

# MBA Business Foundations, Quantitative Methods: Session Five

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# Today

Basics	Functions Linear Inverse Two equations Quadratic
Exponents	Exponents Application: interest rates Exponential functions Logarithmic functions
Logarithms	Logarithmic functions Logarithmic and exponential equations Case: pricing Derivatives
Derivatives	Optimal decisions Case: production Statistics
Uncertainty	Probability & statistics Normal distribution

- **Steeper learning curve** (whether it's worth it, depends on your objectives!)
  - Fast
  - Free
  - Flexible
  - Reproducible
  - Extremely well supported (stackexchange, rbloggers, etc.)
  - Big data friendly (eg. no bound on rows/columns)
  - Visualization and graphics! (with ggplot2)
  - Data science (with the Tidyverse)
  - Machine learning (with Caret, randomForest, nnet, e1071)
  - Bayesian modeling/Markov Chain Monte Carlo (with Bugs, Jags, and Stan)
- ... you learn as you go/as needed, shouldn't try to "learn R"!

- Into to functions (ordinary, inverse, quadratic...)
- Exponents and logs
- Derivatives
- Optimization
- Statistics and probability (last class: measures of location and dispersion)



- Tyche: Goddess of chance (daughter of Zeus).
- The ancient Greeks believed that when no other cause can be attributed to random events such as floods, droughts, frosts, then Tyche is responsible.
- Probability is the study of such random events or, more generally, of randomness.

Examples of random events:



Break downs



Gambling



The weather



The stock market

- But what is probability?
- For our purposes, the probability of an event can be interpreted as the long-run frequency of the occurrence of that event.
- Always???
- What is the probability of a nuclear war in the next year? Of US dollar collapse?
- When Galileo first observed Saturn through a telescope, he saw something like this.



- Are those rings around the planet? Handles? Or is it three planets next to each other? Can you assign a probability?

- Revisit: What is the difference between probability and statistics?
- A probability question: A **fair** die will be tossed twice. What is the probability that it lands on six both times?
- A (descriptive) statistics question: A die was tossed twice and it landed on a five and a six. What is the mean die value?
- An (inferential) statistics question: A die was tossed twice and it landed on a five and a six. How confident are you that the die is **fair**?
- To answer this question, can you use descriptive statistics? Can you use probability? How? Is there a right answer? What is it?

- In probability we know which values our variables  $X$  can take, and we know how probable those values are.
- Ex: if the variable represents the outcome of the toss of a fair die (i.e., which face landed up), what are the values? How likely are they?
- Variables are
  - Discrete, corresponding to natural or counting numbers.
  - Continuous, corresponding to real numbers.
- Classify the following:
  - Height or weight
  - Number of monthly lottery winners in California
  - Temperature tomorrow
  - A parent's number of children
  - The amount of money in your bank account



- A certain couple has two children. At least one of them is a boy. What is the probability that both children are boys?
- Possibilities: BB, BG, GB, GG
- What can we rule out? GG
- What remains: BB, BG, GB
- Probability that both children are boys is  $1/3$ .

- A probability distribution is a function  $f$  of the random variable  $X$ :  $f(X)$
- This function can only take values between 0 and 1.
- Also, this function is additive: for two independent events, the probability of their sum is the sum of their probabilities.
- Ex:  $\Pr(\text{die lands on 4}) + \Pr(\text{die lands on 6}) = \Pr(\text{die lands on 4 or die lands on 6})$ .
- In the discrete case, it tells us how probable it is that the random variable  $X$  will take a specific value  $x$ . We often denote the function  $f$  with  $\Pr$ .
- Ex: For a fair die,  $f(3) = \Pr(X = 3) = 1/6$ .
- In the continuous case, it tells us how likely it is that our variable will be contained within an interval  $[a, b]$ .
- Ex: For a person's weight,  $\Pr(a < x < b) = k$ .
- But what about  $\Pr(x)$  in the continuous case (???)

