



Computer
Science

CSC380: Principles of Data Science

Probability 4

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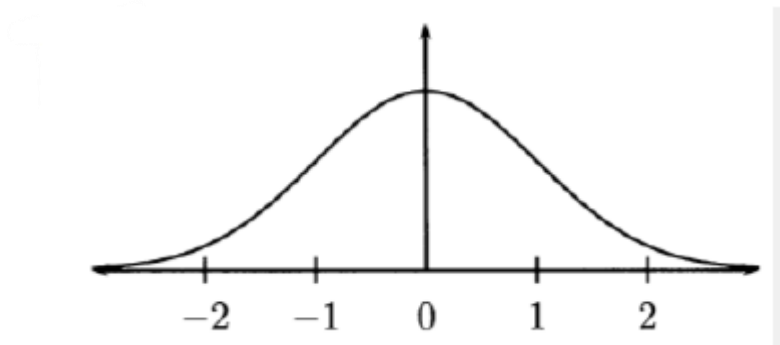
Recap

- PDF of a transformation of a continuous RV
 - $X + b$ has a PDF that is a translation of X 's PDF by b units
 - aX 's PDF is X 's PDF stretched by a factor of a horizontally
- Mean
 - $E[X] = \int x f(x) dx$
 - $E[r(X)] = \int r(x)f(x) dx$
- Variance
 - $\text{Var}(X) = \sigma^2 = E[(X - \mu)^2] = E[X^2] - (E[X])^2$
- Properties
 - $E[aX] = a E[X]$
 - $\text{Var}(aX) = a^2 \text{Var}(X)$
 - $E[X + b] = E[X] + b$
 - $\text{Var}(X + b) = \text{Var}(X)$

- Calculating probabilities about Gaussians
- Multivariate Random Variables
 - Joint distribution
 - Marginal distribution

The standard Gaussian distribution

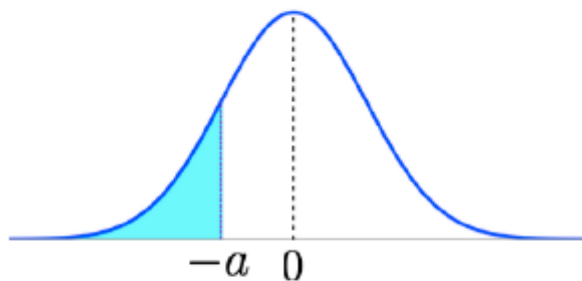
- Gaussian distribution with $\mu = 0$ and $\sigma^2 = 1$



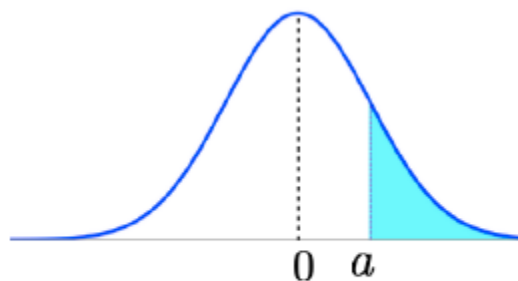
- Denoted by $Z \sim N(0,1)$
- Its PDF denoted by $\phi(z)$, and CDF denoted by $\Phi(z)$

Calculating probabilities about Gaussians

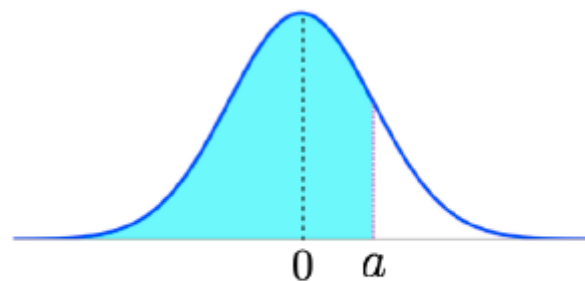
- Symmetry of $\phi \Rightarrow \Phi(-a) = 1 - \Phi(a)$



$$\Phi(-a) = P(Z \leq -a)$$



$$= P(Z \geq a)$$



$$= 1 - P(Z \leq a) = 1 - \Phi(a)$$

Calculating probabilities about Gaussians

- Suppose $X \sim N(5, 2^2)$, how can I calculate $P(1 < X < 8)$?
- Transform X into another variable:
 - $X \sim N(\mu, \sigma^2): E[X] = \mu, Var[X] = \sigma^2$
- What is mean and variance for the following transformations of X ?

