

Consider the following dummy data matrix:

4 # observations

2 # Features

$$X = \begin{bmatrix} 0.5 & -1 \\ 0.5 & 1 \\ -0.5 & 1 \\ -0.5 & -1 \end{bmatrix} \quad 4 \times 2$$

- 10 I. Derive the estimated covariance matrix. Show all your work.
- 10 II. Find the eigen values and eigen vectors of the estimated covariance matrix. Show all your work.
- 5 III. Sketch the eigen vectors in the previous question on the cartesian coordinate. Show the eigen value corresponding to each eigen vector on the sketch.
- 5 IV. In which direction, the above data matrix has the maximum and minimum variance.
- 5 V. If the estimated covariance matrix multiplies the data inside the unit disk [shown below], draw [roughly] the shape of the product.

i

$$\text{Step 1: } \tilde{X} = X - \bar{X} = X$$

$$\text{Step 2: } \hat{C} = \frac{1}{N-1} \tilde{X}^T \tilde{X} = \frac{1}{3} \left(\begin{bmatrix} 0.5 & 0.5 & -0.5 & -0.5 \\ -1 & 1 & 1 & -1 \end{bmatrix} \begin{bmatrix} 0.5 & -1 \\ 0.5 & 1 \\ -0.5 & 1 \\ -0.5 & -1 \end{bmatrix} \right) \quad 2 \times 4 \quad 4 \times 2$$

$$\tilde{v} = \frac{1}{3} \begin{pmatrix} 1 & 0 \\ 0 & 4 \end{pmatrix} v = \begin{pmatrix} \frac{1}{3} & 0 \\ 0 & \frac{4}{3} \end{pmatrix} v$$

ii

e-value / e-vector

$$\hat{C} v = \lambda v \rightarrow |\hat{C} - \lambda I| = 0$$

$$\begin{vmatrix} \frac{1}{3} - \lambda & 0 \\ 0 & \frac{4}{3} - \lambda \end{vmatrix} = 0 \rightarrow (\lambda - \frac{1}{3})(\lambda - \frac{4}{3}) = 0 \quad \left\{ \begin{array}{l} \lambda_1 = \frac{1}{3} \\ \lambda_2 = \frac{4}{3} \end{array} \right.$$

$$\hat{C} v = \lambda v \rightarrow \hat{C} v - \lambda v = 0 \rightarrow \underbrace{(\hat{C} - \lambda I)v}_{=0} = 0 \quad v \neq 0$$

$$|\hat{C} - \lambda I| = 0$$

$$\lambda_1 = \frac{1}{3} \rightarrow v_1 = ?$$

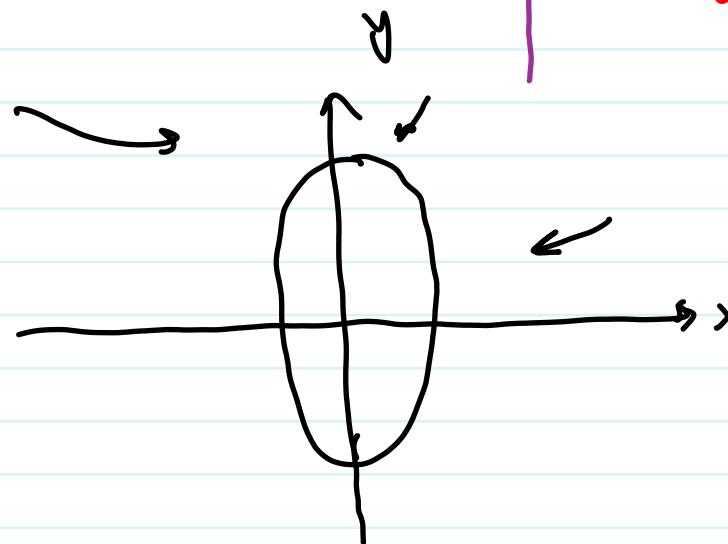
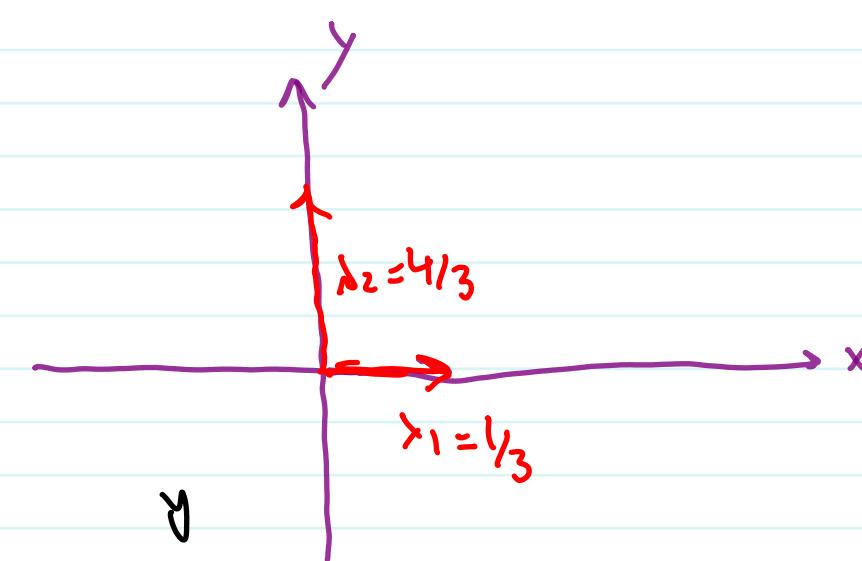
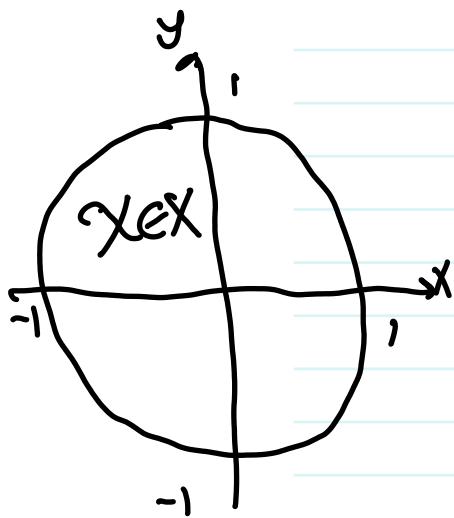
$$cv_1 = \lambda_1 v_1 \rightarrow \begin{bmatrix} \frac{1}{3} & 0 \\ 0 & \frac{4}{3} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \frac{1}{3}x \\ \frac{4}{3}y \end{bmatrix}$$

$$\begin{cases} \frac{1}{3}x = \frac{1}{3}x \\ \frac{4}{3}y = \frac{1}{3}y \end{cases} \quad v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\lambda_2 = \frac{4}{3} \rightarrow v_2 = ?$$

