

# NOTES ON MATH FOR DATA SCIENCE

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*The following collection was compiled while studying for the degree of masters of data science at City University of New York over the years 2018–present.*

## 1 ALGEBRA

### ARITHMETIC OPERATIONS

$$a \left( \frac{b}{c} \right) = \frac{ab}{c} \quad \frac{\frac{a}{b}}{c} = \frac{a}{bc}$$

$$\frac{\frac{a}{b}}{c} = \frac{ac}{b} \quad \frac{a}{b} \pm \frac{c}{d} = \frac{ad \pm bc}{bd}$$

$$\frac{a-b}{c-d} = \frac{b-a}{d-c} \quad \frac{ab+ac}{a} = b+c, \forall a \neq 0$$

$$\frac{a+bc}{c} = \frac{a}{c} + \frac{b}{c} \quad \frac{\frac{a}{b}}{\frac{c}{d}} = \frac{ad}{bc}$$

### EXPONENTS

$$x^a \cdot x^b = x^{a+b} \quad \frac{x^a}{x^b} = x^{a-b}$$

$$(xy)^a = x^a y^a \quad \frac{x^a}{y^a} = \left( \frac{x}{y} \right)^a$$

$$x^0 = 1^\dagger \quad x^{-n} = \frac{1}{x^n}^\dagger$$

$$x^{\frac{1}{a}} = \sqrt[a]{x}^\dagger \quad (xy)^a = x^a y^a$$

$$x^{-a} = \frac{1}{x^a} \quad \left( \frac{x}{y} \right)^{-a} = \left( \frac{y}{x} \right)^a = \frac{y^a}{x^a}$$

$$^\dagger \forall x \neq 0$$

### RADICALS

$$\sqrt[a]{x} = x^{\frac{1}{a}} \quad \sqrt[n]{xy} = \sqrt[n]{x} \sqrt[n]{y}$$

### LOGARITHMS

Defined as  $y = \log_b x \equiv x = b^y$ , with domain of  $x > 0$

$$\log_b b = 1 \quad \log_b 1 = 0$$

$$\log_b b^x = x \quad b^{\log_b x} = x$$

$$\log_b (x^r) = r \log_b x$$

$$\log_b \left( \frac{x}{y} \right) = \log_b x - \log_b y$$

$$\log_b (xy) = \log_b x + \log_b y$$

### FACTORING

$$x^2 - a^2 = (x+a)(x-a)$$

$$x^2 \pm 2ax + a^2 = (x \pm a)^2$$

$$x^2 + (a+b)x + ab = (x+a)(x+b)$$

### FUNCTIONS

- A rule for a relationship between an input and an output quantity where each input *uniquely determines* an output value
- Must be of the form  $y = f(x)$

TYPE	FORM
Constant	$f(x) = c$
Identity	$f(x) = x$
Absolute	$f(x) =  x $
Quadratic	$f(x) = x^2$
Cubic	$f(x) = x^3$
Reciprocal	$f(x) = \frac{1}{x^2}$
Square root	$f(x) = \sqrt{x}$

## TRANSFORMATIONS OF FUNCTIONS

TRANSFORMATION	FORM
Vertical shift	$f'(x) = f(x) + k$
Horizontal shift	$f'(x) = f(x - k)$
Horizontal reflection	$f'(x) = f(-x)$
Vertical reflection	$f'(x) = -f(x)$
Vertical stretch	$f'(x) = kf(x), k > 1$
Vertical compression	$f'(x) = kf(x), 0 < k < 1$

## DISTANCE FORMULA

$$d(P_1, P_2) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

is the distance between points  $P_1 = (x_1, y_1)$  and  $P_2 = (x_2, y_2)$

## LINEAR FUNCTIONS

- Linear functions have a constant rate of change,  $m$

$$m = \frac{\Delta y}{\Delta x} = \frac{y_2 - y_1}{x_2 - x_1} = \frac{f(x_2) - f(x_1)}{x_2 - x_1}$$

