



THE UNIVERSITY OF TEXAS AT AUSTIN

Department of Statistics and Data Sciences

College of Natural Sciences

SDS 321: Introduction to Probability and Statistics

Probability and Statistics-review

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Things you should know

▶ **Single r.v.**

- ▶ How to get a marginal pdf from a joint.
- ▶ How to calculate expectation, variance of a r.v.
- ▶ Derived distributions. Pay special attention to the Uniform. Its always tricky,

▶ **Multiple r.v.'s**

- ▶ When are two r.v.'s independent?
 - ▶ How to calculate covariance, correlation of two r.v.'s
 - ▶ Conditional expectation, iterated expectation rule.
 - ▶ Linearity of expectation. Variance of sum of r.v.'s
 - ▶ Covariance of two sums of r.v.'s
- ▶ Markov's inequality, Chebyshev's inequality. Central limit theorem.

Functions of a continuous random variable: Summary

- ▶ We know that functions $Y = g(X)$ of random variables X are again random variables.
- ▶ In the discrete case, it was pretty easy to find the PMF of the new random variable X :

$$p_Y(y) = \sum_{x|g(x)=y} p_X(x)$$

- ▶ In the continuous case, we need to find

$$F_Y(y) = \mathbf{P}(Y \leq y) = \int_{x|g(x) \leq y} f_X(x) dx$$

- ▶ We can then differentiate wrt y to get the PDF of Y .
- ▶ In certain cases, this procedure simplifies a little:
- ▶ **Linear case:** If $Y = aX + b$, $f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right)$
- ▶ **Monotonic case:** If g is a strictly monotonic function with inverse h , then $f_Y(y) = f_X(h(y)) \left| \frac{dh}{dy}(y) \right|$
- ▶ What about if we have functions of more than one random variable?

Functions of continuous random variables: Example

- ▶ What is the PDF of the area of a circle, if the radius is a uniform random variable on $[0,1]$?
- ▶ If X is the radius, we know that the area $Y = \pi X^2$.
- ▶ We know that

$$f_X(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

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$$f_X(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- ▶ We can calculate the CDF of Y :

$$\begin{aligned} F_Y(y) &= \mathbf{P}(\pi X^2 \leq y) \\ &= \int_{x|\pi x^2 \leq y} f_X(x) dx \end{aligned}$$

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$$\begin{aligned} F_Y(y) &= \mathbf{P}(\pi X^2 \leq y) \\ &= \int_{x|\pi x^2 \leq y} f_X(x) dx \end{aligned}$$

- ▶ $f_X(x)$ is only positive for $0 \leq x \leq 1$, corresponding to $0 \leq y \leq \pi$. So,

$$F_Y(y) = \begin{cases} 0 & y < 0 \\ \int_0^{\sqrt{y/\pi}} f_X(x) dx = \sqrt{\frac{y}{\pi}} & 0 \leq y \leq \pi \\ 1 & y > \pi \end{cases}$$

Functions of continuous random variables: Example

- From the previous slide:

$$F_Y(y) = \begin{cases} 0 & y < 0 \\ \int_0^{\sqrt{y/\pi}} f_X(x) dx = \sqrt{\frac{y}{\pi}} & 0 \leq y \leq \pi \\ 1 & y > \pi \end{cases}$$

- We can differentiate to get

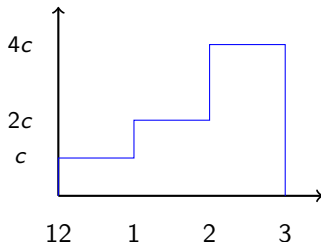
$$f_Y(y) = \begin{cases} 0 & y < 0 \\ \frac{d}{dy} \left(\sqrt{\frac{y}{\pi}} \right) = \frac{1}{2} \sqrt{\frac{1}{y\pi}} & 0 \leq y \leq \pi \\ 0 & y > \pi \end{cases}$$

Conditional expectation

- ▶ I am expecting an email, that will definitely arrive between midday and 3pm.
- ▶ Within a given hour (midday-1, 1-2, 2-3), each time is equally likely.
- ▶ It is twice as likely to arrive between 1 and 2 as it is to arrive between midday and 1.
- ▶ It is twice as likely to arrive between 2 and 3 as it is to arrive between 1 and 2.
- ▶ What does the PDF look like?