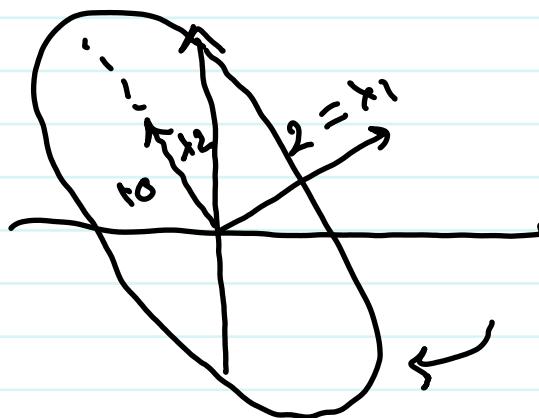
$$cv_2 = \lambda_2 v_2 \rightarrow \begin{bmatrix} \frac{1}{3} & 0 \\ 0 & \frac{4}{3} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \frac{4}{3}x \\ \frac{4}{3}y \end{bmatrix}$$

$$\begin{cases} \frac{1}{3}x = \frac{4}{3}x \\ \frac{4}{3}y = \frac{4}{3}y \end{cases} \quad v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$



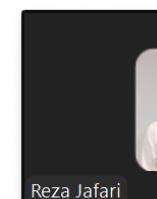
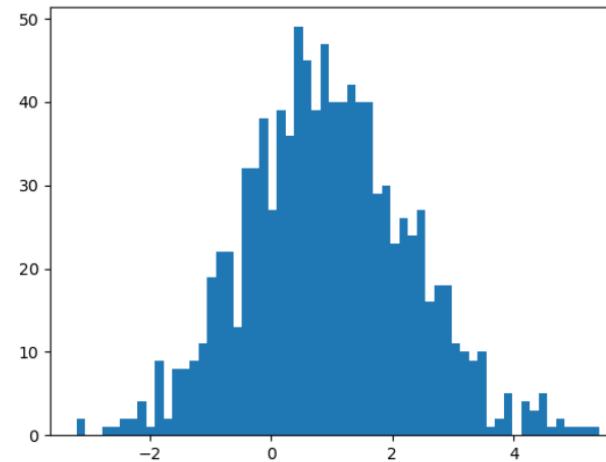
$$C' = \begin{bmatrix} \frac{4}{3} & b \\ b & \frac{1}{3} \end{bmatrix}$$

$$b \neq 0 < 0$$



In_Class_Sep27_pretty_table_example.py
In_Class_Sep27_VT_Data_Analytics.pdf
IQR-Standardization- Lecture-Boston.py
Jaccard Similarity.py
L1 and L2 Norm.pdf
LAB2.docx
LAB2.pdf
LAB_Sol.py
norm-all.py
Normalization -sklearn.py
OOP_normalization_standarziation.py
OOP_skew_call.py
Plotting exponential function.py
Rank-Example.py
relation-between-svd of X and e-values of covariance matrix.p
skewness_object01.py
skewness_object02.py
skewness_object-04.py
skewness_object_04.py

15 H =
16
17 d =
18 HH
19 for
20
21



```
nsole x Covariance-eigne-value-eigen-vector x
[ 0.57665071, -0.50822058, -0.6596945 ],
 [ 0.55822137, -0.32663401,  0.76269203])
```

```
[27]: import numpy as np
[28]: x = np.random.normal(mean, np.sqrt(var),
[29]: mean = 1
[30]: var = 2
[31]: ..
```

```
> C = {ndarray: (3, 3)} [[344.74470229 257.20500807
> H = {ndarray: (2, 2)} [[2.01785688 2.05222957], [2.0
> HH = {DataFrame: (2, 2)} [0, 1] [0 2.017857 2.05222957]
> N = {int} 1000
> X = {ndarray: (1000, 3)} [[8.15389049e-01 2.2557581
> X_tid = {DataFrame: (10000, 2)} [0, 1] [0 1.15059 ...]
> a = {ndarray: (3,)} [825.41207595 79.0593311 86.6
> b = {ndarray: (3, 3)} [[ 0.5965617  0.79688272 -0.09
> d = {PrettyTable} +-----+
> i = {int} 1
> mean = {int} 1
> s = {ndarray: (3,)} [28.72998566 9.30839748 8.891
> u = {ndarray: (1000, 1000)} [[-2.76158847e-02 -2.110
> v = {ndarray: (3, 3)} [[-0.5965617 -0.57663077 -0.55
> var = {int} 2
> vh = {ndarray: (2, 2)} [[ 0.44395552  0.89604882], [
> x = {ndarray: (1000,)} [-7.20016292e-01 -4.1920559,
```

likelihood pure! on
pric

$$\underbrace{P(Y | x_1, x_2, \dots, x_d)}_{\text{posterior}} = \frac{\overbrace{P(x_1, x_2, \dots, x_d | Y) P(Y)}^{\text{prior}}}{P(x_1, \dots, x_d)}$$

$$P(B|A) = \frac{P(A \cap B)}{P(A)} = \frac{\overbrace{P(A|B) \cdot P(B)}^{\text{Bayes Rule}}}{P(A)}$$

$$P(\text{cancer} | x_1, \dots, x_d) < P(\overline{\text{cancer}} | x_1, \dots, x_d)$$