- Estimate horizontal line by solving  $f(x) = 0$
- Find the point where two non-parallel lines meet: set  $f(x) = g(x)$  and solve for  $x$
- Estimate slope from two points by solving for  $m$ :

$$y_1 - y_2 = m(x_1 - x_2)$$

## QUADRATIC FUNCTIONS

Standard form  $f(x) = ax^2 + bx + c$

Vertex or transformation form  $f(x) = a(x - h)^2 + k$

Quadratic eq.  $x = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$

- $b^2 - 4ac > 0 \Rightarrow$  Two real unequal solutions
- $b^2 - 4ac = 0 \Rightarrow$  Repeated real solution
- $b^2 - 4ac < 0 \Rightarrow$  Two complex solutions

## POLYNOMIAL FUNCTIONS

$$f(x) = a_0 + a_1x + a_2x^2 + \cdots + a_nx^n = \sum_{k=0}^n a_kx^{n-k}$$

- $a_0$  Constant term
- $a_n$  Polynomial coefficient
- $a_nx^m$  Term ( $m$  is called *degree*)

## RATIONAL FUNCTIONS

Can be written as a quotient of two polynomials  $P(x)$  and  $Q(x)$ :

$$f(x) = \frac{P(x)}{Q(x)} = \frac{\sum_{k=0}^n a_kx^{n-k}}{\sum_{k=0}^n b_kx^{n-k}}$$

- Horizontal Intercept is the inputs where the output is 0
- Vertical Intercept is where input is 0 (if defined)

## ASYMPTOTES

The "line" a function approaches but never touches

*vertical* A vertical line  $x = a$  where the graph tends towards positive or negative infinity as the inputs approach  $a$ :

$$x \rightarrow a, f(x) \rightarrow \pm \infty$$

*horizontal* A horizontal line  $y = b$  where the graph approaches the line as the input gets larger:

$$x \rightarrow \pm \infty, f(x) \rightarrow b$$

## ASYMPTOTES OF RATIONAL FUNCTIONS

*vertical* Where denominator = 0 but numerator  $\neq 0$

*horizontal* Determined by respective degrees of numerator and denominator:

- Degree of denominator  $>$  degree of numerator  $\Rightarrow$  Horizontal asymptote at  $y = 0$
- Degree of denominator  $<$  degree of numerator  $\Rightarrow$  No horizontal asymptote
- Degree of denominator = degree of numerator  $\Rightarrow$  Horizontal asymptote is ratio of leading coefficients

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### EXPONENTIAL FUNCTIONS

Rate of change is as a percent, i.e., not an constant (absolute) rate. Takes the form:

$$f(x) = a(1 + r)^x$$

or,

$$f(x) = ab^x, \quad b = 1 + r$$

Can always be rewritten in terms of logarithms:

$$b^a = c \equiv \log_b c = a$$

*continuous growth* Use  $e = 2.718282\dots$  for continuous growth, often natural phenomena

$$f(x) = ae^{rx}$$

where

- $a \equiv$  initial quantity
- $r \equiv$  continuous growth rate

*ALMOST ALWAYS USE CONTINUOUS  $e$  FORM!*

Solve by:

1. Isolate exponential expression, where possible
  2. Take log on both sides
  3. Use *exponent property* of logs to pull variables of out exponent
  4. Use algebra to solve for variable
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### GRAPHING EXPONENTIAL FUNCTIONS

- $a \equiv$  vertical intercept
  - $a > 0 \Rightarrow$  concave up
  - $a < 0 \Rightarrow$  concave down
- $b \equiv$  rate of growth
  - $b > 1 \Rightarrow$  growing
  - $0 < b < 1 \Rightarrow$  decaying
- Horizontal asymptote is where  $y = 0$

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### LOGARITHMIC FUNCTIONS

The inverse of exponential functions; use to solve exponential functions. Commonly used to express quantities that vary widely in size. Form:

$$\log_b x$$

which can be rewritten in terms of exponents:

$$b^a = c \equiv \log_b c = a$$

*inverse property of logs*

$$\log_b b^x = x$$

$$b^{\log_b x} = x$$

*exponent property*

$$\log_b A^r = r \log_b A$$

## 2 PROBABILITY

## ASSUMPTIONS

$$0 \leq P(E) \leq 1$$

$$P(S) = 1$$

$$P(S) = \sum_{i=1}^{\infty} P(E_i)$$

Sample space  $S$  contains each event  $E_i$ , e.g.,  $E = \{\text{all outcomes in } S \text{ starting with a 3}\}$

