

Problem set 3: Summarizing distributions

Due October 16, 2023, at 9pm

(Your name here)

NOTE: Start with the file `ps3_2023_summarizing_distributions.qmd` (available from the github repository at <https://github.com/UChicago-pol-methods/IntroQSS-F23/tree/main/assignments>). Modify that file to include your answers. Make sure you can “render” the file (e.g. in RStudio by clicking on the **Render** button). Submit both the qmd file and the PDF via Canvas.

Question 1: Expected value

Consider the random variable X characterized by the following PMF.

x	$f(x)$
-1	.3
2	.2
3	.5

(1a) Compute $E[X]$. Show your work.

Answer:

$$\begin{aligned} E[X] &= \sum_x x f(x) \\ &= -1 \times .3 + 2 \times .2 + 3 \times .5 \\ &= 1.6 \end{aligned}$$

(1b) Write an R function to compute the expectation of any discrete random variable. The arguments to your function should include the values the random variable can take on (**x**) and the probability it takes on each value (**probs**). Use your function to confirm your answer from question (1a).

Answer:

```
exp_func <- function(x, probs){  
  sum(x * probs)  
}  
exp_func(x = c(-1, 2, 3), probs = c(.3, .2, .5))
```

```
[1] 1.6
```

(1c) Compute the MSE ($E[(X - c)^2]$) for $c = 0$ and $c = 1$. Show your work.

Answer:

$$\begin{aligned} E[(X - 0)^2] &= \sum_x (x - 0)^2 f(x) \\ &= (-1)^2 \times .3 + (2)^2 \times .2 + (3)^2 \times .5 \\ &= 1 \times .3 + 4 \times .2 + 9 \times .5 \\ &= 5.6 \end{aligned}$$

$$\begin{aligned} E[(X - 1)^2] &= \sum_x (x - 1)^2 f(x) \\ &= (-1 - 1)^2 \times .3 + (2 - 1)^2 \times .2 + (3 - 1)^2 \times .5 \\ &= 4 \times .3 + 1 \times .2 + 4 \times .5 \\ &= 3.4 \end{aligned}$$

(1d) Write a function to compute the MSE for any discrete random variable at a value c . The arguments to your function should include the values the random variable can take on (x), the probability it takes on each value ($probs$), and the value c being considered. Use your function to confirm your answers from (1c).

Answer:

```
mse_func <- function(x, probs, c){  
  sum(probs*(x - c)^2)  
}  
mse_func(x = c(-1, 2, 3), probs = c(.3, .2, .5), c = 0)
```

```
[1] 5.6
```

```
mse_func(x = c(-1, 2, 3), probs = c(.3, .2, .5), c = 1)
```

[1] 3.4

(1e) Create a vector `cs` that contains numbers in the sequence 1.0, 1.1, 1.2, . . . , 3. Using a for-loop and your function from (1d), compute the MSE for the random variable X whose PMF was given above at each value of `cs` and store the result in a vector called `mses`.

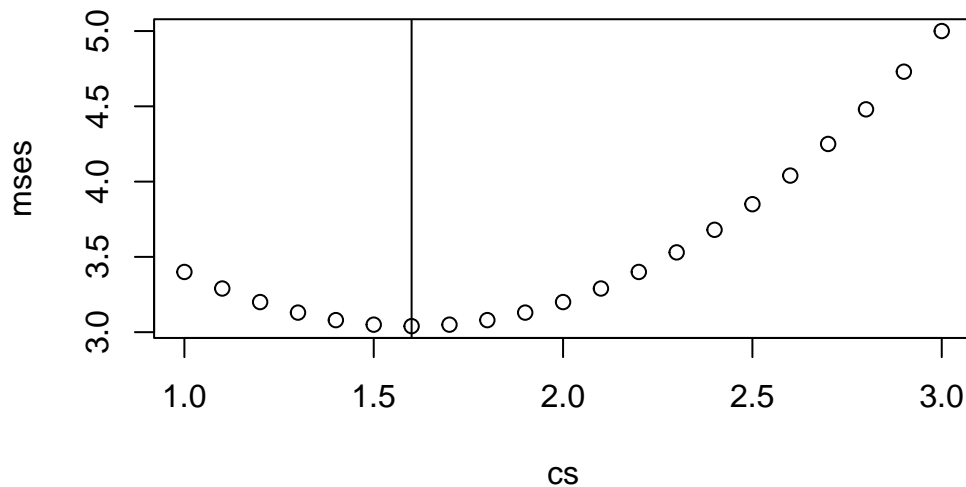
Answer:

```
cs <- seq(from = 1, to = 3, by = .1)
mses <- rep(NA, times = length(cs))
for(i in 1:length(mses)){
  mses[i] <- mse_func(x = c(-1, 2, 3), probs = c(.3, .2, .5), c = cs[i])
}
```

(1f) Using the `plot()` command, make a scatterplot showing the MSE of X (vertical axis) at each value of $c \in \{1, 1.1, 1.2, \dots, 3\}$ (horizontal axis). Use `abline()` to add a vertical line at $E[X]$.