$$\Rightarrow X - \mu$$

$$\Rightarrow \frac{X - \mu}{\sigma}$$

$$\begin{aligned} E[aX] &= a E[X] \\ Var(aX) &= a^2 Var(X) \\ E[X + b] &= E[X] + b \\ Var(X + b) &= Var(X) \end{aligned}$$

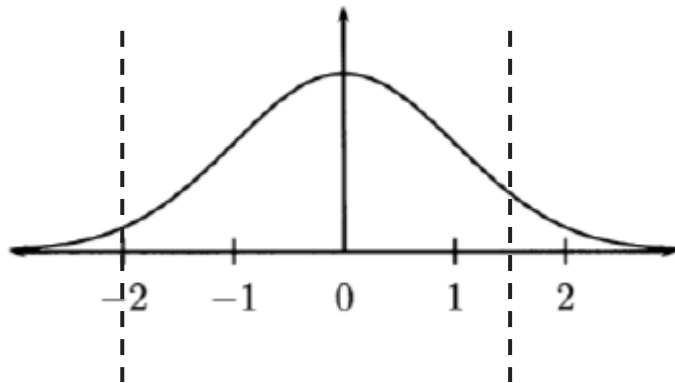
Calculating probabilities about Gaussians

- Suppose $X \sim N(5, 2^2)$, how can I calculate $P(1 < X < 8)$?
- Transform X into standard normal Z :
 - $X \sim N(\mu, \sigma^2)$
 - $\Rightarrow X - \mu \sim N(0, \sigma^2)$
 - $\Rightarrow Z = \frac{X - \mu}{\sigma} \sim N(0, 1)$
- We can write $P(a < X < b)$ using $P(c < Z < d)$, which in turn can be written in Φ .

$$\begin{aligned}E[aX] &= a E[X] \\ \text{Var}(aX) &= a^2 \text{Var}(X) \\ E[X + b] &= E[X] + b \\ \text{Var}(X + b) &= \text{Var}(X)\end{aligned}$$

Calculating probabilities about Gaussians

$$\begin{aligned} & \bullet P(a < X < b) \\ &= P\left(\frac{a - \mu}{\sigma} < \frac{X - \mu}{\sigma} < \frac{b - \mu}{\sigma}\right) \\ &= P\left(\frac{a - \mu}{\sigma} < Z < \frac{b - \mu}{\sigma}\right) \\ &= \Phi\left(\frac{b - \mu}{\sigma}\right) - \Phi\left(\frac{a - \mu}{\sigma}\right) \end{aligned}$$



Example Suppose $X \sim N(5, 2^2)$, calculate $P(1 < X < 8)$

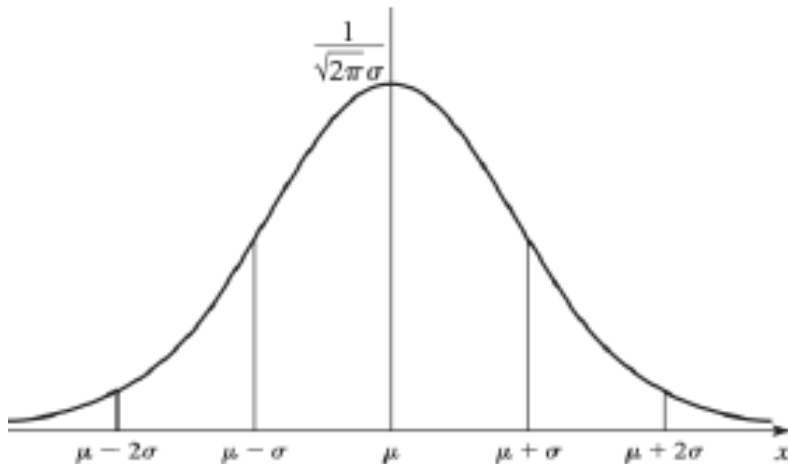
This is $\Phi\left(\frac{8-5}{2}\right) - \Phi\left(\frac{1-5}{2}\right) = \Phi(1.5) - \Phi(-2) = \Phi(1.5) - (1 - \Phi(2))$

```
from scipy.stats import norm
print(norm.cdf(1.5)-(1-norm.cdf(2)))
```

0.9104426667829627

Calculating probabilities about Gaussians

- What is the probability that a Gaussian RV X is within k ($k = 1, 2, \dots$) std of its mean?