## UNIONS AND INTERSECTIONS

$$E \cup F$$

is the union of the two sets  $E$  and  $F$ , i.e., the event where either  $E$  or  $F$  occurs. The intersection of two events, the outcomes in both  $E$  and  $F$  is

$$E \cap F.$$

Commutative	$E \cup F = F \cup E$ $E \cap F = F \cap E$
Associative	$(E \cup F) \cup G = E \cup (F \cup G)$ $(E \cap F) \cap G = E \cap (F \cap G)$
Distributive	$(E \cup F) \cap G = (E \cap G) \cup (F \cap G)$ $(E \cap F) \cup G = (E \cup G) \cap (F \cup G)$

## INDEPENDENT EVENTS

Two events are independent if knowing the outcome of one provides no useful information about the outcome of the other.

## MUTUALLY EXCLUSIVE EVENTS

$$P(A \cap B) = 0$$

are disjoint events, when  $A$  and  $B$  are mutually exclusive and there is no intersection—it is not possible for both to happen.

## UNION AND ADDITION RULE

$$P(A \cup B) \equiv P(A \vee B) \equiv \{x : x \in A \vee x \in B\}$$

Independent	$P(A \cup B) = P(A) + P(B) - P(A \cap B)$
Mutually exclusive	$P(A \cup B) = P(A) + P(B)$

## INTERSECTION AND MULTIPLICATION RULES

$$P(A \cap B) \equiv P(A \wedge B) \equiv \{x : x \in A \wedge x \in B\}$$

Independent	$P(A \cap B) = P(A) \cdot P(B)$
Mutually exclusive	$P(A \cap B) = 0$
Dependent	$P(A \cap B) = P(A) \cdot P(B A)$

## COMPLEMENT AND SUBTRACTION RULE

$$P(A') \equiv P(A^c) \equiv P(\neg A) \equiv \{x : x \notin A\} \equiv 1 - P(A)$$

$$P(A') = 1 - P(A)$$

Some implications:

$$P(A \cup A^c) = 1$$

$$P(A) = 1 - P(A^c)$$

$$P(A|B) = 1 - P(A^c|B)$$

## CONDITIONAL PROBABILITY RULE

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

is the probability of the outcome of event  $A$  given condition  $B$ .

## BAYES THEOREM

$$P(A|B) = \frac{P(A) P(B|A)}{P(A) P(B|A) + P(A^c) P(B|A^c)}$$

## THE FUNDAMENTAL PRINCIPLE OF COUNTING

If an operation can be performed in  $n_1$  ways, and if for each of these a second operation can be performed  $n_2$  ways, and for each of the first two a third operation can be performed in  $n_3$  ways, ..., then the sequence of  $k$  operations can be performed in  $n_1 n_2 n_k$  ways.

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### FACTORIAL

$$n! = n(n-1)(n-2)(2)(1)$$

with  $0! = 1$ .

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### PERMUTATION

A permutation  $\sigma$  any finite set  $A$  is a one-to-one mapping of  $A$  onto itself. An element mapped to itself in the permutation is a *fixed point*.

*Example*

One permutation of the set  $A = \{a, b, c\}$  is:

$$\sigma = \begin{pmatrix} a & b & c \\ b & c & a \end{pmatrix}$$

where  $a$  is sent to  $b$ ,  $b$  to  $c$ , and  $c$  to  $a$ .

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### PERMUTATIONS OF $n$ ELEMENTS

$$n!$$

is the number of permutations of  $n$  objects, that is, the number of arrangements of a set containing  $n$  elements.

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### PERMUTATIONS: $n$ TAKEN $r$ AT A TIME

$${}_nP_k = \frac{n!}{(n-k)!}$$

represents the number of permutations of  $n$  distinct objects taken  $r$  at a time.

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### STIRLING'S APPROXIMATION OF $n!$

$$n^n e^{-n} \sqrt{2\pi n}$$

is a sequence asymptotically equal to  $n!$ .