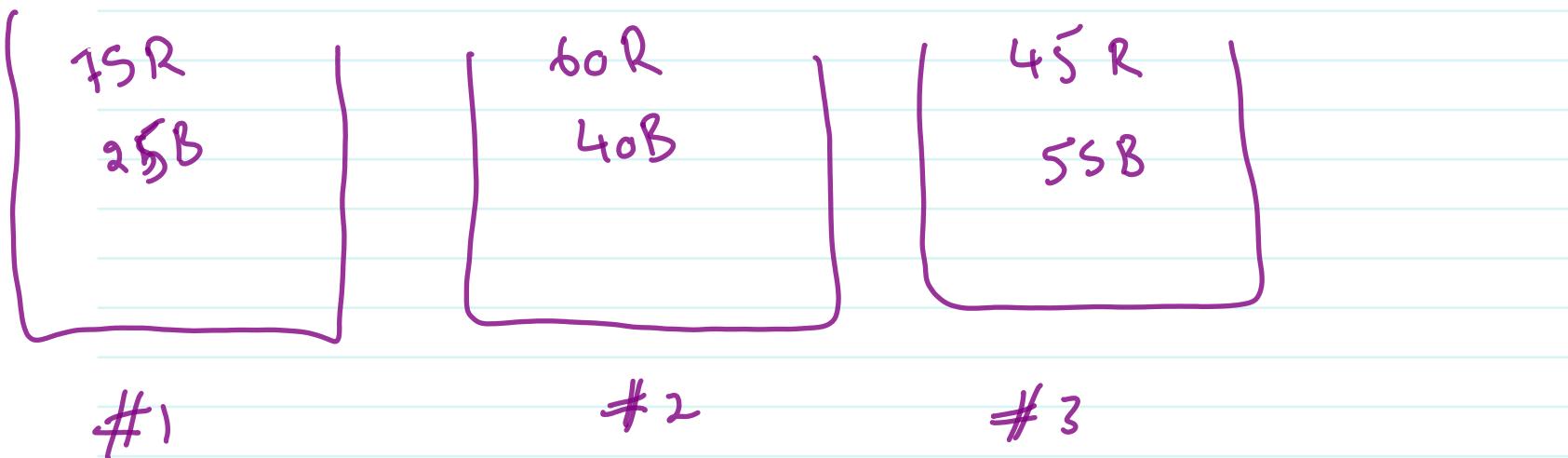
$$P(B|A) = \frac{P(A \cap B)}{P(A)} \xrightarrow[\text{A, B}]{\text{indep}} \frac{P(A) \cdot P(B)}{P(A)} = P(B)$$

$$f_{X|Y}(x_1 | \lambda_1) = f_X(x_1)$$

- **Example:** There are three bags that each contain 100 marbles:

- Bag 1 has 75 red and 25 blue marbles;
- Bag 2 has 60 red and 40 blue marbles;
- Bag 3 has 45 red and 55 blue marbles;

I choose one of the bags at random and then pick a marble from the chosen bag, also at random. What is the probability that the chosen marble is red?



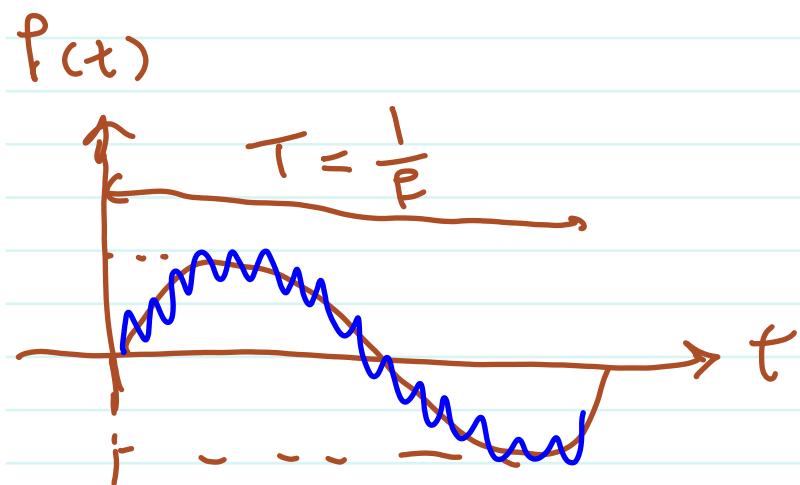
$$P(\text{red}) = \frac{1}{3} \left(\frac{75}{100} + \frac{60}{100} + \frac{45}{100} \right) = \frac{180}{100} \cdot \frac{1}{3} = 60\%$$

$$P(\text{Stock} \uparrow | \text{Lunched}) > P(\text{Stock} \uparrow | \overline{\text{Launch}})$$



$$f(t) = A \sin(\omega t + \theta)$$

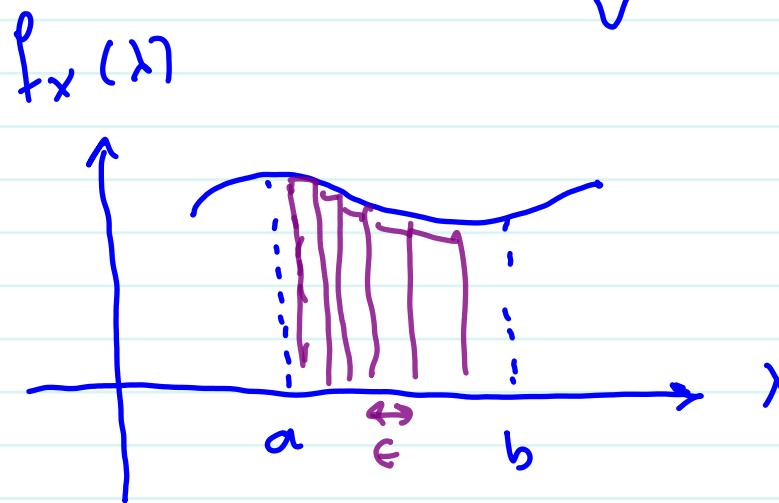
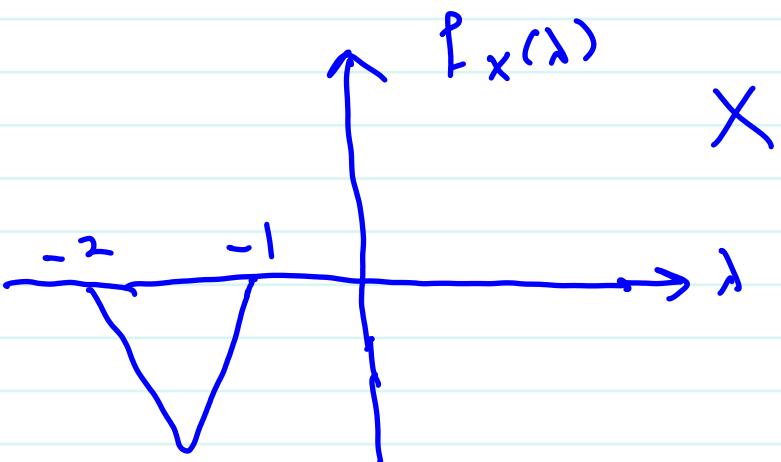
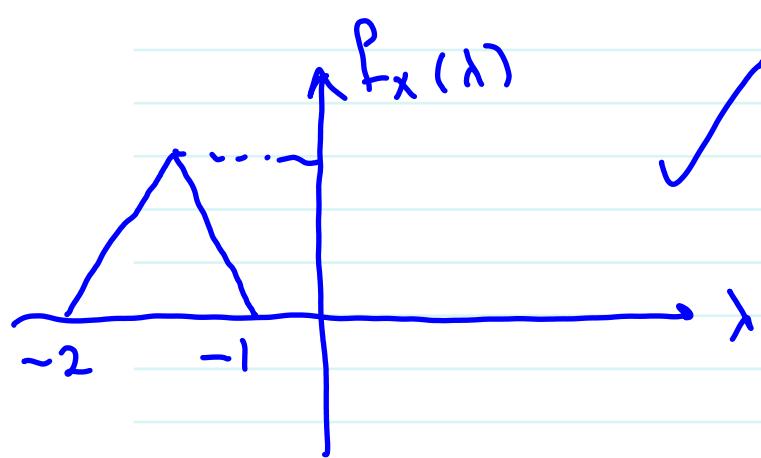
$$\left. \begin{array}{l} A \\ \omega \\ \theta = 0 \end{array} \right\} \text{fixed}$$