- A fair die is rolled once.
- What is the probability that it lands on a number greater than 4?

$$\Pr(X > 4) = \Pr(X = 5) + \Pr(X = 6) = 2/6 = 1/3$$

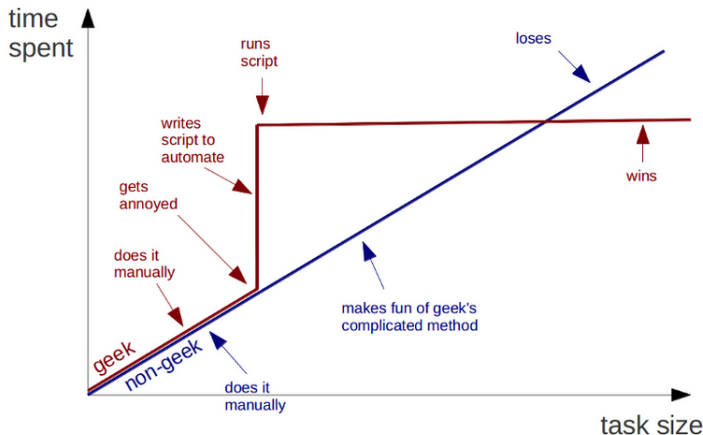
```
library(dice)
getEventProb(nrolls = 1, ndicePerRoll = 1,
  nsidesPerDie = 6, eventList = list(5:6))
[1] 0.3333333
```

- A fair die is rolled 5 times. What is the probability of seeing exactly the pattern 6, 5, 4, 3, followed by a 2 or a 1?

- $\left(\frac{1}{6}\right)^5 + \left(\frac{1}{6}\right)^5 = 0.0002$

```
library(dice)
getEventProb(nrolls = 5, ndicePerRoll = 1, nsidesPerDie = 6,
  eventList = list(6, 5, 4, 3, 1:2), orderMatters = TRUE)
[1] 0.0002572016
```

# Doing things manually vs. automating them: geek vs. non-geek



- We have 30 students in this class. What is the probability that at least one pair of students share the same birthday?
- Algebraic solution:

$$\frac{30 \cdot 29}{2} = 435 \text{ pairs of students}$$

$$\frac{364}{365} \text{ probability that a single pair does not share a birthday}$$

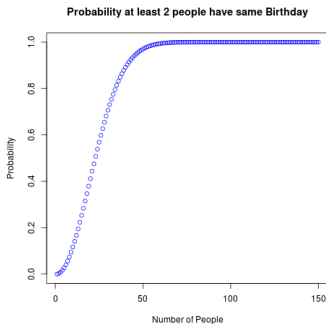
$$\left(\frac{364}{365}\right)^{435} = 0.30 \text{ probability that no pair shares a birthday}$$

$$1 - 0.30 = 0.70 \text{ probability that at least one pair shares a birthday}$$

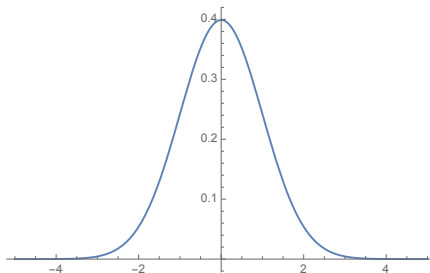
## Birthdays: R solution

```
k = 30
p <- numeric(k) # create numeric vector to store probabilities
for (i in 1:k) {
  q <- 1 - (0:(i - 1))/365 # 1 - prob(no matches)
  p[i] <- 1 - prod(q) }
prob <- p[k]
print(prob)
#BONUS:

plot(p, main="Probability at least 2 people have same Birthday",
     xlab="Number of People", ylab="Probability", col="blue")
[1] 0.7063162
```



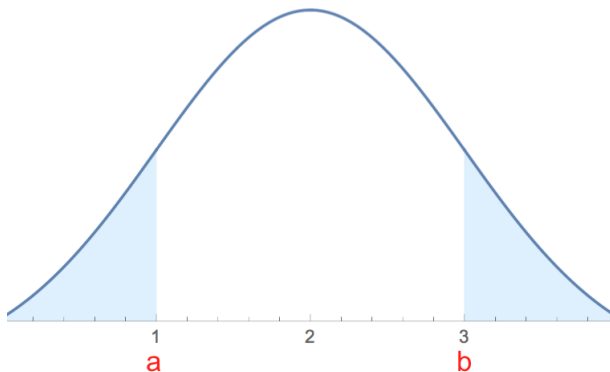
- Most important probability distribution you will encounter (due, in part, to the central limit theorem).
- This distribution belongs to the exponential family of distributions, and it has two parameters, its average  $\mu$  and standard deviation  $\sigma$ .
- Represented by the famous “bell curve”: symmetric around its mean



