This toolbox contains primarily Matlab source codes implementing the robust speaker identification (SID) system proposed in:

Xiaojia Zhao, Yang Shao and DeLiang Wang, "CASA-based Robust Speaker Identification," *IEEE Trans. on Audio, Speech and Language Processing,* vol.20, no.5, pp.1608-1616, 2012.

This README was written by Xiaojia Zhao in Oct'11, adapted in May'13.

There are 6 parts of this toolbox and the detailed description of each part is given below.

1. Experimental Setup (Corpus Generation)

Folder: MixtureGeneration

Description: This folder contains files that create mixtures at different signal-to-noise ratios (SNR) as well as different noisy conditions. The Ideal binary mask (IBM) is also derived.

Structure:

MixtureGeneration

cochleagram.m	Program to create cochleagram
CTL 	Folder to store test file lists
erb2hz.m	Convert equivalent rectangular bandwidth (ERB) scale to linear hertz (Hz) scale
f_af_bf_cf.mat	Contains parameter values of equal-loudness functions from BS3383, "Normal equal-loudness level contours for pure tones under free-field listening conditions", table 1
gammatone.m	Use a bank of gammatone filters to decompose input speech signal with equal loudness compression
gammatoneNorm.m	The same as above excluding equal loudness compression and control the overall gain of the filterbank to be 1
hz2erb.m	Convert linear scale to ERB scale
loadData.m	Load data in HTK format
loudness.m mixSRE.m	Compute loudness level in Phons on the basis of equal-loudness functions Mix clean test files with noises at different SNR levels
mixTraining.m	Mix clean speech data with noises at different SNR levels for mask estimation purposes
synthesis.m	Resynthesize target speech out of mixture with CASA mask
synthesisNorm.m	Resynthesize target speech out of mixture with CASA mask without equal loudness compression and thus not distort energy level
writeHTK.m	Write data in HTK format

2. Feature Extraction

Folder: FeatureExtraction

Description: This folder consists of programs to extract speaker features; particular gammatone features (GF) and gammatone frequency cepstral coefficients (GFCC) as well as their delta features.

Structure:

FeatureExtraction

calDelta.m	Program to calculate delta features of input
cochleagram.m	Program to create cochleagram
erb2hz.m	Convert ERB scale to linear scale
f_af_bf_cf.mat	Contains parameter values of equal-loudness functions from BS3383, "Normal
	equal-loudness level contours for pure tones under free-field listening
	conditions", table 1
fgammaton.m	Use gammatone filterbank to decompose input signal and generate GF features
gammatone.m	Use a bank of gammatone filters to decompose input speech signal with equal
	loudness compression
<pre> gen_gammaton.m</pre>	Create the gammatone filterbank
gtf2gtfcc.m	Convert GF to GFCC
gtfeatures.m	Main file to batch generating GFs and GFCCs for a list of files
hz2erb.m	Convert linear scale to ERB scale
loadData.m	Load data in HTK format
loudness.m	Compute loudness level in Phons on the basis of equal-loudness functions
writeHTK.m	Write data in HTK format

3. Pitch Tracking

Folder: PitchTracking

Description: This part of the toolbox is used to estimate pitches of an input signal. The pitch tracking algorithm is proposed in

Zhaozhang Jin and DeLiang Wang, "HMM-based multipitch tracking for noisy and reverberant speech," *IEEE Trans. on Audio, Speech and Language Processing*, vol. 19, 2011.

Pitch estimation is important because in the following mask estimation part, pitch-based features are used to classify time-frequency (T-F) units that are dominated by speech or noise.

Structure:

PitchTracking

JpitchHP_2pass.jar	Java programs that are developed by Zhaozhang Jin to do pitch tracking of the
	input signal
PitchTracking.m	Main program to batch tracking pitches of a list of files

4. Mask Estimation

Folder: MaskEstimation

Description: Programs in this folder are aimed at training multi-layer perception (MLP) classifiers to estimate CASA masks. The core idea is proposed in the following paper:

Zhaozhang Jin and DeLiang Wang, "A supervised learning approach to monaural segregation of reverberant speech," *IEEE Trans. on Audio, Speech and Language Processing*, vol. 17 2009.

First of all, a 6-dimensional pitch-based feature vector is derived for every single T-F unit. A MLP is trained for each frequency channel using the 6-d feature as input and IBM as desired output. Specifically, a universal MLP will be trained for each channel first across a variety of SNRs and noisy conditions. An initial mask will be derived for a mixture using universal MLPs. With this mask, input SNR is estimated and the closest SNR-specific MLPs (trained at the same time with universal MLPs) will be chosen to refine mask estimation. A folder called "RefineMask" in the section 6 actually sequentially does mask refinement on the initial masks generated by universal MLPs.

