

Speech Recognition

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1. Introduction

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2. Dynamic Time Warping

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2.1 Algorithm

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2.2 Modifications

Major modifications are the step size conditions, step weighting and the global path constraints.

2.3 Applications

3. Speech Recognition

We implemented a speech recognizer using a modified version of the DTW algorithm. Using DTW and speech processing techniques (i.e. Mel Frequency Cepstral Coefficients), our algorithm is able to detect which action we want to perform: open the google search or open our facebook webpage.

This speech recognizer takes as input a sound file and compares it with some template words: *google* and *facebook*. In order to make our recognizer more robust, we have used two template words for each sound. After comparing the input sound with each template word, the algorithm chooses the one with the shortest path.

3.1 Speech Processing

Since each sound has a different length, DTW is a good algorithm for comparing them. However, sound waves are sinusoids and cannot be properly compared on the time domain (amplitude vs seconds). The reason is simple, a sound wave of a word can change radically from a person to another, and even the sound waves of a word generated by a single person can be very different.

If we observe the sound waves of our sounds, we see their differences.

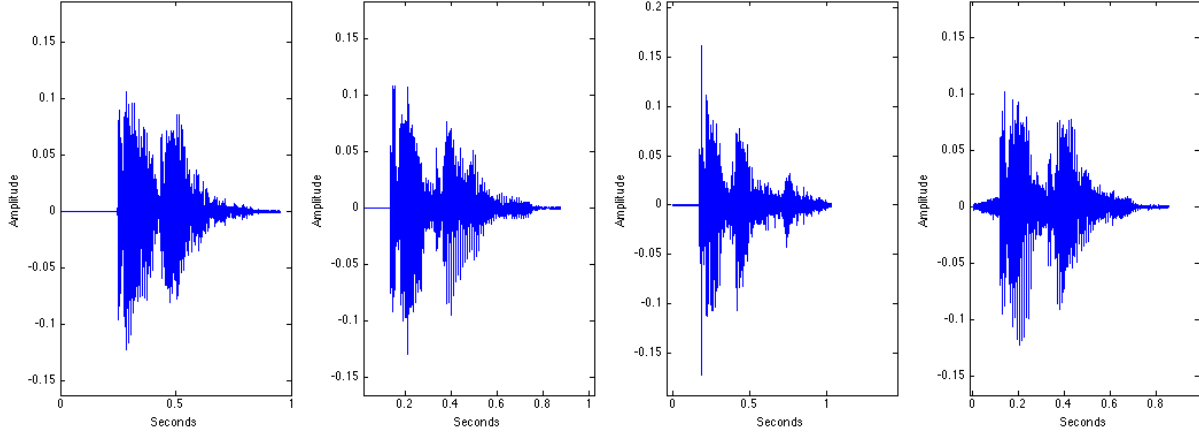


Figure 1. Sound Waves of the word Google.