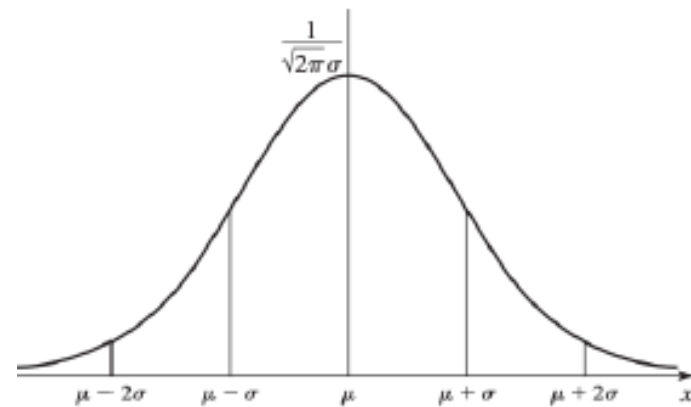


- $P(\mu - k\sigma \leq X < \mu + k\sigma)$

Calculating probabilities about Gaussians

- $p_k = P(\mu - k\sigma \leq X < \mu + k\sigma)$
 $= P\left(-k < \frac{X-\mu}{\sigma} < k\right)$
 $= P(-k < Z < k)$
 $= \Phi(k) - (1 - (\Phi(k)))$
 $= 2\Phi(k) - 1$

k	p_k
1	0.6826
2	0.9544
3	0.9974
4	0.99994



In words,

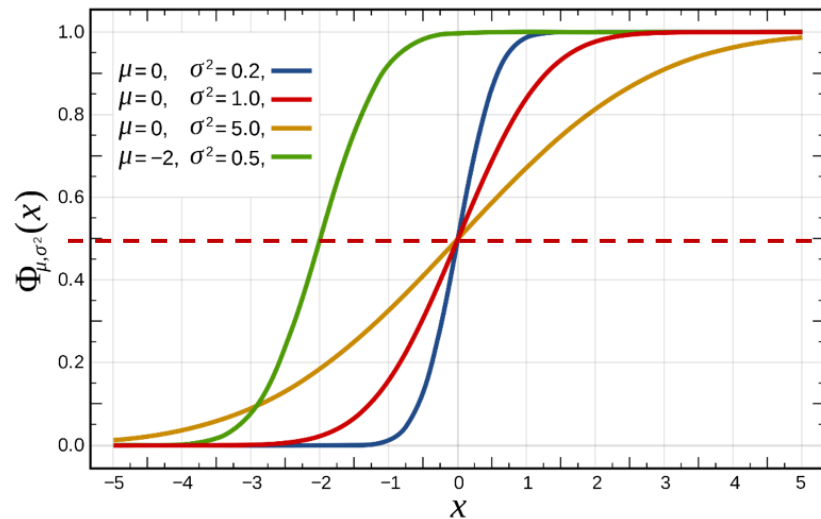
- With probability about 95%, X is within 2 std of its mean
- With overwhelming prob. (99.7%), X within 3 std of mean

CDF of Gaussian Distributions

- F : CDF of Gaussian $N(\mu, \sigma^2)$

Q: what is $F(\mu)$?

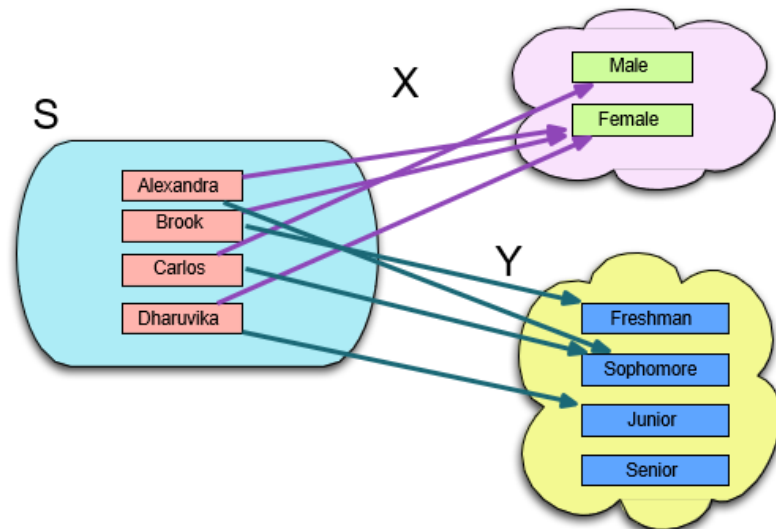
- $F(\mu) = \frac{1}{2}$



- $F(x)$ changes fast near μ
- F 's “sensitive range” is about $[\mu - 3\sigma, \mu + 3\sigma]$