deterministic process

$$f'(t) = f(t) + \epsilon$$

$$\epsilon \sim \mathcal{WN}(\sigma, \sigma_e^2)$$

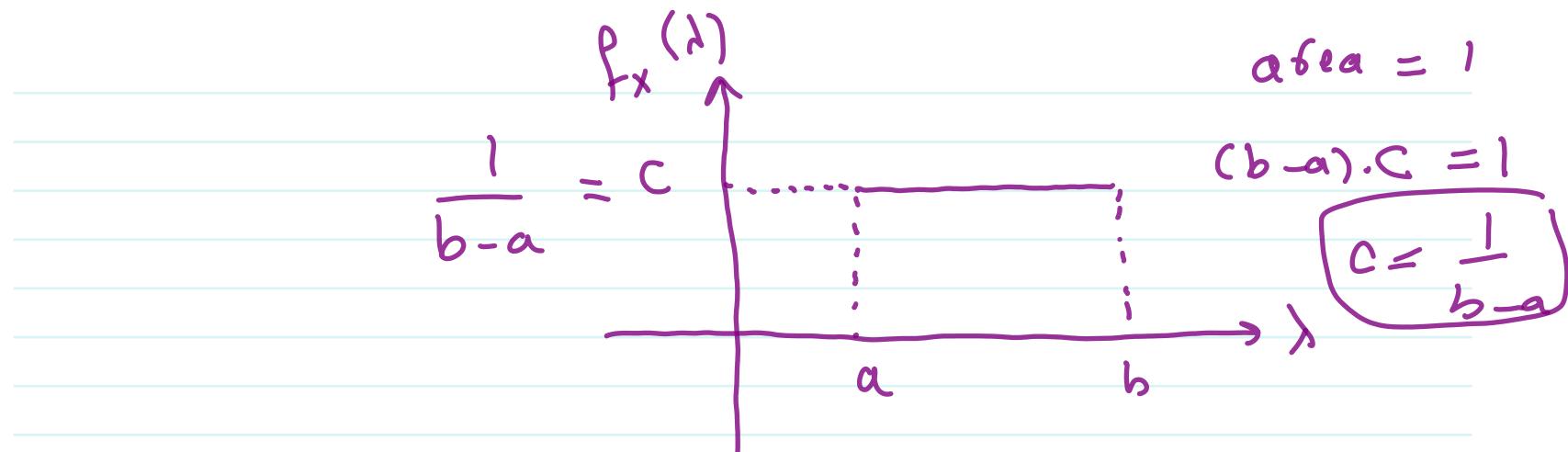


$$\Pr(a < X < b) = \int_a^b f_X(\lambda) d\lambda$$

Let consider a uniform density function for a random variable X. This random variable is said to have $Uniform(a,b)$ distribution. Or it can mathematically be written as:

$$f_X(\lambda) = \begin{cases} c, & 0 < a < \lambda < b \\ 0, & \text{Else} \end{cases}$$

Graph the above density function and find c in terms of a and b.



$$\int_{-\infty}^{\infty} f_X(\lambda) d\lambda = 1 \rightarrow \int_a^b c d\lambda = 1 \rightarrow c \lambda \Big|_a^b = 1$$

$$c(b-a) = 1 \rightarrow c = \frac{1}{b-a}$$

Let X be a continuous random variable with the following pdf:

$$e^{-\infty} = \frac{1}{e^{\infty}} = 0$$

$$f_X(\lambda) = \begin{cases} ce^{-\lambda}, & \lambda \geq 0 \\ 0, & \text{Else} \end{cases}$$

- ① Find c .
- ② Find the probability distribution function $F_X(\lambda)$
- ③ Find $P(1 < X \leq 3)$
- ④ Find $P(X = 2)$
- ⑤ Find $P(X \in [0, 1] \cup [3, 4])$

$$\int_{-\infty}^{\infty} f_X(\lambda) d\lambda = 1 \rightarrow \int_0^{\infty} ce^{-\lambda} d\lambda = 1$$

$$\left. \frac{ce^{-\lambda}}{-1} \right|_0^{\infty} = 1 \rightarrow -c(e^{-\infty} - e^0) = 1 \rightarrow -c(-1) = 1 \rightarrow c = 1$$

$$\int_a^b ae^{-cx} dx = \frac{ae^{-cx}}{-c}$$

$$\int a x^n dx = \frac{ax^{n+1}}{n+1}$$

$$\frac{d(ax^n)}{dx} = anx^{n-1}$$

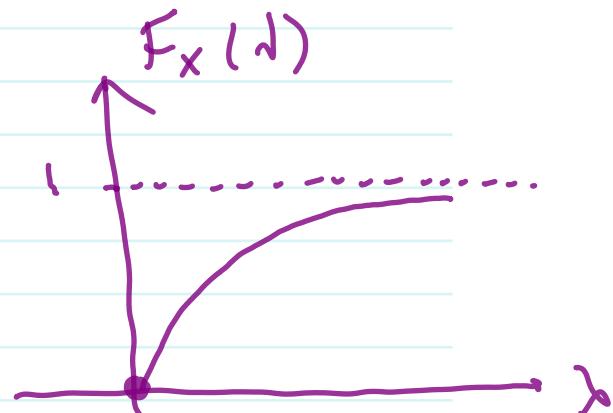
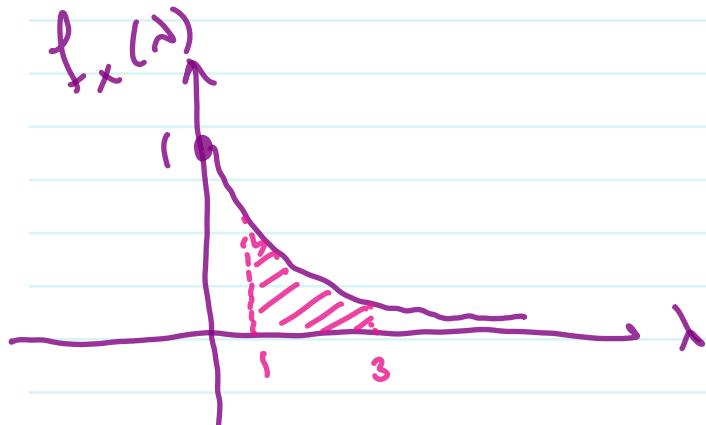
iii)