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### COMBINATIONS

$${}_nC_j = \binom{n}{j} = \frac{n!}{j!(n-j)!}$$

is the number of subsets of size  $j$  that can be assembled given a set of  $n$  elements, for integers  $n$  and  $j$  such that  $0 < j < n$ .

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### BERNOULLI TRIALS PROCESS

A sequence of  $n$  experiments such that

- Each experiment has two possible outcomes, called *success* and *failure*
- The probability of  $p$  of success of each experiment is the same for each, and is independent of previous experiments

The probability of failure is  $q = 1 - p$ .

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### BINOMIAL PROBABILITIES

$$b(n, p, j) = {}_nC_j p^j (1-p)^{n-j}$$

represents the probability that in  $n$  Bernoulli trials there are exactly  $j$  successes, given  $n$  trials, and each trials' success rate is  $p$ .

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### BINOMIAL DISTRIBUTION

$$B \sim b(n, p, k)$$

gives the probability of the number of successes  $k$  in a sequence of Bernoulli trials with parameters  $p$  and  $n$ .

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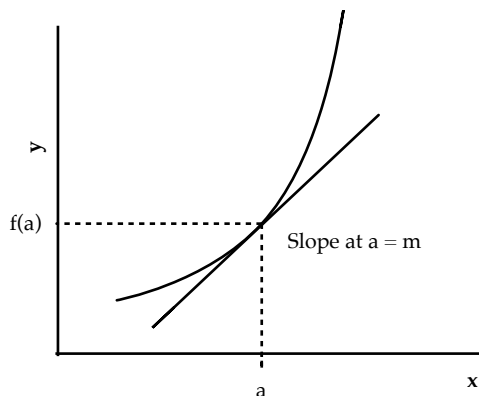
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## 3 CALCULUS

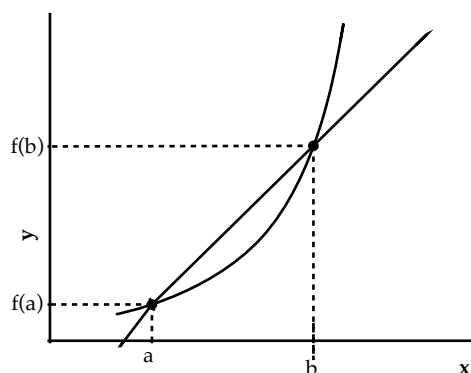
## SIMPLE DERIVATIVES

## TANGENT AND SECANT LINES

- Tangent line is the line on  $f(x)$  at point  $x = a$  that just touches the graph of the function
  - The slope of a non-linear function at one point



- Secant line is the line between A and B on a curve



*Note:* As A gets closer to B the secant slope approaches the tangent line

## AVERAGE RATE OF CHANGE

$$ARC = f(a, a + \Delta) = \frac{\Delta y}{\Delta x} = \frac{f(a + \Delta x) - f(a)}{\Delta x}$$

over interval  $[a, a + \Delta x]$

## DERIVATIVE

The instantaneous rate of change of a function  $f(x)$  at point  $x = a$

$$f'(a) = \lim_{\Delta x \rightarrow 0} \frac{\Delta y}{\Delta x} = \lim_{\Delta x \rightarrow 0} \frac{f(a + \Delta x) - f(a)}{\Delta x}$$

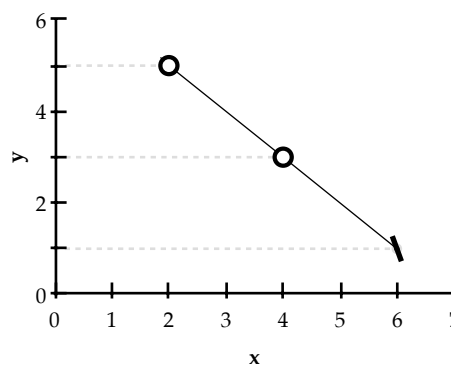
“the derivative of  $y$  w.r.t.  $x$ ”  $\equiv \frac{dy}{dx} \equiv \lim_{\delta \rightarrow 0} \frac{\delta y}{\delta x}$