- Given by

$$f(x|\mu, \sigma^2) = (2\pi\sigma^2)^{-1/2} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

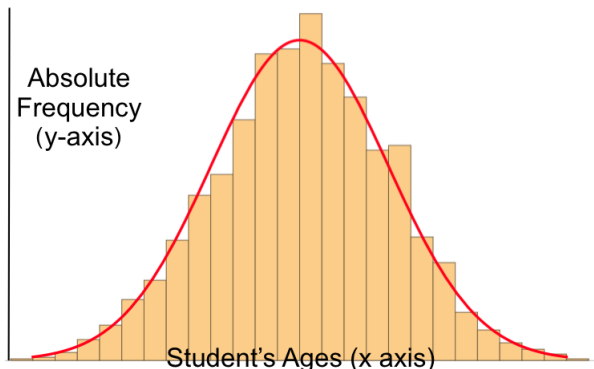
- Bonus: Why is  $\pi$  in there? Because  $\int_{-\infty}^{\infty} e^{-x^2} dx = \sqrt{\pi}$ !



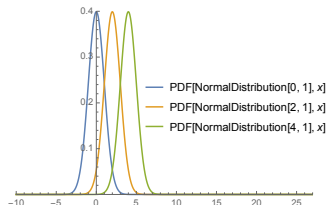
The probability that our random variable is between  $a$  and  $b$  is given by the area under the curve between those two points:

$$\Pr(a < x < b) = (2\pi\sigma^2)^{-1/2} \int_a^b e^{\frac{-(x-\mu)^2}{2\sigma^2}} dx$$

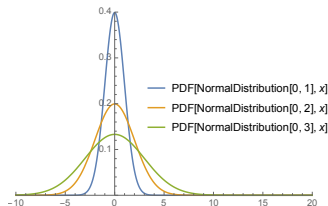




- Variables that have a normal distribution are ubiquitous in real life, provided we have enough data.
- Ex: Age of INSEAD students, height of INSEAD students.

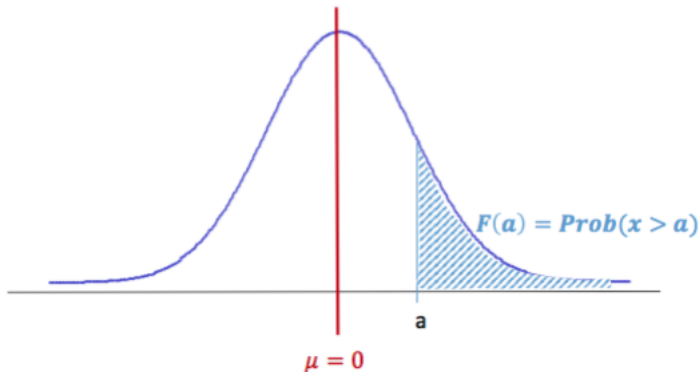


- As the mean changes, the location of the bell shifts  
To the left (for smaller means)  
To the right (for larger means)



- As the standard deviation changes, the bell becomes  
taller and thinner (for smaller standard deviations)  
shorter and thicker (for larger standard deviations)

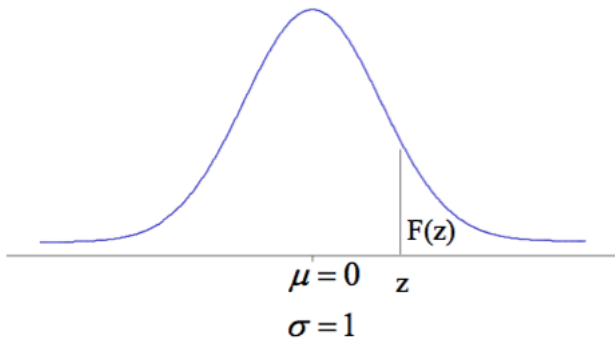
- So how to compute these areas under the curve (=probabilities)?
- The integral does not have a closed form.
- Rescale to a standard normal distribution and then use a table.
- Or, use computational approach, for example in R!



