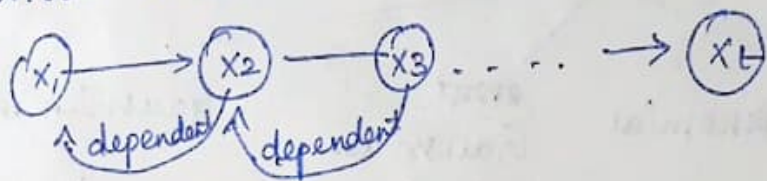


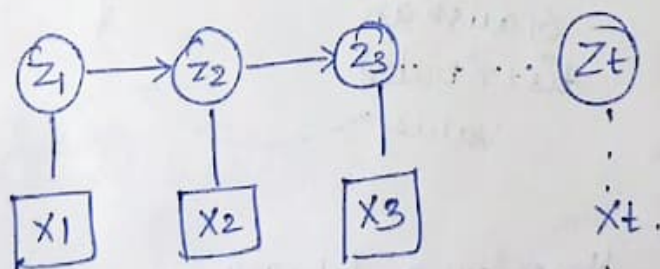
Unit-2 12marks

① Discuss in detail about Building Hidden markov models for sequential data using Python.

→ Markov process: Stochastic model describing sequence of possible events in which the probability of each event depends on the previous state.



→ Hidden markov model most important modification to markov process  $\Rightarrow$  the actual system states assumes to be unobservable or hidden.



unable to observe system  $Z$  &  
can only observe the observable  
Process  $X$ .

## Important parameters

→ Initial hidden state probability

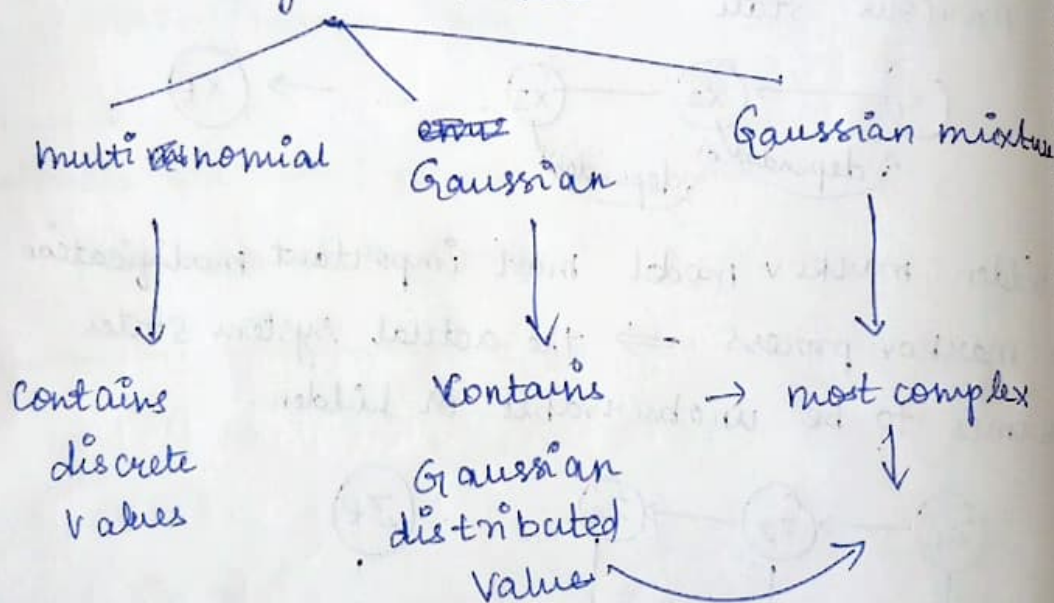
$$\pi = [\pi_0, \pi_1, \pi_2 \dots]$$

→ Hidden state transition matrix is  $A$ .

→ Observable probability  $\theta = [\theta_0, \theta_1, \theta_2 \dots]$

→ The `hmmlearn` library → implements HMM in python.

It gives 3 models



~~Hmm model~~ Hmmlearn library

```
def __init__(self, model_name="Gaussian  
HMM",  
n_components=4, cov_type='diag')
```

In HMM we use Viterbi algorithm.

↓  
max probability to any path  
at time  $t$ .

② Extract the features of Audio signal using MFCC.

→ MFCC → widely used extracting audio signals

→ Steps:

Pre-processing:

- framing into short intervals.
- usually overlapped for continuity.

windowing:

each frame is windowed by Hamming/  
Hanning.

FFT: Fast fourier transform is applied.

Mel filterbank: linear freq. to non-linear  
freq.

Logarithmic compression: - compress the signals.

Discrete cosine transform: is applied.

normalization: normalised to have zero mean

Feature extraction:



③ Build a speech recognition model using Hidden Markov model.

Step 1 - collection & pre-processing of data

↓  
MFCC

Step 2: Define HMM topology

↓  
defining structure of the model  
Probabilities

Step 3: Training using Baum-welch algo

Step 4: Evaluate. used to find the unknown parameters

Step 5: Use the speech recogni

```
import numpy as np
```

```
import hmmlearn.hmm as hmm
```

```
# loading dataset.
```

```
train_data = np.load('train_data.npy')
```

```
# Define HMM
```

```
n_components = 5
```

```
model = hmm.GaussianHMM(n_components =  
                          n_components)
```

```
# Train data
```

```
model.fit(train_data)
```

Stock.

# test data

test\_data = np.load('test\_data.npy')

#### ④ Stock market Prediction using HMM

Step 1 : Data collection : historical stock  
Prices, financial

→ choosing hidden states depending  
on the complexity.

Since its stock market we need 2  
bull / bear (up & down)

→ 'stock.npy'

## ⑤ Speech recognition mechanism for infotainment system.

→ Both h/w & s/w

Step 1:

Suitable mic



Suitable voice recognition Package



Userfriendly UI for user



from mic to s/w (a good software needed)



Intent recognition  
(adjusting volumes)



System response



error handling



Import Speech-recognition as sr

r = sr.recogniser()

with sr.mic() as source:

print("Say something")

audio = r.listen(source)

---

U-3

~~Explain the building conditional Random fields for sequential text data.~~

Describe the time series and sequential data series with examp.

Time series:

→ Data collected over time, where obs is recorded at regular intervals.

→ order imp since it must go by time

Step-1 pre processing

Step-2 Choosing model (ARIMA)/(HMM).

Step-3 - Fit the model.

Step 4 - Evaluate performance

Eg: Stock market explain

## Sequential data:

→ Obs are  $x_i$ .

→ mostly NLP like Speech recog.

→ To build

Step 1: collection & pre-processing

Step 2: Choose modelling approach.

Step 3: Build the model.

Step 4: Train & evaluate

Eg: ~~spe~~ audio Speech recognition

explain

---

## Building CRF for Sequential text data

→ CRF are class of probabilistic model used

to make predictions based on  
Sequential data.

→ Usually uses NLP.

Steps

1) collection & preprocessing data

splitting into words/numbers.



Step 2:

Define the features. they may include current / next / previous words / letters.

Step 3:

Split into training / test data.

Step 4:

Build using python library  
such as pystruct

Step 5:

Evaluate by accuracy, precision, recall  
& F1-Score