Multivariate Random Variables

Multivariate RVs: example



- X : people \rightarrow their genders
- Y : people \rightarrow their class year
- We'd like to answer questions such as: does X and Y have a correlation?
 - I.e., is a student in higher class year more likely to be male?
- We call (X, Y) a multivariate RV, and will study its *joint* distribution

Joint distribution of discrete RVs

- The joint PMF (probability mass function) of discrete random variables X, Y :

$$f(x, y) = P(X = x, Y = y)$$

Examples

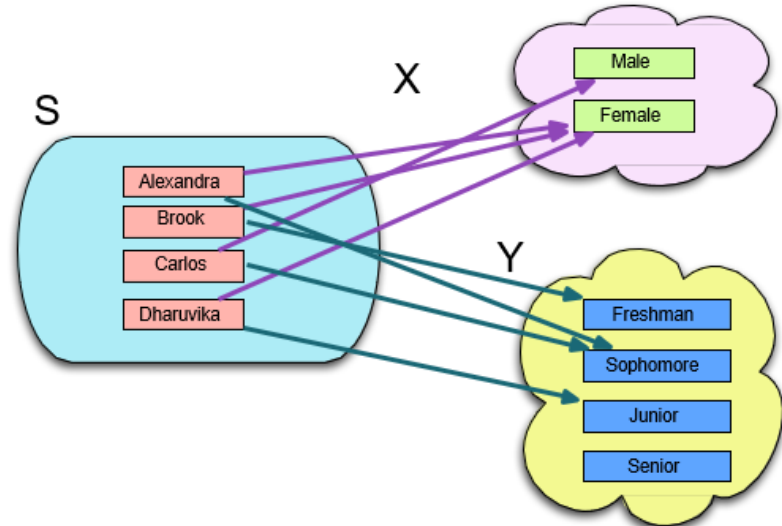
Alexandra

$$P(X = \text{Fem}, Y = \text{Soph}) = \frac{1}{4}$$

Dharuvika

$$P(X = \text{Fem}, Y = \text{Jun}) = \frac{1}{4}$$

...



Joint distribution of discrete RVs

- X : # of cars owned by a randomly selected household
- Y : # of computers owned by the same household

- Joint pmf shown with a table

x	y			
	1	2	3	4
1	0.1	0	0.1	0
2	0.3	0	0.1	0.2
3	0	0.2	0	0

- Probability that a randomly selected household has ≥ 2 cars and ≥ 2 computers?
 - $P(X \geq 2, Y \geq 2) = 0.5$

Marginal distributions

Given joint distribution of (X, Y) , need distribution of one of them, say X .
Named the ***marginal distribution*** of X .

- How to find $P(X = x)$?
- Using law of total probability:

$$f_1(x) = \sum_y f(x, y)$$

x	y			
	1	2	3	4
1	0.1	0	0.1	0
2	0.3	0	0.1	0.2
3	0	0.2	0	0

- This operation is called *marginalization* ('marginalizing out variable Y ', or variable elimination)

Marginal distributions

x	y				Total
	1	2	3	4	
1	0.1	0	0.1	0	0.2
2	0.3	0	0.1	0.2	0.6
3	0	0.2	0	0	0.2
Total	0.4	0.2	0.2	0.2	1.0

f_1 : marginal distribution of X

f_2 : marginal distribution of Y

$$f_1(X = 1) = \sum_y f(1, y) = 0.1 + 0 + 0.1 + 0 = 0.2$$

Joint distribution of continuous RVs

- Any continuous random vector (X,Y) has a *joint probability density function* (PDF) $f(x,y)$, such that for all C ,