- A standard normal distribution is a normal distribution with mean 0 and variance 1.
- The distribution function of a standard normal is given by

$$(2\pi)^{-1/2} e^{-\frac{x^2}{2}}$$

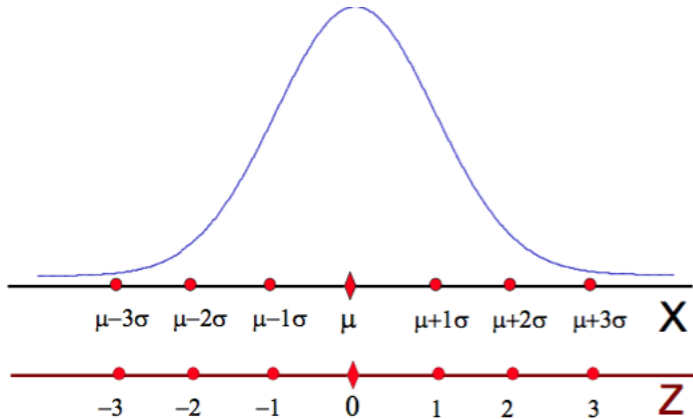
- To denote that a random variable  $X$  has a normal distribution we will use  $X \sim N(\mu, \sigma^2)$ .
- if  $X$  follows a normal distribution with mean  $\mu$  and standard deviation  $\sigma$ ,  $X \sim N(\mu, \sigma^2)$ , then  $Z = \frac{x-\mu}{\sigma}$  follows a standard normal distribution,  $Z \sim N(0, 1)$ .

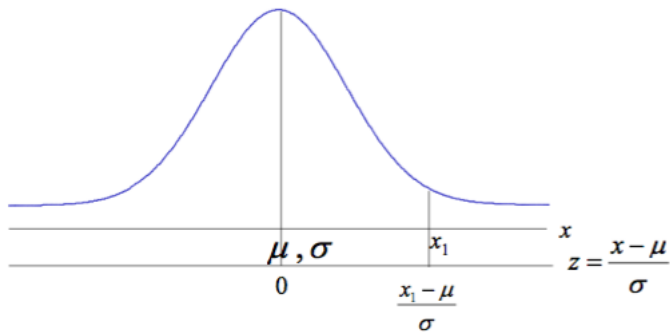


- Capital  $F$  denotes a cumulative distribution:

$$F(z) = \Pr(Z \leq z) = \int_{-\infty}^z f(z) dz$$

- So  $\Pr(a < x < b) = F(b) - F(a)$





- Ex: You want  $\Pr(X > k)$  where  $X \sim N(\mu, \sigma)$ .
- **Step 1:** Transform  $X \rightarrow Z$

$$\begin{aligned}\Pr(X > k) &= \Pr\left(\frac{X - \mu}{\sigma} > \frac{k - \mu}{\sigma}\right) \\ &= \Pr\left(Z > \frac{k - \mu}{\sigma}\right)\end{aligned}$$

- **Step 2:** Look up the probability of  $F(\frac{k - \mu}{\sigma})$  in R or a standard normal table.
- **Step 3:** Get the result:

$$\Pr(X > k) = F\left(\frac{k - \mu}{\sigma}\right)$$



- Suppose that the test scores of a course exam at INSEAD are normally distributed with a mean of 72 and a standard deviation of 15.2. What is the probability that a randomly chosen student received above 84?

$$\Pr(X > 84) \rightarrow \Pr\left(\frac{X - \mu}{\sigma} > \frac{84 - 72}{15.2}\right) \rightarrow \Pr(Z > 0.789)$$

- `pnorm(0.789, lower.tail=FALSE)`
- `1-pnorm(0.789)`
- OR, you can avoid the transformations altogether! `pnorm(84, mean=72, sd=15.2, lower.tail=FALSE)`
- Approximately 21%.
- We use `lower.tail=FALSE` in order to get the area from  $x$  to  $\infty$ .

- The `pnorm` function replaces the lookup table at the end of all statistics textbooks.
- `pnorm` returns the integral from  $-\infty$  to  $k$  of the pdf of the normal distribution. That is,  $F(k)$ .
- If you do not set any further values, then  $k$  is a  $Z$ -score by default.
- However, you can specify the mean and variance as `pnorm(2, mean = 5, sd = 3)`

- The weekly salaries of the employees of a large corporation are assumed to be normally distributed with mean \$450 and standard deviation \$40.
- What is the probability that a randomly chosen employee earns more than \$500 per week?

- Find the Z score and look it up:

$$\Pr(X > 500) = \Pr\left(\frac{X - 450}{40} > \frac{500 - 450}{40}\right) = \Pr\left(Z > \frac{5}{4}\right)$$

- `pnorm(500, mean=450, sd=40, lower.tail=FALSE)`
- approximately 10%.