Answer:

```
plot(cs, mses)
abline(v = 1.6)
```



Question 2: Variance

Consider the random variable X characterized by the following PMF:

x	$f(x)$
0	.6
1	.1
2	.3

(2a) Confirm that the variance of X is the same whether we compute it by the formula in Definition 2.1.12 or the Alternative Formula in Theorem 2.1.13. (We want the two variance computations, not the proof.)

Answer: First we calculate $E[X]$:

$$\begin{aligned} E[X] &= \sum_x x f(x) \\ &= 0 \times .6 + 1 \times .1 + 2 \times .3 \\ &= .7 \end{aligned}$$

Using Definition 2.1.12:

$$\begin{aligned} V[X] &= E[(X - E[X])^2] \\ &= E[(X - .7)^2] \\ &= (-.7)^2 \times .6 + .3^2 \times .1 + 1.3^2 \times .3 \\ &= .81 \end{aligned}$$

To use Theorem 2.1.13, we calculate $E[X^2]$:

$$\begin{aligned} E[X^2] &= \sum_x x^2 f(x) \\ &= 0 \times .6 + 1 \times .1 + 4 \times .3 \\ &= 1.3 \end{aligned}$$

Using Theorem 2.1.13:

$$\begin{aligned} V[X] &= E[X^2] - E[X]^2 \\ &= 1.3 - .7^2 \\ &= .81 \end{aligned}$$

(2b) Write an R function to compute the variance of any discrete random variable. The arguments to your function should include the values the random variable can take on (**x**) and the probability it takes on each value (**probs**). Use your function to confirm your answer from question (2a).

Answer:

```

# two versions, both of which use the expectation function I wrote above

# this version uses the  $E[(X - E[X])^2]$  formulation
var_func2 <- function(x, probs){
  ex <- exp_func(x, probs)
  exp_func((x - ex)^2, probs)
}

# this version uses the  $E[X^2] - E[X]^2$  formulation
var_func <- function(x, probs){
  exp_func(x^2, probs) - exp_func(x, probs)^2
}

# confirming equality
var_func(x = c(0, 1, 2), probs = c(.6, .1, .3))

```

[1] 0.81

```

var_func2(x = c(0, 1, 2), probs = c(.6, .1, .3))

```

[1] 0.81

(2c) Explain definition/property/mathematical operation is being used in each step of the following proof.

For random variable X and $c \in \mathbb{R}$, $V[X + c] = V[X]$,

$$V[X + c] = E[(X + c - E[X + c])^2] \quad (\text{Step 1})$$

$$= E[(X + c - E[X] - c)^2] \quad (\text{Step 2})$$

$$= E[(X - E[X])^2] \quad (\text{Step 3})$$

$$= V[X] \quad (\text{Step 4})$$

Answer:

In Step 1, . . .

In Step 2, . . .

In Step 3, . . .

In Step 4, . . .

Answer:

In Step 1, we use the definition of variance applied to the random variable $X + c$.

In Step 2, we use the linearity of expectations and the fact that the expectation of a constant c is c .

In Step 3, we simplify by observing that $c - c = 0$.

In Step 4, we use the definition of variance.

(2d) Use your variance function to confirm that $V[X + c] = V[X] \forall c \in \mathbb{R}$. (You can just show this for one value of c .)

Answer

```
var_func(x = c(0, 1, 2), probs = c(.6, .1, .3))
```

```
[1] 0.81
```

```
var_func(x = c(0, 1, 2) + 3, probs = c(.6, .1, .3))
```

```
[1] 0.81
```

(2e) Use your function to confirm that $V[aX] = a^2V[X] \forall a \in \mathbb{R}$. (You can just show this for one value of a .)

Answer

```
var_func(x = c(0, 1, 2), probs = c(.6, .1, .3))*9
```

```
[1] 7.29
```

```
var_func(x = c(0, 1, 2)*3, probs = c(.6, .1, .3))
```

```
[1] 7.29
```