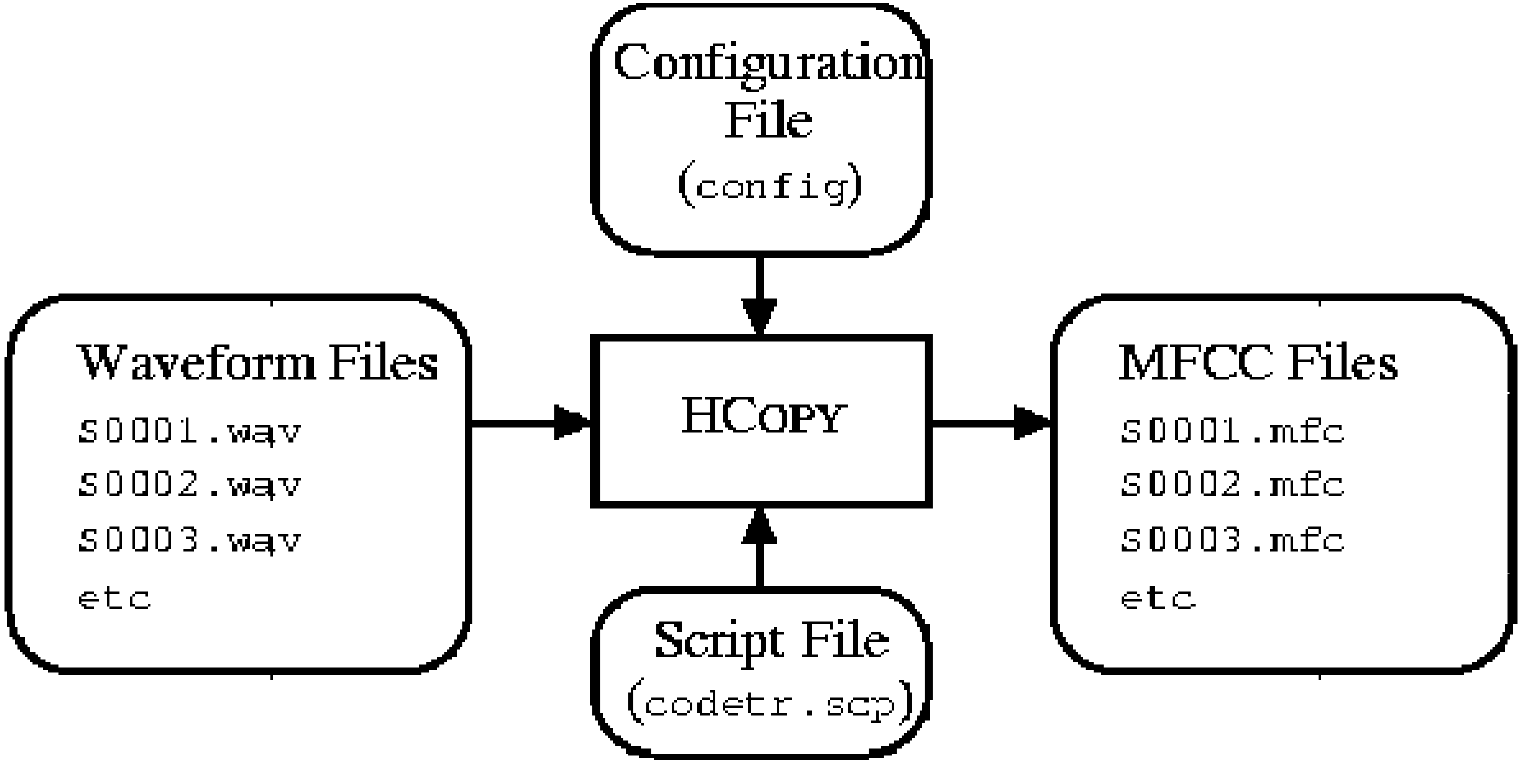
PREEMCOEF: Pre-emphasis filter coefficient

