

COMP3670: Introduction to Machine Learning

Note: For the purposes of this assignment, we let lowercase p denote probability density functions (pdf's), and upper case P denote probabilities. If a random variable Z is characterized by a probability density function p , we have that

$$P(a \leq Z \leq b) = \int_a^b p(z) dz$$

You should show your derivations, but **you may use a computer algebra system (CAS) to assist with integration or differentiation**.¹

Question 1 (40 credits) Bayesian Inference

Let X be a random variable representing the outcome of a biased coin with possible outcomes $\mathcal{X} = \{0, 1\}$, $x \in \mathcal{X}$. The bias of the coin is itself controlled by a random variable Θ , with outcomes² $\theta \in \boldsymbol{\theta}$, where

$$\boldsymbol{\theta} = \{\theta \in \mathbb{R} : 0 \leq \theta \leq 1\}$$

The two random variables are related by the following conditional probability distribution function of X given Θ .

$$p(X = 1 \mid \Theta = \theta) = \theta$$

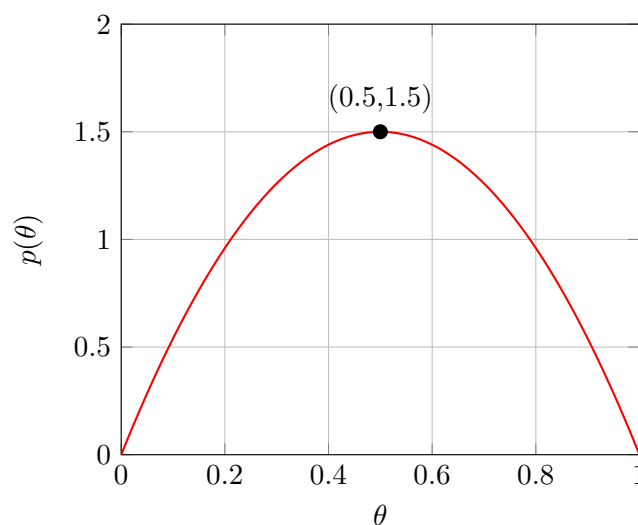
$$p(X = 0 \mid \Theta = \theta) = 1 - \theta$$

We can use $p(X = 1 \mid \theta)$ as a shorthand for $p(X = 1 \mid \Theta = \theta)$.

We wish to learn what θ is, based on experiments by flipping the coin. Before we flip the coin, we choose as our prior distribution

$$p(\theta) = 6\theta(1 - \theta)$$

which, when plotted, looks like this:



¹For example, asserting that $\int_0^1 x^2 (x^3 + 2x) dx = 2/3$ with no working out is adequate, as you could just plug the integral into Wolfram Alpha using the command `Integrate[x^2(x^3 + 2x), {x, 0, 1}]`

²For example, a value of $\theta = 1$ represents a coin with 1 on both sides. A value of $\theta = 0$ represents a coin with 0 on both sides, and $\theta = 1/2$ represents a fair, unbiased coin.

- a) (3 credits) Verify that $p(\theta) = 6\theta(1 - \theta)$ is a valid probability distribution on $[0, 1]$ (i.e that it is always non-negative and that it is normalised.)

We flip the coin a number of times.³ After each coin flip, we update the probability distribution for θ to reflect our new belief of the distribution on θ , based on evidence.

Suppose we flip the coin twice, and obtain the sequence of coin flips⁴ $x_{1:2} = 00$. For each subsequence $x_1, x_{1:2}$ (and for the case before any coins are flipped), compute the:

- b) (15 credits) probability distribution functions
 c) (3 credits) expectation values μ
 d) (3 credits) variances σ^2
 e) (5 credits) The *maximum a posteriori* estimation θ_{MAP} .

Present your results in a table like as shown below.

Posterior	PDF	μ	σ^2	θ_{MAP}
$p(\theta)$	$6\theta(1 - \theta)$?	?	?
$p(\theta x_1 = 0)$?	?	?	?
$p(\theta x_{1:2} = 00)$?	?	?	?