$$F_X(\lambda) = \int_{-\infty}^{\lambda} f_X(\mu) d\mu$$

$$= \int_0^{\lambda} e^{-\mu} d\mu = \frac{e^{-\mu}}{-1} \Big|_0^{\lambda} = -[e^{-\lambda} - e^0] = 1 - e^{-\lambda}$$

$$f_X(\lambda) = e^{-\lambda} : \lambda > 0$$

$$F_X(\lambda) = 1 - e^{-\lambda} : \lambda > 0$$



$$\Pr(X < \lambda < 3) = \int_1^3 e^{-\lambda} d\lambda$$

$$= \frac{-\lambda}{-1} \Big|_1^3 = -(\bar{e}^3 - \bar{e}) = \boxed{\frac{-1 - 3}{\bar{e} - \bar{e}}}$$

$$\Pr(x=2) = \lim_{\epsilon \rightarrow 0} \Pr(2-\epsilon < X < 2+\epsilon)$$

$$\leq \lim_{\epsilon \rightarrow 0} F_x(2+\epsilon) - \lim_{\epsilon \rightarrow 0} F_x(2-\epsilon) = 0$$

$$\Pr(x \in [0, 1] \cup [3, 4]) = \int_0^1 \bar{e}^{-\lambda} d\lambda + \int_3^4 \bar{e}^{-\lambda} d\lambda$$

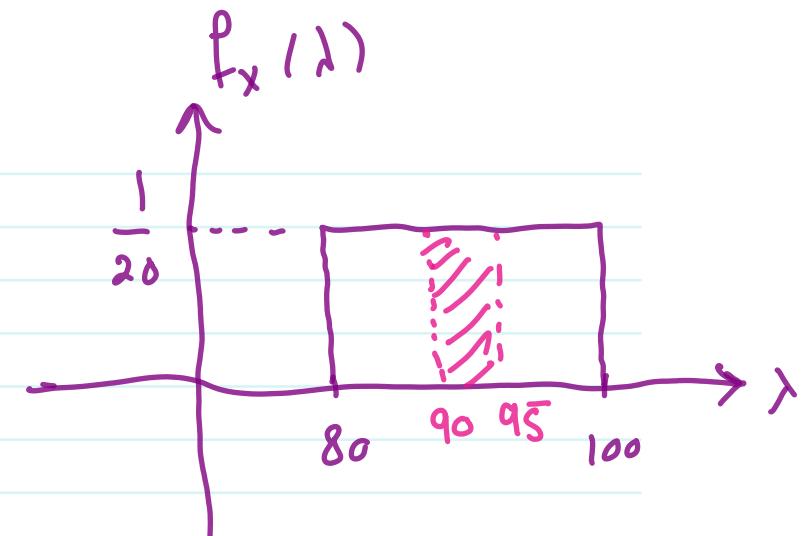
$$= \frac{\bar{e}^{-\lambda}}{-1} \Big|_0^1 + \frac{\bar{e}^{-\lambda}}{-1} \Big|_3^4 =$$

$$\begin{aligned}
 &= -\left(e^{-1} - 1\right) - \left(e^{-4} - e^{-3}\right) = \\
 &= \boxed{1 - e^{-1} - e^{-4} + e^{-3}}
 \end{aligned}$$

Let X be a continuous random variable that is equally likely to be any value between 80 and 100.

- ① Graph the corresponding probability density function.
- ② Find $P(90 < X \leq 95)$

$$f_X(x) = \begin{cases} \frac{1}{20} & : 80 < x < 100 \\ 0 & : \text{Else} \end{cases}$$



$$P(90 < X < 95) = 5 \cdot \frac{1}{20} = \boxed{25\%}$$

Let X and Y be two jointly continuous random variables (uniformly distributed) with joint probability density function:

$$f_{XY}(\lambda_1, \lambda_2) = \begin{cases} \lambda_1 + c\lambda_2^2, & 0 \leq \lambda_1 \leq 1, 0 \leq \lambda_2 \leq 1 \\ 0, & \text{Else} \end{cases}$$

① Find c .

λ_2 is changing
② Find $P(0 < X \leq \frac{1}{2}, 0 < Y \leq \frac{1}{2})$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(\lambda_1, \lambda_2) d\lambda_1 d\lambda_2 = 1 \rightarrow$$

$$\int_0^1 \int_0^1 (\lambda_1 + c\lambda_2^2) d\lambda_1 d\lambda_2 = 1$$

λ_1 is changing

λ_2 is fixed

$$\rightarrow \int_0^1 \left(\frac{\lambda_1^2}{2} + c\lambda_1\lambda_2^2 \right) \Big|_0^1 d\lambda_2 = 1$$

$$\int_0^1 \left(\frac{1}{2} + c\lambda_2^2 \right) d\lambda_2 = 1$$

$$\left. \left(\frac{\lambda_2}{2} + \frac{c\lambda_2^3}{3} \right) \right|_0^1 = 1 \rightarrow \frac{1}{2} + \frac{c}{3} = 1$$

$$\frac{c}{3} = \frac{1}{2} \rightarrow \boxed{c = \frac{3}{2}}$$

$$P(X \leq \frac{1}{2}, Y \leq \frac{1}{2}) = \int_0^{\frac{1}{2}} \int_0^{\frac{1}{2}} \left(\lambda_1 + \frac{3}{2}\lambda_2^2 \right) d\lambda_1 d\lambda_2$$

$$= \int_0^{\frac{1}{2}} \left(\frac{\lambda_1^2}{2} + \frac{3}{2}\lambda_1 \lambda_2^2 \right) \Big|_0^{\frac{1}{2}} d\lambda_2 = \int_0^{\frac{1}{2}} \left(\frac{1}{8} + \frac{3}{2} \cdot \frac{1}{2} \cdot \lambda_2^2 \right) d\lambda_2$$

$$= \left(\frac{\lambda_2}{8} + \frac{3}{4} \cdot \frac{\lambda_2^3}{8} \right) \Big|_0^{\frac{1}{2}} = \frac{1}{16} + \frac{1}{32} = \boxed{\frac{3}{32}}$$

Let X and Y be two jointly continuous random variables with the joint probability density function:

$$f_{XY}(\lambda_1, \lambda_2) = \begin{cases} c\lambda_1^2 + \frac{\lambda_1\lambda_2}{3}, & 0 \leq \lambda_1 \leq 1, 0 \leq \lambda_2 \leq 2 \\ 0, & \text{Else} \end{cases}$$

① Find c .