NUMCHANS: Filter bank numbers

CEPLIFTER: Cepstral Liftering Setting

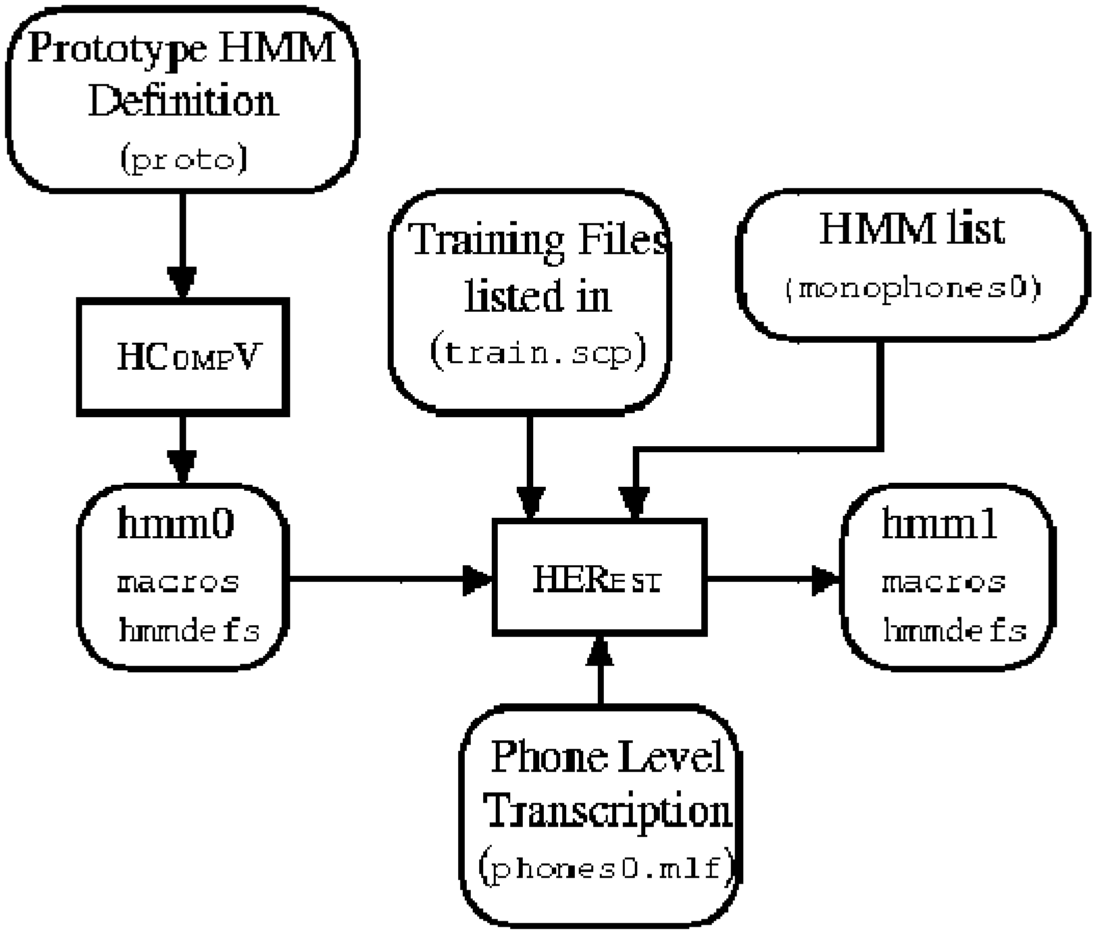
NUMCEPS: Number of output cepstral coefficients

Coding The Data (cont.)



