Quiz 3 - Latent Variable Models in Financial Asset Regime Detection

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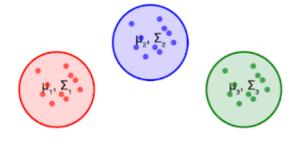
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From GMMs to HMMs

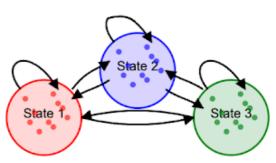
What is the main limitation of Gaussian Mixture Models (GMMs) when modeling sequential data?

1 point

Gaussian Mixture Model (GMM)



Hidden Markov Model (HMM)



- GMMs assume a fixed number of hidden states
- GMMs require labeled data
- GMMs ignore temporal dependencies

In a Hidden Markov Model, the current observation X_t is conditionally independent of all $\ \ _1$ point other variables given the current hidden state H_t

- True
- False

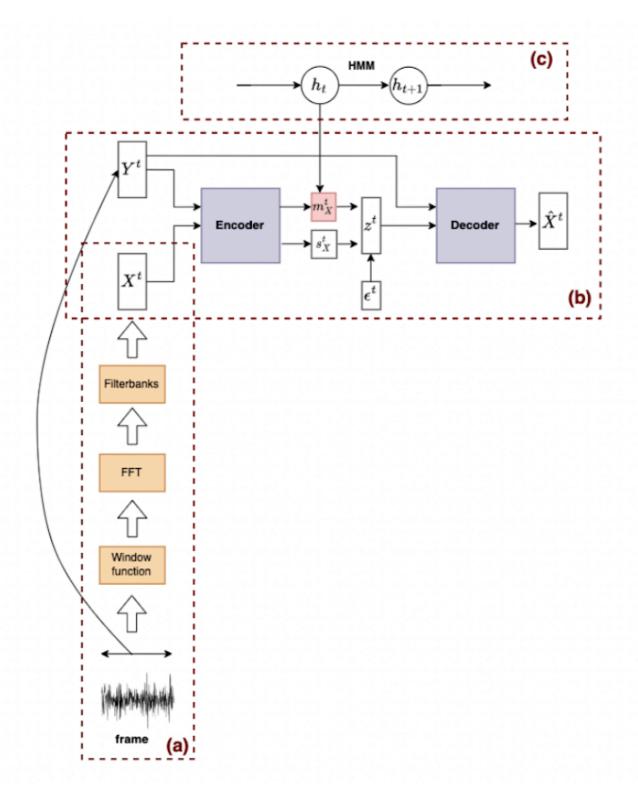
HMM Parameterization

In the sandwich example, if Ross observes his sandwich is safe today, what is the most 1 point likely hidden state?



- State 0
- State 1
- State 2

| For an HMM with M hidden states and D possible observations, how many parameters are needed to fully specify the model with discrete emissions? | 1 point |
|-------------------------------------------------------------------------------------------------------------------------------------------------|---------|
| $ M + M^2 + MD $ | |
| \bigcirc M + M + MD | |
| \bigcirc M × M × D | |
| | |
| Which parameter set fully specifies an HMM with continuous Gaussian emissions? | 1 point |
| O π, Q, O | |
| π, Q, μ, Σ | |
| Ο π, μ, Σ | |
| | |
| | |
| HMMs - Estimation Problems | |

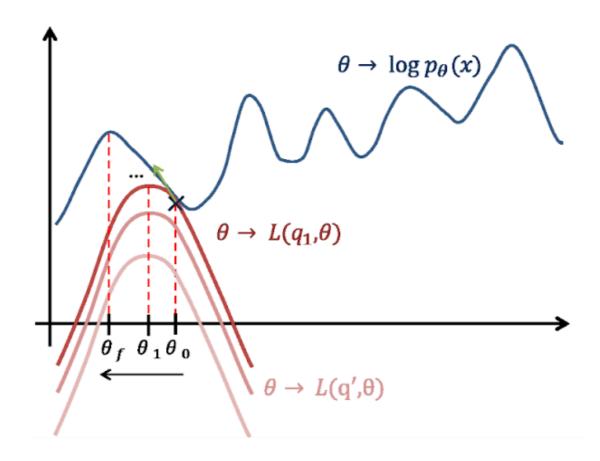


- O To extract spectral information from return data
- O To select individual assets
- To forecast market regimes

- lt efficiently computes the likelihood of an observation sequence without enumerating all possible hidden paths
- It directly estimates the model parameters from the observed data

What is the fundamental purpose of using the EM algorithm for HMMs?

1 point



- To reduce the computational complexity of the Forward-Backward algorithm
- O To convert continuous emissions to discrete emissions for easier processing
- To estimate the parameters of the HMM

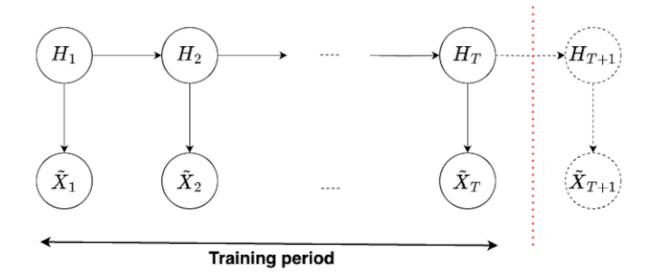
In the Expectation-Maximization (EM) algorithm for Hidden Markov Models, which probabilities are calculated and used during the E-step?

1 point

- Filtering probabilities only
- Smoothing probabilities only
- Neither filtering nor smoothing probabilities

In a Hidden Markov Model, after observing a sequence up to time T, which specific probabilities are used to predict the distribution of the next hidden state (at T+1)?

1 point



- All filtering probabilities $\xi(1,\cdot)$, $\xi(2,\cdot)$, ..., $\xi(T,\cdot)$ from time 1 to T
- Only the most recent filtering probability $\xi(T,\cdot)$
- The smoothing probabilities $\psi(T,\cdot)$

Questions

Any comment ?