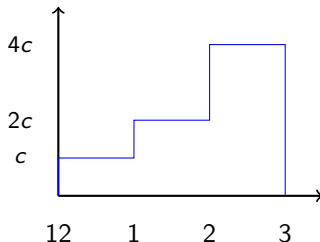
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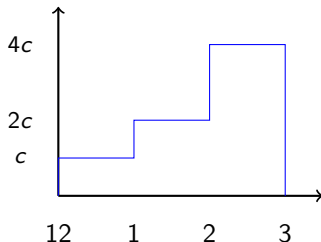
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- ▶ What is c ?

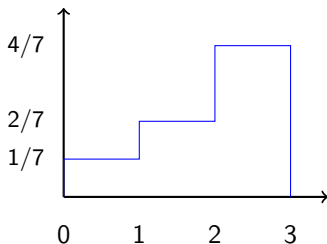
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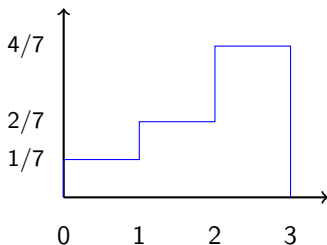
- ▶ What is c ? $1/7$

Conditional expectation



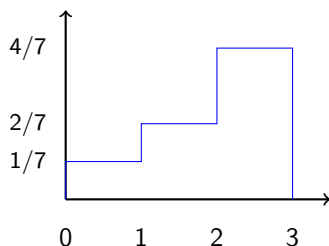
- ▶ I wait until 2pm. It still hasn't arrived. What is the expected value of the arrival time?
- ▶ What is the expected time without any conditioning?

Conditional expectation



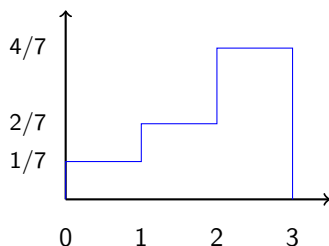
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Conditional expectation



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- ▶ What is the expected time without any conditioning?
- ▶ First, what is the conditional probability, $f_{X|X>2}(x)$?

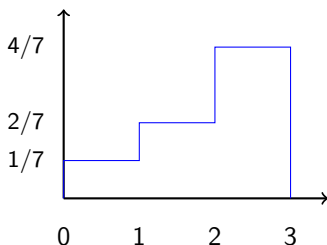
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- ▶ What is the expected time without any conditioning?
- ▶ First, what is the conditional probability, $f_{X|X>2}(x)$?

$$f_{X|X>2}(x) = \begin{cases} 1 & \text{if } 2 < x \leq 3 \\ 0 & \text{otherwise} \end{cases}$$

Conditional expectation

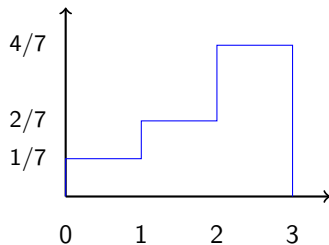


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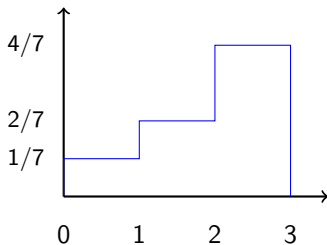
- ▶ So, $E[X|X > 2] = \int_{-\infty}^{\infty} x f_{X|X>2}(x) dx = \int_2^3 x dx = 2.5.$

Conditional expectation



- What is the (unconditional) probability that $X > 2$?

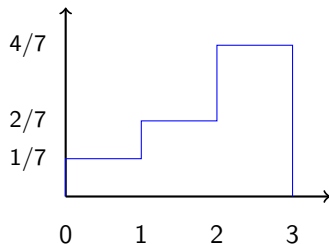
Conditional expectation



- ▶ What is the (unconditional) probability that $X > 2$?

- ▶ $\mathbf{P}(X > 2) = \int_2^3 f_X(x) dx = 4/7$

Conditional expectation



- ▶ What is the (unconditional) probability that $X > 2$?
- ▶ $\mathbf{P}(X > 2) = \int_2^3 f_X(x) dx = 4/7$
- ▶ Similarly, $\mathbf{P}(X < 1) = \int_0^1 f_X(x) dx = 1/7$ and $\mathbf{P}(1 \leq X \leq 2) = 2/7$.

Total expectation theorem

- ▶ What is the total expectation of X ?