$y$	$y'$
$k$	$0$
$x$	$1$
$x^n$	$nx^{n-1}$
$\sqrt{x}$	$\frac{1}{2\sqrt{x}}$
$\sqrt[n]{x}$	$\frac{1}{n\sqrt[n]{x^{n-1}}}$
$\frac{1}{x}$	$-\frac{1}{x^2}$
$e^x$	$e^x$
$a^x$	$a^x \ln(a)$
$x^x$	$xx^{x-1} + x^x \ln(x)$
$\ln(x)$	$\frac{1}{x}$
$\log_a(x)$	$\frac{1}{x} \log_a(e)$

## COMPOSITE DERIVATIVES

## Example

$y$	$y'$
$u^n$	$nu^{n-1}u'$
$\sqrt{u}$	$\frac{u'}{2\sqrt{u}}$
$\sqrt[n]{u}$	$\frac{u'}{n\sqrt[n]{u^{n-1}}}$
$\frac{1}{u}$	$\frac{-u'}{u^2}$
$e^u$	$e^u u'$
$a^u$	$a^u \ln(a)u'$
$u^v$	$vu^{v-1}u' + u^v \ln(u)v'$
$\ln(u)$	$\frac{u'}{u}$
$\log_a(u)$	$\frac{u'}{u} \log_a(e)$



where hollow points are undefined and solid points are defined

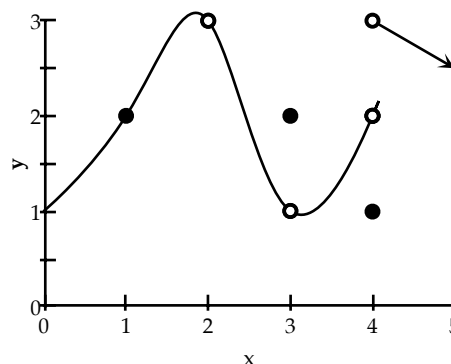
- $\lim_{x \rightarrow 6} f(x) = 1$

- $\lim_{x \rightarrow 4} f(x) = 3$

- Note: even though  $x = 4$  is undefined, we're only concerned with the area around 4, so we can still find the limit

- $\lim_{x \rightarrow 2} f(x) = 5$

## Example: Determining Limits of Non-Linear Functions



where hollow points are undefined and solid points are defined

- $\lim_{x \rightarrow 1} f(x) = 2$ —values where  $x$  is close to but not equal to 1 are near 2

- $\lim_{x \rightarrow 2} f(x) = 3$ —even though  $f(2)$  is undefined, only values near  $f(2)$  are important

- $\lim_{x \rightarrow 3} f(x) = 1$ —even though  $f(3)$  is actually 2

- $\lim_{x \rightarrow 4} f(x)$  = does not exist: Can't determine a single number because  $f(4)$  from the right is about 2, and from the left about 3

## DERIVATIVE OPERATIONS

Sum	$(f(x) + g(x))' = f'(x) + g'(x)$
Difference	$(f - g)'(x) = f'(x) - g'(x)$
Product	$(fg)'(x) = f'(x)g(x) + f(x)g'(x)$
Quotient	$\left(\frac{f(x)}{g(x)}\right)' = \frac{f'(x)g(x) - f(x)g'(x)}{g(x)^2}$
Chain rule	$(f(g))'(x) = f'(g(x))g'(x)$
Inverse	$(f^{-1})'(x) = \frac{1}{f'(x)}$

## LIMITS

$$\lim_{x \rightarrow c} f(x) = L$$

or, “the limit of  $f(x)$ , as  $x$  approaches  $c$ , is  $L$ .”  $\lim_{x \rightarrow c} f(x)$  is a *single number* that describes the behavior of the function  $f(x)$  *near* but not *at* the point  $x = c$ .