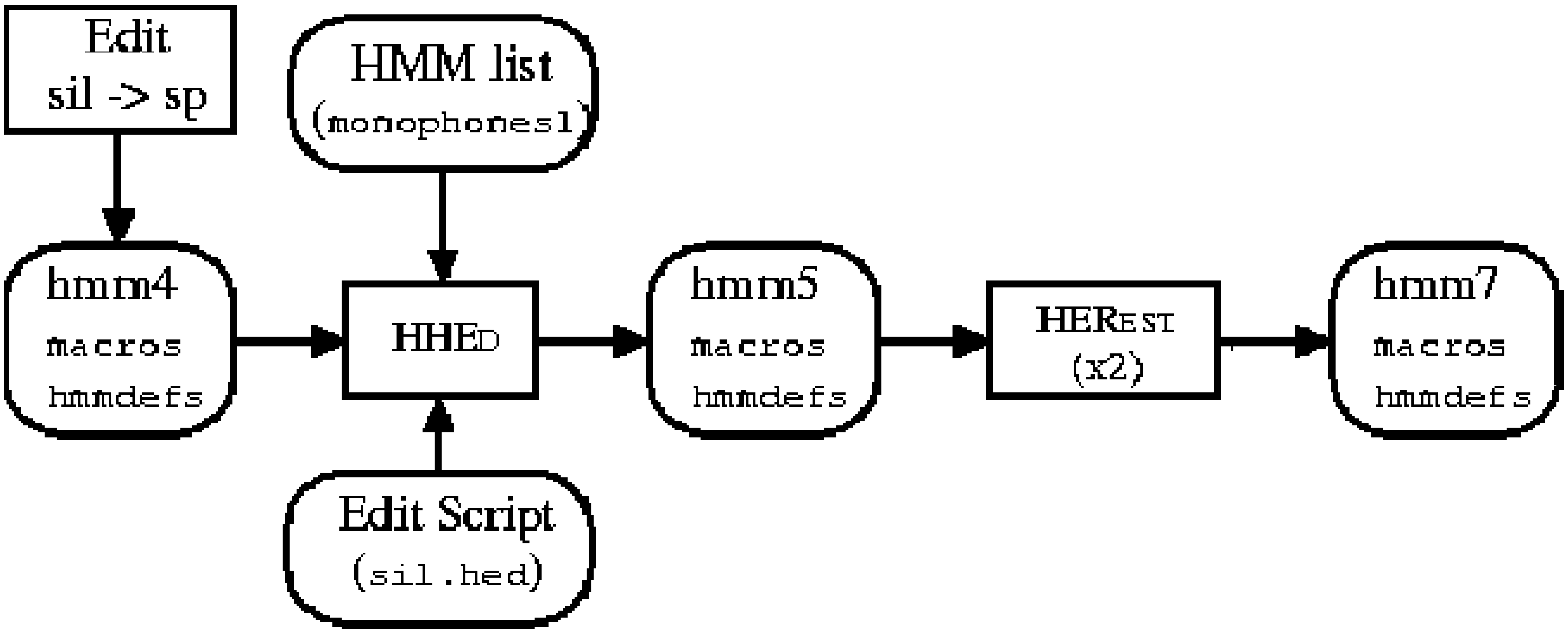
HCopy -T 1 -C config -S codetr.scp

Training



Prototype HMM Definition
(proto)
HCompV
hmm0
macros
hmmdefs
Training Files listed in
(train.scp)
HMM list
(monophones0)
HERest
hmm1
macros
hmmdefs
Phone Level Transcription
(phones0.mlf)

Tee Model


Edit
sil -> sp
HMM list
(monophones1)
hmm4
macros
hmmdefs
HHED
hmm5
macros
hmmdefs
HEREST
(x2)
hmm7
macros
hmmdefs
Edit Script
(sil.hed)

Recognition

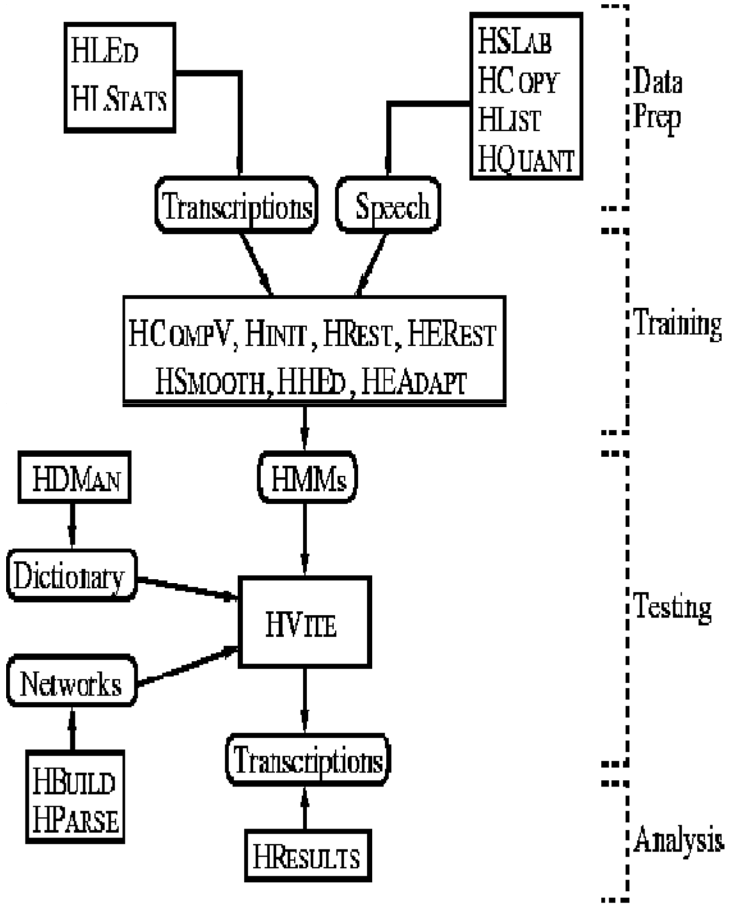
- HVite -T 1 -S test.scp -H hmmset -i results -w wdnet dict hmmlist
- HResults -I refs wlist results