Total expectation theorem

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- ▶ $E[X] = \int_{-\infty}^{\infty} xf_X(x)dx$

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Total expectation theorem

- ▶ What is the total expectation of X ?

- ▶ $E[X] = \int_{-\infty}^{\infty} xf_X(x)dx$

- ▶ Total expectation theorem gives:

$$\begin{aligned} E[X] &= E[X|X \leq 1]\mathbf{P}(X \leq 1) + E[X|1 \leq X \leq 2]\mathbf{P}(1 \leq X \leq 2) \\ &\quad + E[X|X > 2]\mathbf{P}(X > 2) \\ &= 0.5 \cdot 1/7 + 1.5 \cdot 2/7 + 2.5 \cdot 4/7 = 27/14 \end{aligned}$$

Independent r.v.'s

A man and a woman agree to meet at a certain location at about 12:30 p.m. Suppose the man arrives at a time uniformly distributed between 12:00 and 12:45, and the woman independently arrives at a time uniformly distributed between 12:15 and 1:00. Let X be the man's arrival time and Y be the woman's arrival time.

- ▶ Find the probability that the first to arrive shows up before 12 : 20.
- ▶ Find the probability that the first to arrive waits no longer than 5 minutes.
- ▶ What is the probability that the man arrives first?

Independent r.v.'s

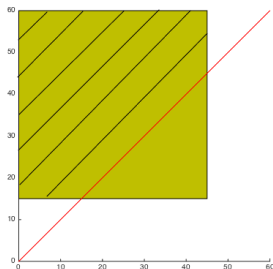
Find the probability that the first to arrive shows up before 12 : 20.

$$\begin{aligned}P(\min(X, Y) \leq 12 : 20) &= 1 - P(\min(X, Y) \geq 12 : 20) \\&= 1 - P(X \geq 12 : 20)P(Y \geq 12 : 20) \\&= 1 - (1 - P(X \leq 12 : 20))(1 - P(Y \leq 12 : 20)) \\&= 1 - (1 - \frac{5}{45})(1 - \frac{20}{45}) \\&= \frac{41}{81}\end{aligned}$$

Independent r.v.'s

Find the probability that the man arrives first.

$$P(X < Y) = 1 - 1/45^2 \times .5 \times 30^2 = 1 - \frac{2}{9} = 7/9$$



Derived distributions

- ▶ We know that a function of a random variable is a random variable.
- ▶ We know that a function of two random variables is a random variable.
- ▶ Sometimes, we need to figure out the distribution of that random variable!
- ▶ This is often pretty easy in the discrete case.
- ▶ e.g. I roll two fair dice. Let X be the number on the first one, and Y be the number on the second one.
- ▶ Let $Z = XY$. What is the PMF of Z ?
- ▶ We can just enumerate through all the options!

Derived distributions

- ▶ $\mathbf{P}(Z = 1) = \mathbf{P}(X = 1)\mathbf{P}(Y = 1) = 1/36$
- ▶ $\mathbf{P}(Z = 2) = \mathbf{P}(X = 1)\mathbf{P}(Y = 2) + \mathbf{P}(X = 2)\mathbf{P}(Y = 1) = 2/36$
- ▶ $\mathbf{P}(Z = 3) = \mathbf{P}(X = 1)\mathbf{P}(Y = 3) + \mathbf{P}(X = 3)\mathbf{P}(Y = 1) = 2/36$
- ▶ $\mathbf{P}(Z = 4) = \mathbf{P}(X = 1)\mathbf{P}(Y = 4) + \mathbf{P}(X = 2)\mathbf{P}(Y = 2) + \mathbf{P}(X = 4)\mathbf{P}(Y = 1) = 3/36...$
- ▶ For general $Z = z$, we can write:

$$\mathbf{P}(Z = z) = \sum_{x=1}^6 \mathbf{P}(X = x)\mathbf{P}(Y = z/x)$$

Independence, joint and marginal

- ▶ $f_{X,Y}(x,y) = ce^{-(x+y^2)/2}$ for $x \geq 0 - \infty < y < \infty$
- ▶ What is c ?
- ▶ Are X, Y independent?
- ▶ What is $\text{var}(X + Y)$?
- ▶ What is $\text{cov}(X, Y)$