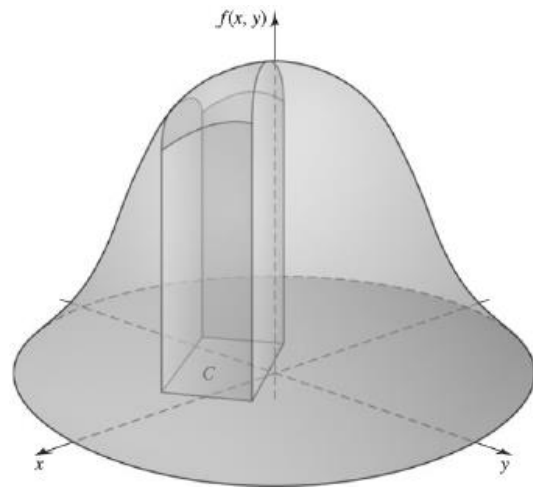
$$P((X,Y) \in C) = \iint_C f(x,y) dx dy$$

$f(x,y)$: represent a 2D surface

double integral: the *volume* under the surface

Properties:

- f is nonnegative
- $\iint_{R^2} f(x,y) dx dy = 1$ (R^2 = the whole x-y plane)
 - $P((X,Y) \in R^2) = 1$



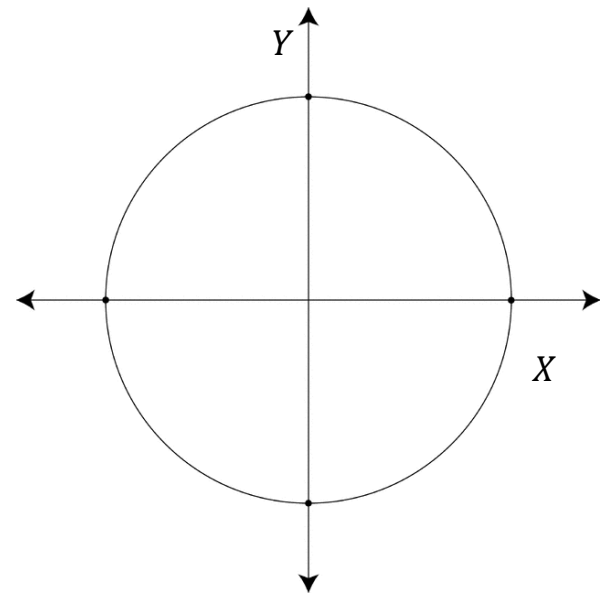
Example: dartboard

- Dartboard with center $(0,0)$ and radius 1; dart lands uniformly at random on the board

- What is the joint PDF of (X, Y) ?

- Fact: the PDF is

$$f(x, y) = \begin{cases} c, & x^2 + y^2 \leq 1 \\ 0, & \text{otherwise} \end{cases}$$



- This is called “the Uniform distribution over the unit disk”

Example: dartboard

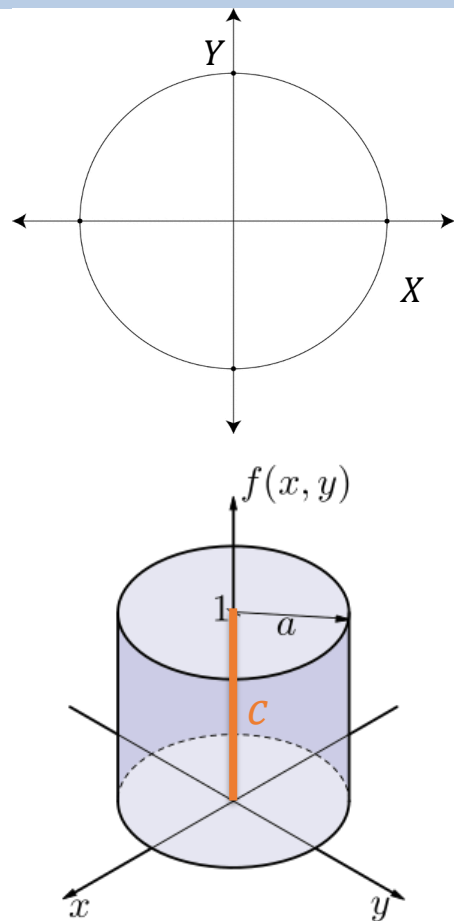
The PDF of X, Y is

$$f(x, y) = \begin{cases} c, & x^2 + y^2 \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

Can we find c ?

Observe: volume under $f(x, y)$ is πc (cylinder)
which must also be 1

Therefore, $c = 1/\pi$



Marginal distribution of continuous RV

Given joint distribution of continuous RV (X, Y) , how to find X 's PDF f_1 ?