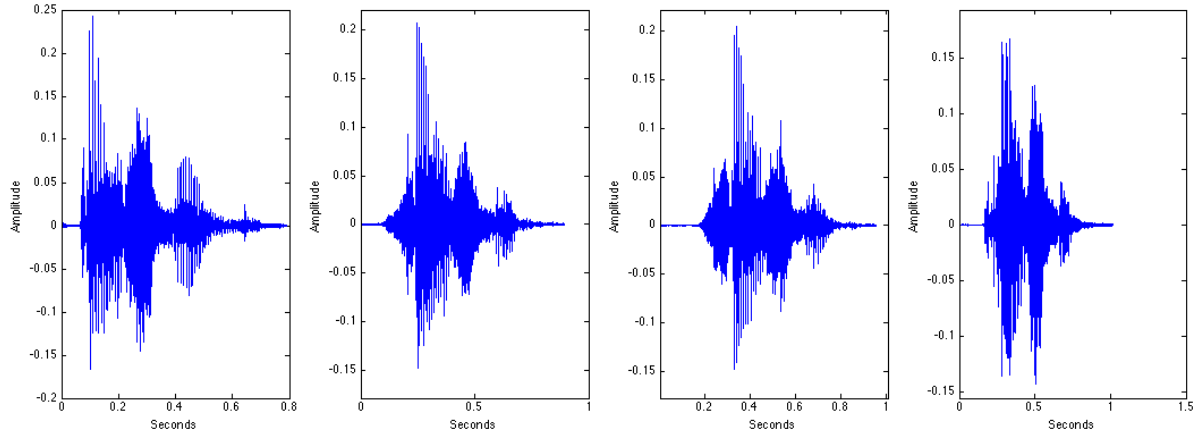


Figure 2. Sound Waves of the word Facebook.

An extended approach is to transform the sound waves into the frequency domain and then, extract some coefficients ([1], [2]). In this project, we tried with two different coefficients: Mel Frequency Cepstral Coefficients (MFCC) and Linear Prediction Coefficient (LPC). Unfortunately, we didn't obtained good results with LPC and we decided to use only MFCC.

1. Pre-emphasizing: In the process to digitalize audio, it passes through a low-pass filter and the high frequencies are attenuated. In this step a pre-emphasizing filter is applied to emphasise the highest frequency components of the speech, that is increase the energy in the higher frequencies.

$$y(n) = x(n) - 0.95x(n - 1)$$

2. Framing: From the previous images, we can clearly see that sound signals are not stationary. However, we can assume that if we take a small sample called frame (approximately of 10ms or 20 ms) it will be stationary. We created a set of frames of 16ms.

3. Windowing: In order to make the transition between frames smoother, a window is multiplied with each frame. The most used window for speech processing is a *Hamming Window*:

$$w(n) = 0.54 - \cos(\frac{2\pi n}{N-1})$$

where N is the length of the window, in this case the length of the frame.

3.2 Dynamic Time Warping in this project

Blablabla

3.3 Results

Blablabla

4. Conclusions

Blablabla

5. References

- [1] Anand, D. A., & Meher, P. K. Combined LPC and MFCC Features based technique for Isolated Speech Recognition.
- [2] Shinde, M. R., & Pawar, V. P. Dynamic time Warping using MATLAB & PRAAT.

6. Annex

6.1 DTW function

```
TimeWarp<-function(x,y,w=4){

  # define distance function
  distance<-function(a,b){
    dist(rbind(a,b))
  }

  # 1. Compute matrix 11xM

  # set parameters
  m<-dim(x)[2]
  n<-dim(y)[2]
  colnames(x)<-1:m
  colnames(y)<-1:n
  w = max(w, abs(n-m))

  # Create matrix
  DTW<-matrix(Inf,n,m)
  rownames(DTW)<-n:1
  colnames(DTW)<-1:m

  # Initial values
  DTW['1','1']<-distance(x[, '1'], y[, '1'])

  # First row
  for(j in 2:(w+1)){
    cost<-distance(x[,as.character(j)], y[,as.character(1)])
    DTW['1',as.character(j)]<- cost + DTW['1', as.character(j-1)]
  }

  # First column
  for(i in 2:(w+1)){
    cost<-distance(x[,as.character(1)],y[,as.character(i)])
    DTW[as.character(i), '1']<- cost + DTW[as.character(i-1), '1']
  }

  # Fill matrix
  for(i in 2:n){
    for(j in (max(2, i-w)):(min(m, i+w))){

      #current cost
      cost<-distance(x[,as.character(j)], y[,as.character(i)])

      #cumulated cost
      d.cost<-min(DTW[as.character(i-1), as.character(j)] ,
                  DTW[as.character(i), as.character(j-1)],
                  2*DTW[as.character(i-1), as.character(j-1)])
    }
  }
}
```

```

    #combined cost
    DTW[as.character(i),as.character(j)]<-cost + d.cost

  }
}

# 2. Find path
path<-matrix(c(n,m), 1,2)
full.path<-(tail(path,1)[1] ==1 & tail(path,1)[2] ==1)

while(full.path==FALSE ){

  l.path<-tail(path,1)

  if(l.path[1]==1 | l.path[2]==1){
    p<-which(l.path==1)

    if(p==1){new.point<-c(l.path[1], l.path[2]-1)
    }else{
      new.point<-c(l.path[1]-1, l.path[2])
    }

  } else {

    # nearest point
    min.step<-min(DTW[as.character(l.path[1]-1), as.character(l.path[2]-1)],
      DTW[as.character(l.path[1]), as.character(l.path[2]-1)],
      DTW[as.character(l.path[1]-1), as.character(l.path[2])])
    min.step<-which(c(DTW[as.character(l.path[1]-1), as.character(l.path[2]-1)],
      DTW[as.character(l.path[1]), as.character(l.path[2]-1)],
      DTW[as.character(l.path[1]-1), as.character(l.path[2])])==min.step)
    min.step<-min.step[1]

    #path to nearest point
    if(min.step==1){
      new.point<-c(l.path[1]-1, l.path[2]-1)
    } else{
      if(min.step==2){
        new.point<-c(l.path[1], l.path[2]-1)
      } else{
        new.point<-c(l.path[1]-1, l.path[2])
      }
    }
  }
}
path<-rbind(path,new.point)
full.path<-(tail(path,1)[1] ==1 & tail(path,1)[2] ==1)

}

return(list(path=path, DTW=DTW))
}