Independence, joint and marginal

- ▶ Seems like its a product of an exponential and a normal with 0 mean variance 1.
- ▶ $f_X(x) = .5e^{-.5x}$
- ▶ $f_Y(y) = 1/\sqrt{2\pi}e^{-y^2/2}$
- ▶ $c = \frac{.5}{\sqrt{2\pi}}$
- ▶ Are X, Y independent? Yes!
- ▶ What is $\text{var}(X + Y) = \text{var}(X) + \text{var}(Y) = 2^2 + 1 = 5$.
- ▶ What is $\text{cov}(X, Y)$? 0

Conditional expectation: Conditioning on a concrete event

- ▶ A bird leaves its nest heading north, at 1pm. Let X be the speed at which a bird flies, and Y be the bird's distance north from the nest at 1:30pm.
- ▶ Assume $X \sim \text{Uniform}(3, 6)$ and $Y|X = x \sim \text{Normal}(x/2, 1)$
- ▶ What is $E[Y]$? What is $\text{var}(Y)$?
- ▶ $E[Y|X = x] = x/2$ and so $E[Y|X] = X/2$ and so $E[Y] = E[E[Y|X]] = E[X/2] = 4.5$

Covariance and Correlation

Assume that X and Y are correlated with $E[X] = 1$, $E[Y] = 2$, $E[X^2] = 5$.

- ▶ $\text{var}(X) =$
- ▶ $\text{var}(Y) =$
- ▶ $\text{cov}(X, Y) =$
- ▶ $\rho(X, Y) =$

Covariance

Assume that X and Y are correlated with $E[X] = 1$, $E[Y] = 2$, $E[X^2] = 5$, $E[Y^2] = 20$, and $E[XY] = 3$

- ▶ $\text{var}(X) = 5 - 1 = 4$
- ▶ $\text{var}(Y) = 20 - 4 = 16$
- ▶ $\text{cov}(X, Y) = E[XY] - E[X]E[Y] = 3 - 2 = 1$
- ▶ $\rho(X, Y) = 1/\sqrt{4 \times 16} = 1/8$

Assume that X and Y are correlated with $E[X] = 1$, $E[Y] = 2$, $E[X^2] = 5$.

- ▶ $\text{var}(X) =$
- ▶ $\text{var}(Y) =$
- ▶ $\text{cov}(X, Y) =$
- ▶ $\rho(X, Y) =$

Statistics

- ▶ Inequalities: Markov and Chebyshev
- ▶ We have used the Weak Law of Large numbers and the Central Limit theorem.
- ▶ In Frequentist statistics, we've looked at **the MLE**, **Type I error** and **significance testing**.
- ▶ Today, we will review these with some examples.

Frequentist statistics

- ▶ The frequentist viewpoint is that probability = frequency.
- ▶ Frequentists treat parameters as fixed (they don't change if we repeat the experiment) and observations as random (they do change if we repeat the experiment).
- ▶ Frequentists use **estimators** to estimate the parameters, based on some sequence of n observations.
- ▶ These estimators might be **unbiased** (right on average for any number of observations), **asymptotically unbiased** (right on average as $n \rightarrow \infty$), or **biased** (wrong on average).
- ▶ In general, if we decrease the bias of an estimator, we increase its variance.

The Maximum Likelihood estimator

You have n i.i.d. datapoints $X_1, \dots, X_n \sim N(\mu, \sigma_1^2)$ and n i.i.d. datapoints $Y_1, \dots, Y_n \sim N(\mu, \sigma_2^2)$. What's the MLE of μ ? You can assume that you know σ_1^2 and σ_2^2 .

- ▶ Write the log likelihood: $f(x_1, \dots, x_n, y_1, \dots, y_n) = \prod_i \exp(-(x_i - \mu)^2 / 2\sigma_1^2) \prod_j \exp(-(y_j - \mu)^2 / 2\sigma_2^2)$
- ▶ Take log: $L = - \sum_i \frac{(x_i - \mu)^2}{2\sigma_1^2} - \sum_j \frac{(y_j - \mu)^2}{2\sigma_2^2}$
- ▶ Differentiate: $\frac{dL}{d\mu} = - \sum_i \frac{x_i - \mu}{\sigma_1^2} - \sum_j \frac{y_j - \mu}{\sigma_2^2}$
- ▶ Set to zero and solve for $\hat{\mu}$. $\hat{\mu} = \frac{\bar{x}/\sigma_1^2 + \bar{y}/\sigma_2^2}{1/\sigma_1^2 + 1/\sigma_2^2}$

The Maximum Likelihood estimator

- ▶ Is this biased?