Introduced to make calculating rate of change at 0 feasible, by making the  $\Delta$  so infinitesimal the difference is between it and 0 is negligible—“allows” division by 0

*Example: Determining Limits Using Algebra*

Factor equation to simplest form and plug in  $c$  (assuming function at  $c$  is defined):

$$\lim_{x \rightarrow 5} \frac{x^2 - 6x + 8}{x - 4} = \frac{25 - 30 + 8}{1} = 3$$

## LIMITS OF BROKEN FUNCTIONS

Some functions are continuous but in an unusual way—they appear “broken” when graphed—and so there is a *left* and *right* limit

*left* “The limit coming from the left”; values of  $f(x)$  as  $f(x)$  nears  $x$  and left of  $c$ ,  $x < c$

$$\lim_{x \rightarrow c^-} f(x) = L$$

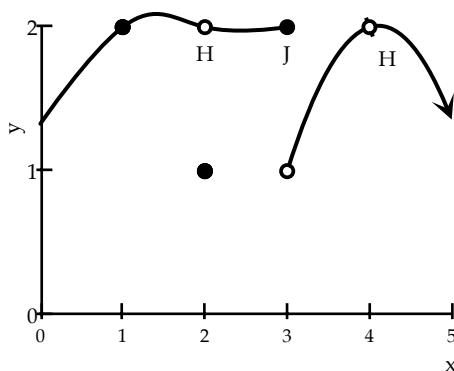
*right* “This limit coming from the right”; values of  $f(x)$  as  $f(x)$  nears  $x$  and right of  $c$ ,  $x > c$

$$\lim_{x \rightarrow c^+} f(x) = L$$

*Note:* If left and rights limits are not the same, limit *doesn't exist*.

## CONTINUITY

A function  $f$  is continuous at  $x = a$  iff  $\lim_{x \rightarrow a} f(x) = f(a)$ , i.e., if the limit of  $x$  at  $a$  is equals  $f(a)$ , i.e., no breaks or jumps

*Example*

where H indicates a hole—where the graph is defined but could be made continuous by changing the point—and J a jump—where the left and right limits are not the same.

- Continuous at 1 since  $\lim_{x \rightarrow 1} f(x) = f(1) = 2$
- Not continuous at 2, 3, or 4
  - $\lim_{x \rightarrow 2} f(x) = 2 \neq f(2) = 1$
  - $\lim_{x \rightarrow 3} f(x) = \text{doesn't exist} \neq f(3) = 2$
  - $\lim_{x \rightarrow 4} f(x) = 2 \neq f(4) = \text{undefined}$

## CALCULATING DERIVATIVES

*Example: Using Formal Definition*

Find derivative of  $f(x) = 2x^2 - 16x + 35$ .

1. Assemble using the formal definition

$$\frac{[2(x-h)^2 - 16(x+h) + 35] - [2x^2 - 16x + 35]}{h}$$

2. Factor—cannot plug  $h = 0$  because no division by zero!

$$\begin{aligned} &= \frac{2x^2 + 4xh + 2h^2 - 16x - 16h + 35 - 2x^2 + 16x - 35}{h} \\ &= \lim_{h \rightarrow 0} \frac{4xh + 2h^2 - 16h}{h} \end{aligned}$$

3. Factor out  $h$  in numerator to cancel  $h$  in denominator

$$\begin{aligned} f'(x) &= \lim_{h \rightarrow 0} \frac{h(4x + 2h - 16)}{h} \\ &= \lim_{h \rightarrow 0} 4x + 2h - 16 \\ &= 4x - 16 \end{aligned}$$

## IMPLICIT DIFFERENTIATION

The process to find  $y' = f'(x)$  when  $f(x)$  is difficult or impossible to use with explicit differentiation, by assuming  $y$  is a function of  $x$ :

*Example*

Implicitly differentiate  $x^2 + y^2 = 25$

1. Differentiate each side, treating  $y$  as a function

$$\begin{aligned} \frac{d}{dx} (x^2 + y^2) &= \frac{d}{dx} 25 \\ \Rightarrow \frac{d}{dx} x^2 + \frac{d}{dx} y^2 &= 0 \\ \Rightarrow 2x + 2y \frac{dy}{dx} &= 0 \\ \Rightarrow 2x + 2yy' &= 0 \end{aligned}$$



2. Algebraically solve for  $y'$

$$\begin{aligned} 2yy' &= -2x \\ \Rightarrow y' &= \frac{-2x}{2y} \\ \Rightarrow y' &= -\frac{x}{y} \end{aligned}$$

