

Student Name: TODO

Collaboration Statement:

Total hours spent: TODO hours

I consulted the following resources:

- TODO

Contents

1a: Problem Statement

CS 136 - 2022s - HW-CP5

Prove the following property under a Hidden Markov Model.

$$p(z_{t+1}|x_t, z_t) = p(z_{t+1}|z_t) \quad (1)$$

1a: Solution

TODO

1b: Problem Statement

CS 136 - 2022s - HW-CP5

Prove the following property under a Hidden Markov Model.

$$p(x_{t+1}|x_{1:t}, z_{1:t}) = p(x_{t+1}|z_t) \quad (2)$$

1b: Solution

TODO

2a: Problem Statement

CS 136 - 2022s - HW-CP5

Write out an expression for the expected complete log likelihood:

$$\mathbb{E}_{q(z_{1:T}|s)} [\log p(z_{1:T}, x_{1:T}|\theta)] \quad (3)$$

Use the HMM probabilistic model $p(z_{1:T}, x_{1:T}|\theta)$ and the approximate posterior $q(z_{1:T}|s)$ defined above.

Your answer should be a function of the data x , the local sequence parameters s and $r(s)$, as well as the HMM parameters π, A, ϕ .

2a: Solution

TODO

2b: Problem Statement

CS 136 - 2022s - HW-CP5

Provide a short verbal summary of the update for ϕ_{kd} given below. How should we interpret the numerator? The denominator?

$$\phi_{kd} = \frac{\sum_{t=1}^T r_{tk} x_{td}}{\sum_{t=1}^T r_{tk}} \quad (4)$$

2b: Solution

TODO