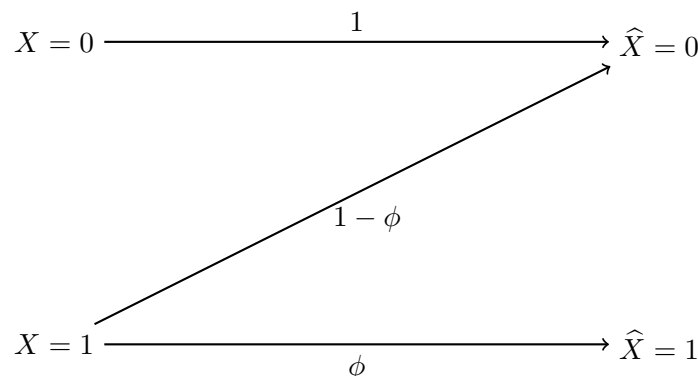
- f) (5 credits) Plot each of the probability distributions $p(\theta)$, $p(\theta|x_1 = 0)$, $p(\theta|x_{1:2} = 00)$ over the interval $0 \leq \theta \leq 1$ on the same graph to compare them.
- g) (6 credits) What behaviour would you expect of the posterior distribution $p(\theta|x_{1:n})$ if we updated on a very long sequence of alternating coin flips $x_{1:n} = 10101010 \dots$?
 What would you expect $\mu, \sigma^2, \theta_{MAP}$ to look like for large n ?
 Sketch/draw an estimate of what $p(\theta|x_{1:n})$ would approximately look like against the other distributions.

Question 2

Bayesian Inference on Imperfect Information

(50 credits)

We have a Bayesian agent running on a computer, trying to learn information about what the parameter θ could be in the coin flip problem, based on observations through a noisy camera. The noisy camera takes a photo of each coin flip and reports back if the result was a 0 or a 1. Unfortunately, the side of the coin with a "1" on it is very shiny, and the reflected light causes the camera to sometimes report back the wrong result.⁵ The probability that the camera correctly reads a one is parameterised by $\phi \in [0, 1]$. The camera always correctly identifies zeros. Letting X denote the true outcome of the coin, and \hat{X} denoting what the camera reported back, we can draw the relationship between X and \hat{X} as shown.



³The coin flips are independent and identically distributed (i.i.d.).

⁴We write $x_{1:n}$ as shorthand for the sequence $x_1 x_2 \dots x_n$.

⁵The errors made by the camera are i.i.d, in that past camera outputs do not affect future camera outputs.

So, we have

$$\begin{aligned} p(\hat{X} = 0 \mid \phi, X = 0) &= 1 \\ p(\hat{X} = 0 \mid \phi, X = 1) &= 1 - \phi \\ p(\hat{X} = 1 \mid \phi, X = 1) &= \phi \\ p(\hat{X} = 1 \mid \phi, X = 0) &= 0 \end{aligned}$$

We would now like to investigate what posterior distributions are obtained, as a function of the parameter ϕ . Let $\hat{x}_{1:n}$ be a sequence of coin flips as observed by the camera.

- a) (5 credits) Briefly comment about how the camera behaves for $\phi = 0, \phi = 0.5, \phi = 1$. How you expect this would change how the agent updates its prior to a posterior on θ , given an observation of \hat{X} . (No equations required.)
- b) (10 credits) Compute $p(\hat{X} = x \mid \theta, \phi)$ for all $x \in \{0, 1\}$.
- c) (15 credits) The coin is flipped, and the camera reports seeing a zero. (i.e. that $\hat{x}_1 = 0$.) Given the same choice of prior $p(\theta \mid \phi) = 6\theta(1 - \theta)$ as before, compute the posterior $p(\theta \mid \hat{x}_1 = 0, \phi)$. What term (from Question 1) does $p(\theta \mid \hat{x}_1 = 0, \phi)$ simplify to when $\phi = 1$? When $\phi = 0$? Explain your observations.
- d) (10 credits) The experiment is reset. The coin is flipped, and the camera reports seeing a one. (i.e. that $\hat{x}_1 = 1$.) Given the same choice of prior $p(\theta \mid \phi) = 6\theta(1 - \theta)$ as before, compute the posterior $p(\theta \mid \hat{x}_1 = 1, \phi)$. Comment on how the result depends on ϕ . Does the result make sense?
- e) (10 credits) Plot $p(\theta \mid \hat{x}_1 = 0, \phi)$ as a function of θ , for all $\phi \in \{0, 0.2, 0.4, 0.6, 0.8, 1\}$ on the same graph to compare them. Comment on how the shape of the distribution changes with ϕ . Explain your observations.

Question 3

Relating Random Variables

(10 credits)

Let X be a random variable, on $[0, 1]$, with probability density function

$$p(x) = 2 - 2x$$

Let Y be a random variable on $[1, 2]$, such that $Y = X^2 + 1$. Find the probability density function for Y .