

1. (a)

$$\begin{aligned}
 P(X < x) &= P(F^{-1}(U) < x) \\
 &= P(U < F(x)) \\
 &= F(x)
 \end{aligned}$$

So  $X \sim F$ .(b) For  $Z = X - Y$ :

$$\begin{aligned}
 P(Z = z) &= P(X - Y = z) \\
 &= \int_0^1 P(X = z + y | Y = y) P(Y = y) dy \\
 &= \int_0^1 I(0 < z + y < 1) dy \\
 &= \int_0^1 I(-z < y < 1 - z) dy \\
 &= \int_{\max\{0, -z\}}^{\min\{1, 1-z\}} dy \\
 &= \max\{0, |1 - z|\}
 \end{aligned}$$

For  $Z = \min\{X, Y\}$ :

$$\begin{aligned}
 P(Z < z) &= P(Z < z | Y \geq z) P(Y \geq z) + P(Z < z | Y < z) P(Y < z) \\
 &= P(X < z | Y \geq z) P(Y \geq z) + P(Y < z) \\
 &= I(0 < z < 1)(2z - z^2)
 \end{aligned}$$

(c)

$$\begin{aligned}
 E(Y) &= E(e^X) \\
 &= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{x - \frac{x^2}{2}} dx \\
 &= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{1}{2}(x-1)^2 + \frac{1}{2}} dx \\
 &= e^{\frac{1}{2}} \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{1}{2}(x-1)^2} dx \\
 &= e^{\frac{1}{2}}
 \end{aligned}$$

$$\begin{aligned}
 V(Y) &= V(e^X) \\
 &= E((e^X)^2) - E(e^X)^2 \\
 &= E(e^{2X}) - e \\
 &= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{2x - \frac{x^2}{2}} dx - e \\
 &= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{1}{2}(x-2)^2 + 2} dx - e \\
 &= e^2 \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{1}{2}(x-2)^2} dx - e \\
 &= e^2 - e
 \end{aligned}$$

2. (a) We can calculate the sum of squares for estimate  $\hat{\beta}$  as

$$\|Y - X\hat{\beta}\|^2 = Y'Y - 2Y'X\hat{\beta} - \hat{\beta}'X'X\hat{\beta}$$

Since this is a convex function, we will find the minimum when the gradient with respect to  $\hat{\beta}$  is 0:

$$\begin{aligned}\nabla_{\hat{\beta}}(Y'Y - 2\hat{\beta}'X'Y + \hat{\beta}'X'X\hat{\beta}) &= -2X'Y + 2X'X\hat{\beta} \\ 0 &= -2X'Y + 2X'X\hat{\beta} \\ X'Y &= X'X\hat{\beta} \\ (X'X)^{-1}X'Y &= \hat{\beta}\end{aligned}$$

So  $\hat{Y} = X\hat{\beta} = X(X'X)^{-1}X'Y = HY$  is the least squares estimate.

(b)

$$HX = X(X'X)^{-1}X'X = X$$

(c)

$$\begin{aligned}H' &= (X(X'X)^{-1}X')' \\ &= X(X(X'X)^{-1})' \\ &= X((X'X)^{-1})'X' \\ &= X((X'X)')^{-1}X' \\ &= X(X'X)^{-1}X' \\ &= H\end{aligned}$$

(d)

$$H^2 = X(X'X)^{-1}X'X(X'X)^{-1}X' = X(X'X)^{-1}X' = H$$

- (e) Since  $HY$  is a linear combination of the columns of  $X$ , it must be in the column space of  $X$ . Further, since

$$X'HY = X'X(X'X)^{-1}X'Y = X'Y$$

We can conclude it is actually the projection of  $Y$  into this space.

- (f) Using the SVD of  $X = U\Sigma V$ :

$$\begin{aligned}\text{rank}(X) &= \text{rank}(U) \\ &= \text{rank}(UU') \\ &= \text{rank}(H) \\ &= \text{tr}(H) \\ &= d\end{aligned}$$

The last step is since the rank of an idempotent matrix (which  $H$  is) is its trace.

3. 1. There are 30 rolls in the data set  
 2. The data is stored in unicode format, so the '+' operator does a string join instead of addition. The end result is a string with all the numbers in order.  
 3. The sum this time is correct and gives us a total of 92  
 4. Running this on my home computer, the first time took .00472 seconds, and the second .00352  
 5. I didn't do the hello world part because I can't get on AWS for the moment, but the reduce is the same as count()