② Find $P(X + Y \geq 1)$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left(c\lambda_1^2 + \frac{\lambda_1\lambda_2}{3} \right) d\lambda_1 d\lambda_2 = 1$$

$$\int \int \phi$$

$$\int_0^1 \int_0^2 \left(c\lambda_1^2 + \frac{\lambda_1\lambda_2}{3} \right) d\lambda_2 d\lambda_1 = 1$$

$$\int_0^1 \left(c\lambda_2 \lambda_1^2 + \frac{\lambda_2^2 \lambda_1}{6} \right) \Big|_0^2 = 1$$

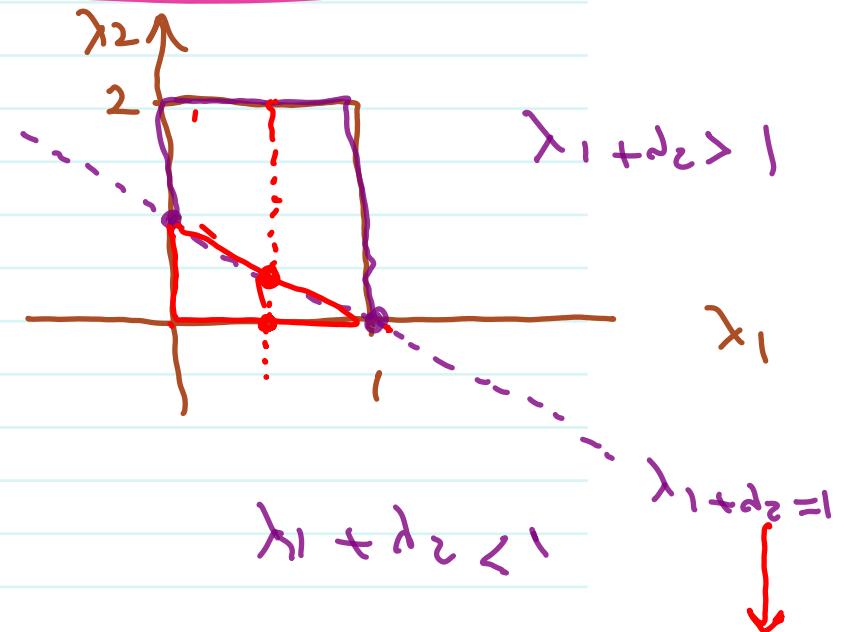
$$\int_0^1 \left(2c\lambda_1^2 + \frac{2}{3}\lambda_1 \right) d\lambda_1 = 1$$

$$\left(\frac{2c\lambda_1^3}{3} + \frac{2}{3} \cdot \frac{\lambda_1^2}{2} \right) \Big|_0^1 = 1$$

$$\frac{2c}{3} + \frac{1}{3} = 1 \rightarrow \frac{2c}{3} = \frac{2}{3} \rightarrow c = 1$$

$$\Pr(\underline{\lambda_1 + \lambda_2 > 1})$$

$$= 1 - \Pr(\lambda_1 + \lambda_2 < 1)$$



$$= 1 - \int_0^1 \int_0^{1-\lambda_1} (\lambda_1^2 + \frac{\lambda_1 \lambda_2}{3}) d\lambda_2 d\lambda_1$$

$$= 1 - \int_0^1 \left(\lambda_2 \lambda_1^2 + \frac{\lambda_1 \lambda_2^2}{6} \right) \Big|_0^{1-\lambda_1} d\lambda_1$$

$$\lambda_2 = 1 - \lambda_1$$

$$= 1 - \int_0^1 \left(\lambda_1^2 (1-\lambda_1) + \frac{\lambda_1 (1-\lambda_1)^2}{6} \right) d\lambda_1$$

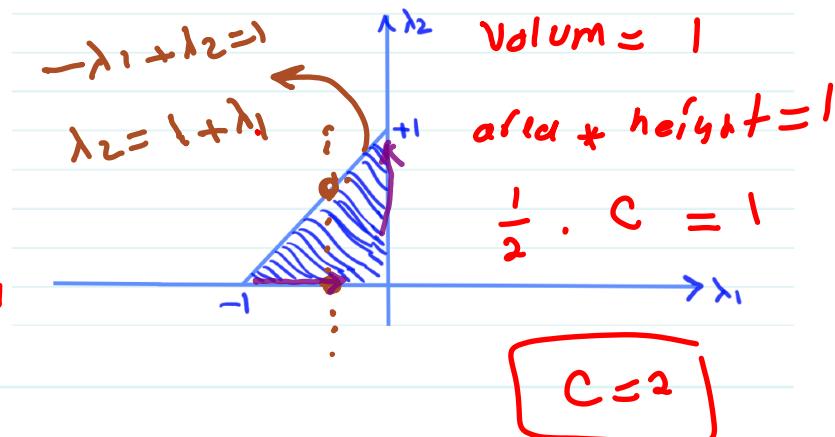
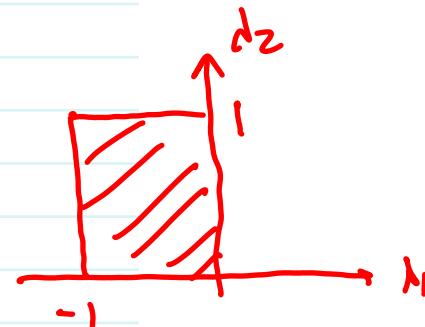
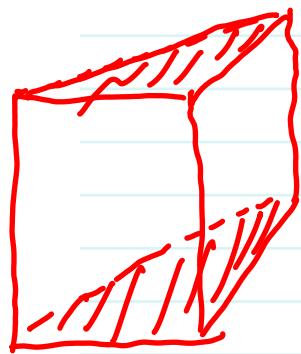
$$= 1 - \int_0^1 \left(\lambda_1^2 - \lambda_1^3 + \frac{\lambda_1}{6} - \frac{1}{3} \lambda_1^2 + \frac{\lambda_1^3}{6} \right) d\lambda_1$$

$$= 1 - \left(\frac{\lambda_1^3}{3} - \frac{\lambda_1^4}{4} + \frac{\lambda_1^2}{12} - \frac{\lambda_1^3}{9} + \frac{\lambda_1^4}{24} \right) \Big|_0^1 .$$

$$= 1 - \underbrace{\left(\frac{1}{3} - \frac{1}{4} + \frac{1}{12} - \frac{1}{9} + \frac{1}{24} \right)}_{\frac{1}{12}} = 1 - \left(\frac{1}{6} - \frac{1}{9} + \frac{1}{24} \right)$$
$$= \boxed{1 - \left(\frac{5}{24} - \frac{1}{9} \right) =}$$

