

University of Technology Sydney
School of Mathematical and Physical Sciences

Probability and Random Variables (37161) –
Class 6 Preparation Work
SOLUTIONS

1.

- i) If the first flip is Heads (i.e. $T = 1$) then $E(X) = 1$ hence $E(X | T = 1)$.
- ii) If the first flip is Tails (i.e. $T = 0$) then we have added one flip to the total but are no closer to obtaining the first Heads, hence $E(X | T = 0) = 1 + E(X)$.

iii) Conditioning on this first flip, we have that

$$E(E(X | T)) = E(E(X | T = 1))P(T = 1) + E(E(X | T = 0))P(T = 0).$$

As $T \sim \text{Bern}(p)$, we have $P(T = 1) = p$ and $P(T = 0) = 1 - p$.

Substituting this (along with parts i and ii) gives .

$$E(E(X | T)) = p + (1 - p)[E(X) + 1]$$

- iv) The Law of Total Expectation tells us that $E(E(X | T)) = E(X)$ hence $E(X) = p + (1 - p)[E(X) + 1]$.

Rearranging gives $E(X) - (1 - p)E(X) = 1$ hence that $E(X) = \frac{1}{p}$.

- b) The coin is now flipped n times and Y is defined to be the number of times (out of n) that the coin lands Heads. That is $Y \sim \text{Bin}(n, p)$.

Again, let T be the number of Heads shown on the first flip of the coin.

- i) If we know the outcome of the first flip, the remaining uncertainty arises from the next (unknown) $n - 1$ flips.

Both $\text{Var}(Y | T = 0)$ and $\text{Var}(Y | T = 1)$ therefore equal $(n - 1)p(1 - p)$.

- ii) $E(\text{Var}(Y | T)) = (n - 1)p(1 - p)$.

- iii) Once we have observed the first outcome (i.e. have knowledge of the value of T), all of our remaining uncertainty about the value of Y comes from the fact that we have $n - 1$ additional flips which we have not yet observed. $E(Y | T)$ is therefore equal to T

(i.e. 0 or 1) plus however many heads are obtained from the remaining $n - 1$ flips.

This is because both are equivalent to observing an additional $n - 1$ flips and adding on the first outcome (either 0 or 1.)

iv) Taking the variance of the result from iii gives

$Var(E(Y | T)) = Var(T + (n - 1)p) = Var(T)$ since $(n - 1)p$ is constant and does not add to the variance (the only uncertainty is in the value of T .) As $T \sim Bern(p)$, we have that

$$Var(E(Y | T)) = Var(T) = p(1 - p).$$

Part ii gives that $E(Var(Y | T)) = (n - 1)p(1 - p)$. Summing these two terms gives

$$Var(Y) = E(Var(Y | T)) + Var(E(Y | T)) = (n - 1)p(1 - p) + p(1 - p) = np(1 - p)$$

as required for $Y \sim Bin(n, p)$.

2. With change of variable $V = Z^3$, we have that $\frac{dv}{dz} = 3z^2 = 3(v^{\frac{2}{3}})$.

We know that the density function of V is given by $f(z(v))\frac{dz}{dv}$

$$f(z(v))\frac{dz}{dv} = \left(3v^{\frac{2}{3}}\right)\left(\frac{1}{3v^{\frac{2}{3}}}\right) = 1.$$

As the range of Z is $[0, 1]$, this is also the range of V , since

$$V(0) = 0^2 = 0 \text{ and } V(1) = 1^2 = 1.$$

This gives $f(v) = \begin{cases} 1 & v \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$ i.e. $V \sim U[0, 1]$.