Fact (marginalization) $f_1(x) = \int_R f(x, y) dy$

Replacing summation with integration in the continuous case ('marginalizing / integrating out variable Y ')

How about Y 's PDF f_2 ?

- Marginalize out X

Example: dartboard

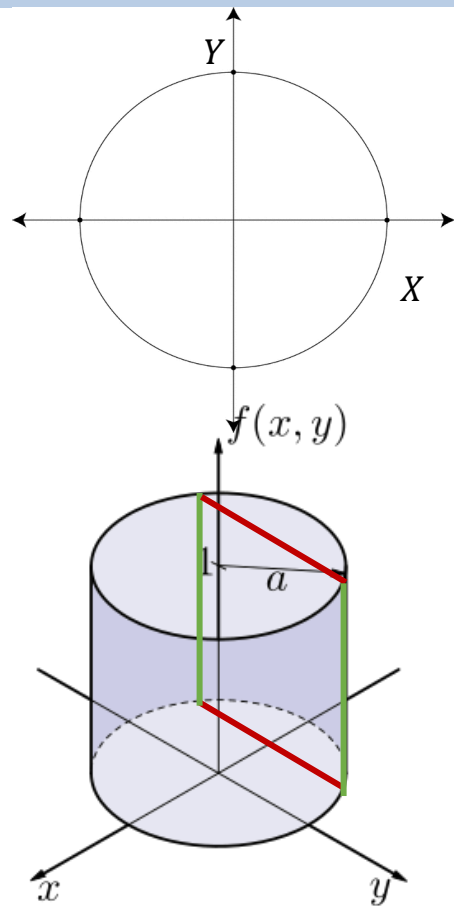
The PDF of X, Y is

$$f(x, y) = \begin{cases} \frac{1}{\pi}, & x^2 + y^2 \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

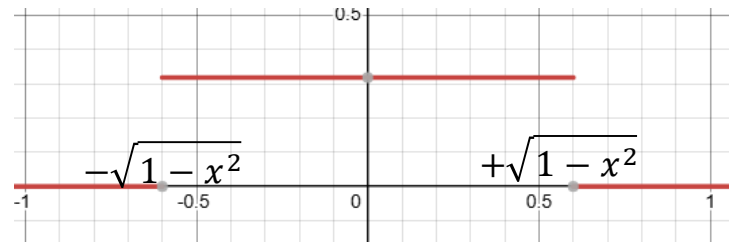
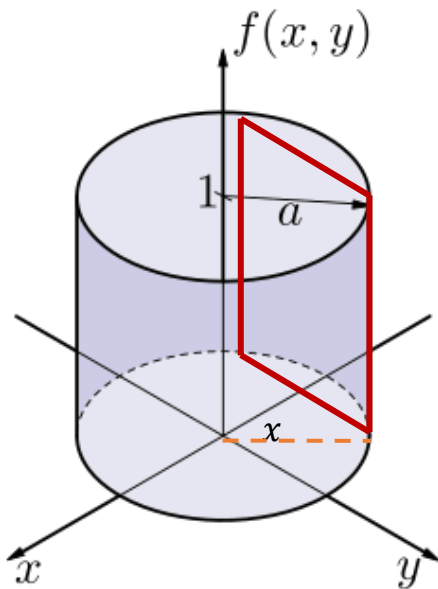
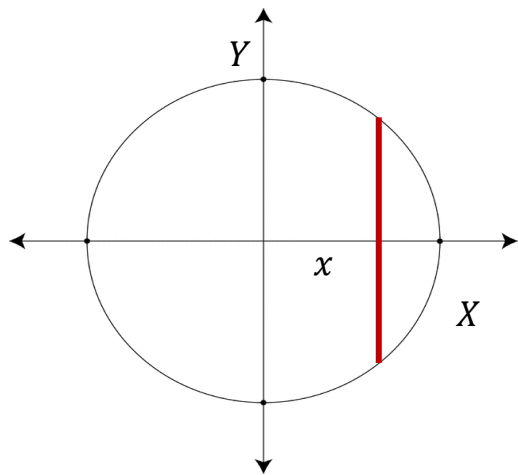
What is the marginal distribution over X ?

$$f_1(x) = \int_{-\infty}^{+\infty} f(x, y) dy$$

How to find this integral?