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

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

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

```
#Coding parameters
SOURCEFORMAT=ALEN
HEADERSIZE=0
SOURCERATE=625
TARGETKIND=MFCC_Z_E_D_A
TARGETRATE=100000.0
#frameshift 10ms
SAVECOMPRESSED=F
SAVEWITHCRC=F
WINDOWSIZE=320000.0
# framesize = 32ms
USEHAMMING=T
PREEMCOEF=0.97
NUMCHANS=26
CEPLIFTER=22
NUMCEPS=12
ENORMALIZE=T
NATURALREADORDER=TRUE
NATURALWRITEORDER=TRUE
```

- SOURCEFORMAT=ALEN: user defined wave format
- HEADERSIZE=0: file header (set to 0 here)
- SOURCEFORMAT / HEADERSIZE: 2 bytes per sample
- SOURCERATE=625: in accordance with sampling rate; 1e7/16000
- TARGETKIND=MFCC_Z_E_D_A: Z(zero mean), E(Energy), D(delta) A(Delta Delta)
- TARGETRATE=100000.0: 10e-3 *1e7
- # framesize = 32ms: 32e-3 *1e7
- USEHAMMING=T: Hamming window
- PREEMCOEF=0.97: Pre-emphasis
- NUMCHANS=26: filter bank no
- CEPLIFTER=22: liftering setting
- NUMCEPS=12: Cepstral coefficient no
- NATURALREADORDER / NATURALWRITEORDER: Intel PC byte Order

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

Step02_HCompv_S1.bat

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

mean will be updated

a list of coefficient files

the resultant prototype HMM (with the global mean and variance setting)

```
HCompV -C ..\Config\Config.fig  -m -S ..\script\HCompV.scp -M ..\Global_pro_hmm_def39
..\HTK_pro_hmm_def39\pro_39_m1_s1
```

The prototype 1-state HMM with zero mean and variance of value 1

- 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

```
~o
<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

```
~o
<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_m1_s2                 ..\Init_pro_hmm\sic_e
copy  ..\Global_pro_hmm_def39\pro_39_m1_s2                 ..\Init_pro_hmm\sic_a
copy  ..\Global_pro_hmm_def39\pro_39_m1_s2                 ..\Init_pro_hmm\sic_o
copy  ..\Global_pro_hmm_def39\pro_39_m1_s2                 ..\Init_pro_hmm\sic_i
copy  ..\Global_pro_hmm_def39\pro_39_m1_s2                 ..\Init_pro_hmm\sic_u
copy  ..\Global_pro_hmm_def39\pro_39_m1_s2                 ..\Init_pro_hmm\sic_iu
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\u
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\ua
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\uai
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\uan
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\uang
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\uei
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\uen
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\ueng
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\uo
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\a
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\ai
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\an
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\ang
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\au
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\e
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\ei
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\en
copy  ..\Global_pro_hmm_def39\pro_39_m1_s4                 ..\Init_pro_hmm\eng
copy  ..\Global_pro_hmm_def39\pro_39_m1_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:

dir of the proto HMMs

dir of the resultant HMMs

HHEd -C ..\Config\ConfigHHEd.fig -d ..\Init_pro_hmm -M ..\Init_pro_hmm_mixture ..\Script\HEdCmd.scp ..\Script\rcdmodel_sil

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_empt.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_empt
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:

Dir to look the corresponding label files

Dir of initial models

```
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_empt
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_empt-empt1  |
ch_empt-empt1  |
sh_empt-empt1  |
r_empt-empt1  |
.
.
.
ji_iu-iue  |
k_u-uo  |
sil  >
)
```

a composite syllable model

Regular expression

or

loop

HParse SYL_WORD_NET SYL_WORD_NET.netparsed

```
VERSION=1.0
N=407    L=1213
I=0      W=!NULL
I=1      W=!NULL
I=2      W=j_empt-empt1
I=3      W=!NULL
I=4      W=ch_empt-empt1
I=5      W=sh_empt-empt1
I=6      W=r_empt-empt1
I=7      W=tz_empt-empt2
I=8      W=ts_empt-empt2
I=9      W=s_empt-empt2
I=10     W=sic_a-a
I=11     W=j_a-a
```

The search/recognition network generated by HParse command

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

```
j_empt-empt1    j_empt   empt1
ch_empt-empt1    ch_empt   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
```

Step08_HResults_Test.bat

- Function:
  - Analyze the recognition performance
- Command

The extension file name for the label files

ignore 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

Dir lookup the reference label files

```
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
```

```
==================== HTK Results Analysis =====================
  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=600]
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