Structure:

MaskEstimation

|-- Code

Ι	GetTestFeatures.m	Batch generating 6-dimensional pitch-based features for test data
Ι	GetTrainFeatures.m	Batch generating 6-dimensional pitch-based features for training data
		of mask estimation
Ι	16 kHz	All the necessary java source and class files for 6-dimensional pitch-
		based feature generation for 16 kHz sampling frequency signals
	*.java	

8 kHzAll the necessary java source and class files for 6-dimensional pitch- based feature generation for 8 kHz sampling frequency signals	*.class	
based feature generation for 8 kHz sampling frequency signals	8 kHz	All the necessary java source and class files for 6-dimensional pitch-
		based feature generation for 8 kHz sampling frequency signals
*.java	*.java	
*.class	*.class	
praat_pd.m Program to get ground truth pitch value	praat_pd.m	Program to get ground truth pitch value
testAll.m Use universal MLPs to derive voiced masks for a number of files	testAll.m	Use universal MLPs to derive voiced masks for a number of files
testSNR.m Use SNR-specific MLPs to derive voiced masks for a number of files	testSNR.m	Use SNR-specific MLPs to derive voiced masks for a number of files
trainAll.m Train universal MLPs for every channel	trainAll.m	Train universal MLPs for every channel
trainSNR.m Train SNR-specific MLPs for every channel	trainSNR.m	Train SNR-specific MLPs for every channel
UsePraat.m Batch generating ground truth pitch contours for a number of files	UsePraat.m	Batch generating ground truth pitch contours for a number of files

5. Model Training

Folder: ModelTraining

Description: This part of the toolbox trains speaker models by adapting a universal background model (UBM). The main idea is proposed by Reynolds et al. in

Reynolds et al., "Speaker verification using adapted Gaussian mixture models", *Digital Signal Processing*, vol. 10, 2000.

We assume you have experience of using HTK to train a gigantic UBM. If not, please refer to HTK book or anyone familiar with HTK-based speech recognition for how to train a Gaussian mixture model (GMM).

Structure:

ModelTraining

AdaptGMM.m	Adapt the UBM to get a speaker model. During this adaptation process,
	only means of Gaussian components are adapted
calMixtureLL.m	Calculate likelihood of an observation given a Gaussian component
GetPosterior.m	Get posterior probability of each Gaussian component given an
	observation
loadData.m	Load data in HTK format
loadGMM.m	Load a GMM created by HTK in a HMM prototype
spkrList	A list of enrolled speakers
TrainGMM.m	Main file of training speaker models

6. Evaluation

Folder: testing

Description: This folder has main programs to perform robust SID. It can evaluate not only the proposed combined system, but also individual modules as well as baseline systems.

Structure:

testing

calGMMLL_hub_AdaptGMM.m Calculate the SID score of an input signal given a speaker model	
calMixtureLL.m	Calculate likelihood of a frame given a Gaussian component
CTL	Main folder to store file lists
DetectSNR.m	Estimate input SNR of a mixture
GetMLP.m	Get the closest SNR-specific MLPs given an input SNR
GetRunningMask.m	Use selected SNR-specific MLPs to do finer mask estimation
getSID_ssn.m	Get the ground truth speaker ID
gtf2gtfcc.m	Program to convert GF to GFCC
loadData.m	Load data in HTK format
loadGMM.m	Load a GMM created by HTK under its HMM definition
MDL	Main folder to store model files (speaker models, UBMs, etc.)
preCptLBound.m	Pre-compute lower bound of bounded marginalization to speed up score calculation
readmodelfile_diagonal.m	Read a speech prior model / UBM with diagonal covariance matrix provided no Gaussian component is dropped by HTK during training
readPrior.m	Read a speech prior model / UBM with diagonal covariance matrix provided some Gaussian components are dropped by HTK during training
recon_gtf.m	Perform reconstruction on GF
RefineMask	
WriteMask.m	This program batch refines initial masks created by universal MLPs
RmBottom10.m	Remove the first 10 channels of GF as they correspond to frequency
	range below 200 Hz. This processing is for telephone recordings and it is
	unnecessary for high quality microphone recordings
speakerID_hub_AdaptGMN	1_Comb.m Main file to get SID scores using the proposed combined
	system
<pre> speakerID_hub_AdaptGMN</pre>	
spkrList	A list of enrolled speakers
testBaseline	
main_etsi.m	Main file of baseline system using ETSI_D features
main_gfcc.m	Main file of baseline system using GFCC features

	Main file of baseline system using 12-dimensional MFCC features
main_mfc.m	Main file of baseline system using 22-dimensional MFCC features
testClean	
main_ETSI.m	Main file of a SID system using clean ETSI_D features
main_GFCC.m	Main file of a SID system using clean GFCC features
main_GTF.m	Main file of a SID system using clean GF features
	Main file of a SID system using clean 12-dimensional MFCC features
main_MFC.m	Main file of a SID system using clean 22-dimensional MFCC features
<pre> testCombination</pre>	
main_Combination.m	Main file of the proposed combined system
testIndividualModule	
main_REC.m	Main file of the proposed reconstruction module
main_MAR.m	Main file of the proposed marginalization module
var2ccVar.m	Program to convert spectral uncertainties to cepstral uncertainties

7. How to use this toolbox?

Each part of this toolbox is relatively independent. You can easily find main programs in each folder to start with. Be careful with the programs as there are places that you need to make necessary changes such as designated folders. These programs have been reorganized based on our initial code so there could be some minor bugs, though we have done a quick sanity check to fix any obvious errors. Please feel free to contact us if you have any problem using the code.