Example: dartboard



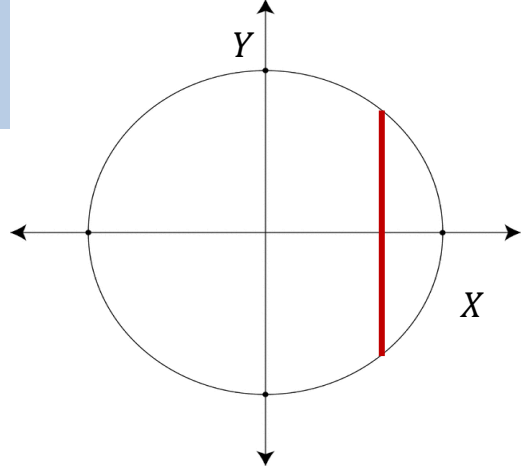
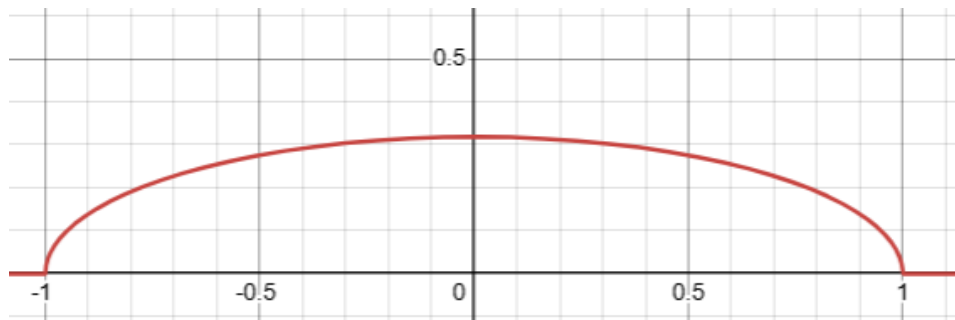
For a fixed $x \in [-1, 1]$, we can think of $f(x)$ is the area of the slice:

- height: $\frac{1}{\pi}$, width: $2 \cdot \sqrt{1 - x^2}$
- $f_1(x) = \frac{2}{\pi} \cdot \sqrt{1 - x^2}$

Example: dartboard

- In summary,

$$f(x) = \begin{cases} \frac{2}{\pi} \cdot \sqrt{1 - x^2}, & x \in [-1, 1] \\ 0, & \text{otherwise} \end{cases}$$



X 's distribution is NOT Uniform($[-1, 1]$)!

Actually makes sense: X closer to 1 is harder to be hit

Joint distribution of more than 3 RVs

- We can consider the joint distribution of more than 3 random variables,
 - E.g. (A,B,C), A = gender, B = class year, C = blood type
- Discrete RVs: can still define joint PMFs

a	b	c	$P(A = a, B = b, C = c)$
0	0	0	0.06
0	0	1	0.09
0	1	0	0.08
0	1	1	0.12
1	0	0	0.06
1	0	1	0.24
1	1	0	0.10
1	1	1	0.25

Marginalization

a	b	c	$P(A = a, B = b, C = c)$
0	0	0	0.06
0	0	1	0.09
0	1	0	0.08
0	1	1	0.12
1	0	0	0.06
1	0	1	0.24
1	1	0	0.10
1	1	1	0.25

Given the joint distribution of (A, B, C)

- What is the distribution of A ?

- Need to find $P(A = 0)$ and $P(A = 1)$

$$P(A = 0) = \sum_{b,c} P(A = 0, B = b, C = c)$$

Marginalization: summing over irrelevant variables

- What is the joint distribution of (A, B) ?

- Need to find $P(A = 0, B = 0), \dots, P(A = 1, B = 1)$

$$P(A = 0, B = 0) = \sum_c P(A = 0, B = 0, C = c)$$

Marginalization for continuous RVs

Suppose joint PDF of (A, B, C) is $f(a, b, c)$

- What is the PDF of A ?

$$f_A(a) = \iint_{\mathbb{R}^2} f(a, b, c) \, db \, dc$$

- What is the joint PDF of (A, B) ?

$$f_{A,B}(a, b) = \int_{\mathbb{R}} f(a, b, c) \, dc$$

Marginalization: summing over irrelevant variables

- These operations generalize to joint PDFs of more RVs..