```

6.2 Speech Recognizer code

```
# input: isound is the path to the wav file with the sound.

SpeechRecognizer <- function(isound){

  if (!require("tuneR")) install.packages("tuneR");library(tuneR)

  # Read the wav file
  sound <- readWave(isound)
  sr    <- sound@samp.rate

  # Compute the mel frequency cepstrum coefficients
  inputWord <- t(melfcc(sound,
                        sr,
                        wintime=0.016,
                        lifterexp=0,
                        minfreq=133.33,
                        maxfreq=6855.6,
                        sumpower=FALSE))

  # Upload the four template sounds and compute their melfcc
  g1 <- readWave("Project\google1.wav")
  g2 <- readWave("Project\google2.wav")
  f1 <- readWave("Project\facebook1.wav")
  f2 <- readWave("Project\facebook2.wav")

  sr1 <- g1@samp.rate
  sr2 <- g2@samp.rate
  sr3 <- f1@samp.rate
  sr4 <- f2@samp.rate

  google1 <- t(melfcc(g1, sr1, wintime=0.016, lifterexp=0,
                     minfreq=133.33, maxfreq=6855.6, sumpower=FALSE))
  google2 <- t(melfcc(g2, sr2, wintime=0.016, lifterexp=0,
                     minfreq=133.33, maxfreq=6855.6, sumpower=FALSE))
  facebook1 <- t(melfcc(f1, sr3, wintime=0.016, lifterexp=0,
                      minfreq=133.33, maxfreq=6855.6, sumpower=FALSE))
  facebook2 <- t(melfcc(b2, sr4, wintime=0.016, lifterexp=0,
                      minfreq=133.33, maxfreq=6855.6, sumpower=FALSE))

  # Compute the distance of the input sound with the template sounds
  distance.sound<-rep(NA, 4)

  dtwg1 <- TimeWarp(google1, inputWord)
  distance.sound[1]<- tail(dtwg1$DTW[,1],1)

  dtwg2 <- TimeWarp(google2, inputWord)
  distance.sound[2]<- tail(dtwg2$DTW[,1],1)

  dtwf1 <- TimeWarp(facebook1, inputWord)
  distance.sound[3]<- tail(dtwf1$DTW[,1],1)
```

```

dtwf2 <- TimeWarp(facebook2, inputWord)
distance.sound[4]<- tail(dtwf2$DTW[,1],1)

# If the minimum distance is to the word gmail, open gmail
if (which.min(distance.sound) == 1 | which.min(distance.sound) == 2){
  system(paste("open http://google.com"))
}

# If the minimum distance is to the word facebook, open facebook
if (which.min(distance.sound) == 3 | which.min(distance.sound) == 4){
  system(paste("open http://facebook.com"))
}
}

```