- ▶
$$E[\hat{\mu}] = \frac{\mu/\sigma_1^2 + \mu/\sigma_2^2}{1/\sigma_1^2 + 1/\sigma_2^2} = \mu$$

Confidence intervals

- ▶ Usually, our estimators will correspond (exactly or approximately – central limit theorem!) to the sample mean of a normal distribution.
- ▶ If our estimator is the sample mean of n samples from a normal distribution with known variance σ^2 , the sample mean is distributed according to

$$\bar{X} \sim \text{Normal}\left(\mu, \frac{\sigma^2}{n}\right)$$

- ▶ So, the difference between the true mean and the sample mean can be described as

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim \text{Normal}(0, 1)$$

- ▶ The $(1 - \alpha)$ confidence region for the true mean corresponds to the region where

$$\mathbf{P}\left(\frac{|\bar{X} - \mu|}{\sigma/\sqrt{n}} \leq z\right) = 1 - \alpha$$

- ▶ We can find an appropriate value for z from our standard normal tables.
- ▶ Then, our $(1 - \alpha)$ confidence region is $[\bar{X} - z\sigma/\sqrt{n}, \bar{X} + z\sigma/\sqrt{n}]$

Confidence intervals: unknown variance

- ▶ If our estimator is the sample mean of n samples from a normal distribution with unknown mean and variance, the distribution over \bar{X} is no longer normal.
- ▶ However we still know it's distribution!
- ▶ If we knew σ , then $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$ would be a standard normal random variable.
- ▶ Let $\hat{s}_{n-1}^2 = \sum_i (x_i - \bar{X})^2 / (n - 1)$.
- ▶ Then $T_n = \frac{\bar{X} - \mu}{\hat{s}_{n-1}/\sqrt{n}}$ follows a **t -distribution with $(n - 1)$ degrees of freedom**.
- ▶ We can look up the corresponding probabilities using the t -distribution table.

Exponential confidence interval

I have $X_1, \dots, X_n \sim \text{Exp}(\lambda)$. Find a 95% confidence interval for λ

- ▶ Start with what we know.
- ▶ We want θ^- and θ^+ such that $P(\theta^- \leq \lambda \leq \theta^+) = .95$
- ▶ But $\bar{X} \sim N(1/\lambda, 1/(n\lambda^2))$ as n gets large by CLT
- ▶ $P\left(\sqrt{n} \frac{|\bar{X} - 1/\lambda|}{1/\lambda} \leq 1.96\right) = 0.95$
- ▶ What does this translate to?
- ▶ $P\left(\frac{1}{\lambda} \left(1 - \frac{1.96}{\sqrt{n}}\right) \leq \bar{X} \leq \frac{1}{\lambda} \left(1 + \frac{1.96}{\sqrt{n}}\right)\right) = .95$
- ▶ $P\left(\frac{1}{\bar{X}} \left(1 - \frac{1.96}{\sqrt{n}}\right) \leq \lambda \leq \frac{1}{\bar{X}} \left(1 + \frac{1.96}{\sqrt{n}}\right)\right) = .95$

Hypothesis testing

I have two datasets, X_1, \dots, X_n and Y_1, \dots, Y_n , with known variance 1 and unknown means μ_X and μ_Y . I want to test the null hypothesis that $\mu_X = 2\mu_Y$ and $H_1 : \mu_X \neq 2\mu_Y$ using a two-sided significance test.

- ▶ Suggest an appropriate statistic – specifically, a function of \bar{X} and \bar{Y} – that has zero mean under the null hypothesis. $S = \bar{X} - 2\bar{Y}$?
- ▶ What is the variance of this statistic? $\text{var}(S) = 5/n$
- ▶ Construct an appropriate significance test for the null hypothesis, at significance level $\alpha = 0.05$

$$P(|S| \geq \xi) = P(|S|\sqrt{n}/\sqrt{5} \geq \xi\sqrt{n}/\sqrt{5}) = .95 \text{ So } \xi = 1.96\sqrt{5/n}$$

What you should know

► Frequentist Statistics

- Markov, Chebyshev inequalities and when to use them?
- Weak Law of Large numbers, and Central limit theorem.
- What are biased/unbiased/asymptotically unbiased estimators?
- How to get the MLE. Pay special attention to the Uniform.
- How to get a confidence interval with a given coverage probability?
Normal vs t tables.
- For any rejection region how to get type I error.
- How to get the rejection region with a fixed type I error?
- How to do a significance test with a given level of significance?