#### DEFINITE INTEGRAL

The definite integral of a positive function  $f(x)$  over an interval  $[a, b]$  is the area between  $f$ , the  $x$ -axis,  $x = a$ , and  $x = b$ .

$$\int_a^b f(x) dx$$

where  $a$  and  $b$  are the “limits of integration” and  $f(x)$  is the integrand

$$\begin{aligned} \int_a^a f(x) dx &= 0 \\ \int_a^b f(x) dx &= -\int_b^a f(x) dx \\ \int_a^b f(x) dx &= \int_a^c f(x) dx + \int_c^b f(x) dx \end{aligned}$$

where  $a < c < b$

#### INTEGRATION OPERATIONS

Sum	$\int u + v dx = \int u dx + \int v dx$
Difference	$\int u - v dx = \int u dx - \int v dx$
Product	$\int af(x) dx = a \int f(x) dx$
Parts	$\int u dv = uv - \int v du$
Substitution	$\int f(u)u' dx = \int f(u) du$

#### GROWTH, CONCAVITY, AND EXTREMA

*growth*

- $\forall x \in I \ f'(x) \geq 0 \Rightarrow f$  is increasing in  $I$ .
- $\forall x \in I \ f'(x) \leq 0 \Rightarrow f$  is decreasing in  $I$ .

*concavity*

- $\forall x \in I \ f''(x) \geq 0 \Rightarrow f$  is concave up in  $I$ .
- $\forall x \in I \ f''(x) \leq 0 \Rightarrow f$  is concave down in  $I$ .

*extrema* If  $f'(a) = 0$  (critical point)

- $f''(a) < 0 \Rightarrow f$  has a local maximum at  $x = a$ .

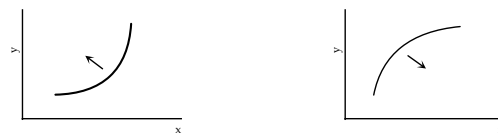


Figure 1: Concave up (left) and concave down (right)

- $f''(a) > 0 \Rightarrow f$  has a local minimum at  $x = a$ .

#### INCREASING AND DECREASING FUNCTIONS

- $f(x)$  is *increasing* iff  $\forall x_1, x_2$  in interval  $I$  is such that  $x_1 < x_2$  and  $f(x_1) < f(x_2)$
- $f(x)$  is *decreasing* iff  $\forall x_1, x_2$  in interval  $I$  is such that  $x_1 > x_2$  and  $f(x_1) > f(x_2)$
- Determine all intervals where  $f(x)$  is in/increasing/decreasing:
  1. Find all critical points (via first derivative)
  2. For each critical point, select a number  $a$  in that range and see if  $f'(a)$  is positive or negative

#### INFLECTION POINTS

- Where the second derivative changes signs
- The point(s) on a graph where the concavity of a function changes from up to down
- functions can be increasing (positive derivative) or decreasing (negative derivative) regardless of concavity,

#### MAXIMA AND MINIMA

A *critical point* is a point where either  $f'(a) = 0$  or  $f'(a)$  is undefined, and is a *candidate* of being a local or global extreme

*Local*

*maximum* at  $a$  if  $f(a) \geq f(x) \ \forall x$  near  $a$

*minimum* at  $a$  if  $f(a) \leq f(x) \ \forall x$  near  $a$

*extreme* at  $a$  if  $f(a)$  is a local maximum or minimum

*Global*

*maximum* at  $a$  if  $f(a) \geq f(x) \forall x$  in domain of  $f$

*minimum* at  $a$  if  $f(a) \leq f(x) \forall x$  in domain of  $f$

*extreme* at  $a$  if  $f(a)$  is a global maximum or minimum

*Example*

Find the critical point of  $f(x) = x^3 - 6x^2 + 9x + 2$

1. Find  $f'(x)$

$$\begin{aligned} f'(x) &= 3x^2 - 12x + 9 \\ &= 3(x^2 - 4x + 3) \\ &= 3(x - 1)(x - 3) \end{aligned}$$