- Let consider two random variables X and Y with the following joint probability density function:

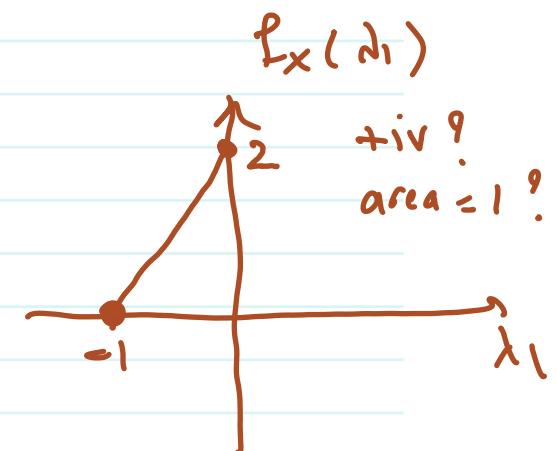
$$f_{XY}(\lambda_1, \lambda_2) = \begin{cases} C, & -1 \leq \lambda_1 \leq 0, 0 \leq \lambda_2 \leq 1 \\ 0, & \text{Else} \end{cases}$$



$$f_X(\lambda_1) = \int_{-\infty}^{\infty} f_{X,Y}(\lambda_1, \lambda_2) d\lambda_2$$

$$= \int_0^{1+\lambda_1} 2 d\lambda_2 = 2(1+\lambda_1)$$

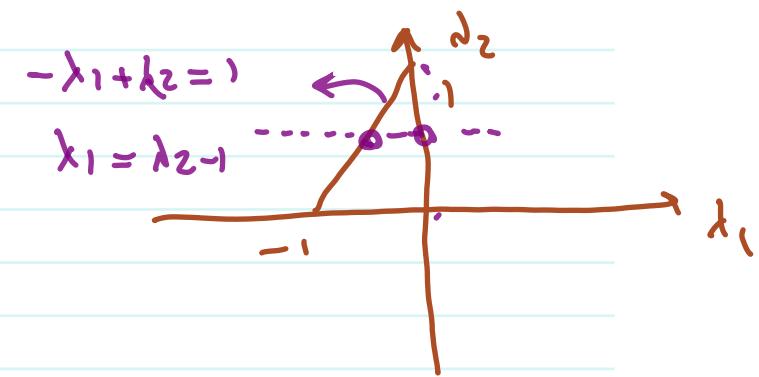
$$f_X(\lambda_1) = \begin{cases} 2(1+\lambda_1) : & -1 < \lambda_1 < 0 \\ 0 & : \text{Else} \end{cases}$$



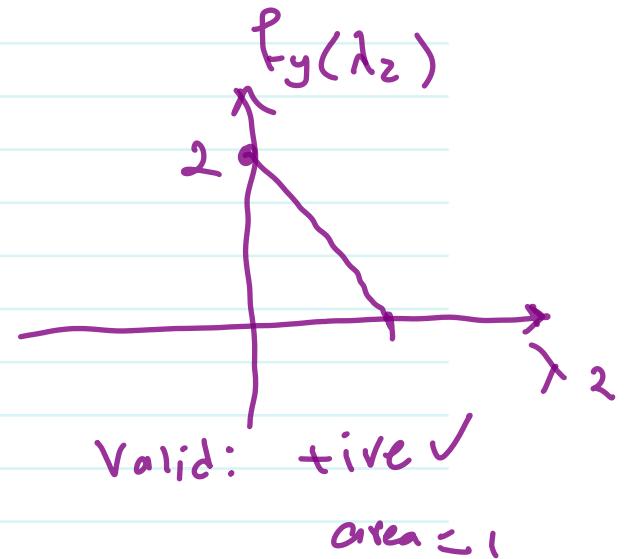
$$f_y(\lambda_2) = \int_{-\infty}^{\infty} f_{xy}(\lambda_1, \lambda_2) d\lambda_1$$

$$= \int_{\lambda_2-1}^{0} 2 d\lambda_1$$

$$= 2(1-\lambda_2)$$



$$f_y(\lambda_2) = \begin{cases} 2(1-\lambda_2) & : 0 < \lambda_2 < 1 \\ 0 & : \text{Else} \end{cases}$$

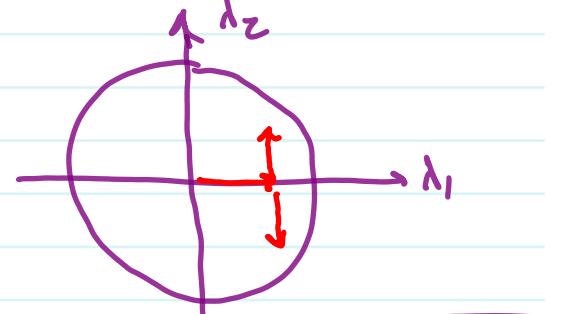


independency :

$$f_{xy}(\lambda_1, \lambda_2) \stackrel{!}{=} f_x(\lambda_1) \cdot f_y(\lambda_2) \quad \forall \lambda_1, \lambda_2$$

$$2 \neq 2(1+\lambda_1) \cdot 2(1-\lambda_2)$$

x, y are not independent = dependent.