- Probabilities correspond to areas
- Probabilities sum to 1:  $\Pr(Z < k) = 1 - \Pr(Z > k)$
- Symmetry:  $\Pr(Z < -k) = \Pr(Z > k)$
- For intervals, use subtraction:  $\Pr(a < Z < b) = \Pr(Z > a) - \Pr(Z > b)$

# Practice: but how much money will I make???

- The average post INSEAD MBA starting salary is  $170k$  with a standard deviation of  $30k$ .
- You want to know the probability of earning between  $150$  to  $200k$  upon graduation.
- So you want to calculate the probability that a randomly chosen post-MBA INSEAD student starts at  $150 - 200k$ .
- Find the answer algebraically, then confirm in R.

Algebraic:

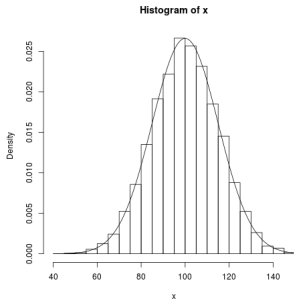
$$\begin{aligned}\Pr(150 < x < 200) &= \Pr(x > 150) - \Pr(x < 200) \\ &= \Pr\left(\frac{x - 170}{30} > \frac{150 - 170}{30}\right) - \Pr\left(\frac{x - 170}{30} > \frac{200 - 170}{30}\right) \\ &= \Pr\left(Z > \frac{150 - 170}{30}\right) - \Pr\left(Z > \frac{200 - 170}{30}\right) \\ &= \Pr\left(Z > -\frac{2}{3}\right) - \Pr(Z > 1) \\ &= .7454 - .1587 \\ &= .5867\end{aligned}$$

In R:

```
pnorm(150, mean=170, sd=30, lower.tail=FALSE) -  
pnorm(200, mean=170, sd=30, lower.tail=FALSE)
```

# The rnorm() function

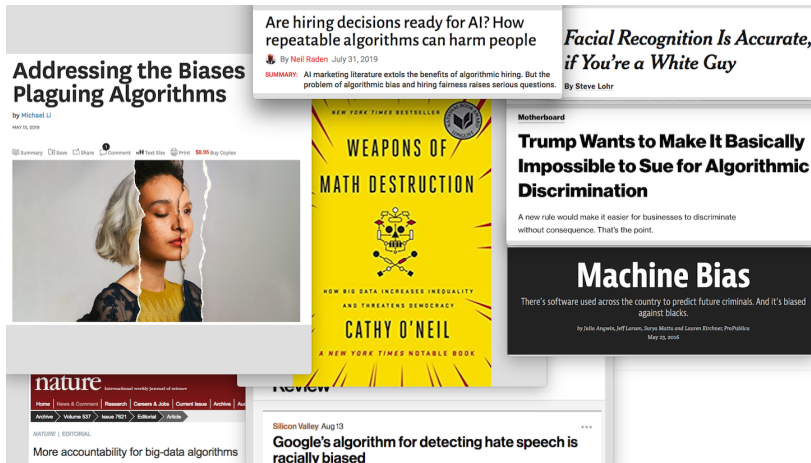
```
x <- rnorm(10000, mean=100, sd=15)
hist(x, probability=TRUE)
xx <- seq(min(x), max(x), length=150)
lines(xx, dnorm(xx, mean=100, sd=15))
```



This generates 10000 random numbers from a specified normal distribution (first line), plots their histogram (second line), and graphs the distribution function of the same normal distribution (third and fourth lines).



- I will teach the MBA elective Management Decision Making (or MDM) in P4 (March/April)!
- You should all take it! Tell your friends too!
- Based on prototype I designed at Caltech called “Statistics, Ethics and Law”



Some topics we will look at:

- **Human bias**
  - Risk and rational choice
  - Statistical reasoning (p-hacking, small samples, overfitting)
  - Implicit bias, anchoring, loss aversion, ambiguity
  - Moral reasoning
- **Machine bias**
  - Algorithmic justice (ex: facial recognition systems)
  - Discrimination in lending, credit scoring, policing
  - Bias in advertising
  - Transparency in algorithmic decision-making (cf: adversarial attacks)
  - Privacy in data collection (for example in health care, on social media)
  - Ethics of self-driving cars and medical technology

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Uncertainty	Probability & statistics Normal distribution



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