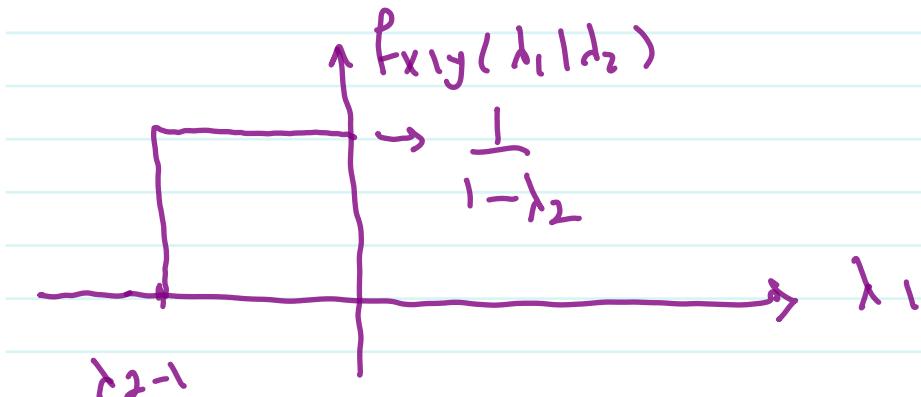
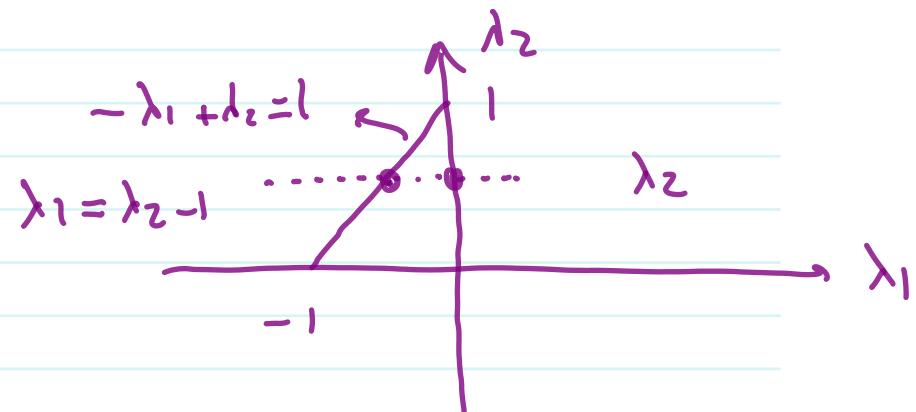
$$f_x(\lambda_1) = \begin{cases} 2(1+\lambda_1) : -1 < \lambda_1 < 0 \\ 0 : \text{else} \end{cases}$$

$$f_y(\lambda_2) = \begin{cases} 2(1-\lambda_2) : 0 < \lambda_2 < 1 \\ 0 : \text{else} \end{cases}$$

$$f_{x,y}(\lambda_1, \lambda_2) = \frac{f_{x,y}(\lambda_1, \lambda_2)}{f_y(\lambda_2)} = \frac{2}{2(1-\lambda_2)} = \frac{1}{1-\lambda_2}$$

$$f_{x,y}(\lambda_1 | \lambda_2 = \frac{1}{4}) = \frac{1}{1-\frac{1}{4}} = \frac{4}{3}$$

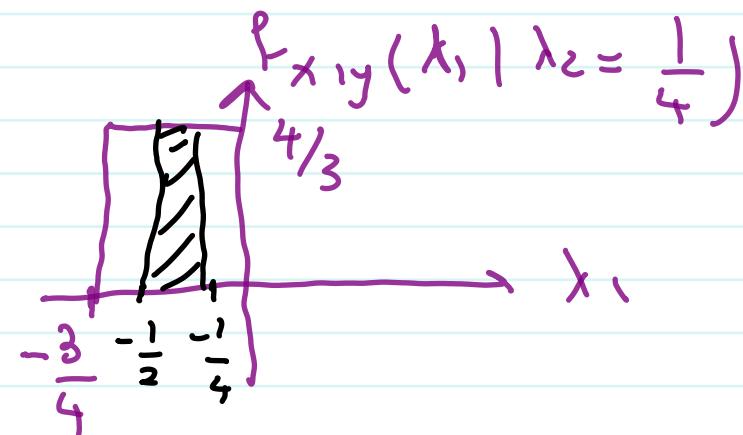
$$f_{X_1 Y}(\lambda_1 | \lambda_2) = \frac{1}{1-\lambda_2}$$



Valid = five

area = 1

$$f_{X_1 Y}(\lambda_1 | \lambda_2 = \frac{1}{4}) = \frac{4}{3}$$

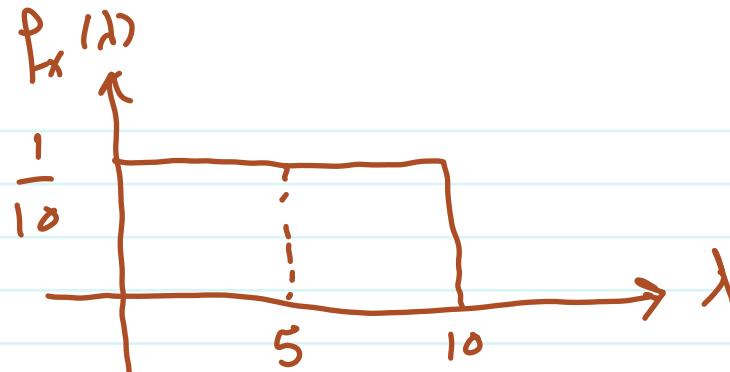


$$\Pr\left(-\frac{1}{2} < \lambda_1 < -\frac{1}{4} \mid \lambda_2 = \frac{1}{4}\right) = \frac{4}{3} * \frac{1}{4} = \frac{1}{3}$$

λ_1, λ_2

$$\alpha_{xy} = E[(x - \mu_x)(y - \mu_y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x - \mu_x)(y - \mu_y) f_{xy}(x, y) dx dy$$

- Let consider a continuous random variable X to be defined uniformly between 0 and 10. What is the mean of X and what is the variance?
- Answer :** $\mu_x = 5$ and $\sigma_x^2 = 8.3$



$$f_x(\lambda) = \begin{cases} \frac{1}{10} & : 0 < \lambda < 10 \\ 0 & : \text{else} \end{cases}$$

$$\begin{aligned} E[x] &= \int_{-\infty}^{\infty} x f_x(x) dx = \int_0^{10} x \frac{1}{10} dx = \frac{x^2}{20} \Big|_0^{10} \\ &= \frac{100}{20} = 5 \end{aligned}$$

$$\sigma_x^2 = E[(x-5)^2] = \int_0^{10} \frac{(\lambda-5)^2}{10} d\lambda$$

$$= \frac{1}{10} \int_0^{10} (\lambda^2 - 10\lambda + 25) d\lambda = \frac{1}{10} \left(\frac{\lambda^3}{3} - 5\lambda^2 + 25\lambda \right) \Big|_0^{10}$$

$$= \frac{1}{10} \left(\frac{1000}{3} - \overbrace{500+250}^{-250} \right) = \frac{100}{3} - 25 = \frac{25}{3} \approx 8.3$$