2. Find where  $f'(x) = 0$ , which is 1 and 3
3. Put  $x = 1$  and  $x = 3$  into  $f(x)$  to find the critical points

$$\begin{aligned} (1, f(1)) &= (1, 6) \\ (3, f(2)) &= (3, 2) \end{aligned}$$

## ANTIDERIVATIVES

- An antiderivative of a function  $f(x)$  is any function  $F(x)$  such that  $F'(x) = f(x)$
- The antiderivative is an entire family of functions, written  $F(x) + c$
- Also known as the *indefinite integral* (with no limit markers):

$$\int f(x) \, dx$$

*Example*

An antiderivative of  $\int 2x \, dx$  is  $x^2 - 5.2$ ; the antiderivative is  $x^2 + C$

## DEFINITE V. INDEFINITE INTEGRALS

Indefinite integrals do not have limits to integration where definite integrals do

## INTEGRATION BY SUBSTITUTION

A method to algebraically manipulate an integrand so it is amenable to antiderivative rules; especially useful when there is a product in the integral.

Substitute  $u$  for  $g(x)$  where necessary, making  $\frac{du}{dx} = g'(x)$ , so  $du = g'(x) \, dx$ . Since

$$\frac{du}{dx} = g'(x) \equiv du = g'(x) \, dx$$

we can substitute so that

$$\int f'(g(x))g'(x) \, dx \equiv \int f'(u) \, du$$

Now integrate  $f'(u) \, du$ . (Note that  $g(x) \equiv u$  and  $g'(x) \, dx \equiv du$ .)

1. Set one part of the integrand to  $u$ , one “level” into the integral
2. Compute  $du = \frac{du}{dx} \, dx$  (the derivative of  $u$ )
3. Convert  $x$ 's to  $u$ 's in original integral, even including in  $dx$
4. Integrate new  $u$  integral
5. Substitute  $u$ 's back to  $x$ 's in integral

*Example*

Integrate  $\int (x+1)^3 \, dx$

1. Rearrange so that  $u = x + 1$  and  $du = 1 \, dx$

$$= \int (x+1)^3 \cdot 1 \, dx$$

2. Substitute in  $u$  and  $du$

$$\int u^3 \, du$$

3. Integrate

$$= \frac{u^4}{4} + C$$

4. Add  $u$  back in

$$= \frac{(x+1)^4}{4} + C$$

## INTEGRATION BY PARTS

Integrate a complex function by rewriting it as a product of two simpler functions  $u$  and  $du$ , using two possible forms:

$$\int u \, dv = uv - \int v \, du$$

$$\int_a^b u \, dv = uv \Big|_a^b - \int_a^b v \, du$$

### 3.1 Example: First Form

Integrate  $\int x e^x \, dx$ :

1. Break into two parts:  $u = x$  and  $dv = e^x \, dx$
2. Calculate the derivative of  $u$ ,  $du$ , and  $v$ , the integral of  $dv$

$$du = \left( \frac{d}{dx} x \right) dx = 1 \, dx$$

$$v = \int dv = \int e^x \, dx = e^x$$

3. Using the first formula, noting the prior forms from 1 and 2

$$\begin{aligned} \int u \, dv &= \int x e^x \, dx \\ &= x e^x - \int e^x \, dv \\ &= x e^x - e^x + C \end{aligned}$$

### 3.2 Example: Second Form

Integrate  $\int_1^4 6x^2 \ln x \, dx$ :

1. Break into two parts:  $u = \ln x$  and  $dv = 6x^2$
2. Calculate the derivative of  $u$ ,  $du$ , and the integral of  $dv$ ,  $v$ :

$$du = \frac{d}{dx} \ln x = \frac{1}{x} \, dx$$

$$v = \int 6x^2 \, dx = 6 \int x^2 \, dx = 6 \cdot \frac{x^3}{3} = 2x^3$$

3. Use the second formula

$$\begin{aligned} \int_1^4 6x^2 \ln x \, dx &= 2x^3 \ln x \Big|_1^4 - \int_1^4 2x^3 \frac{1}{x} \, dx \\ &= 2x^3 \ln x \Big|_1^4 - 3x^2 \Big|_1^4 \end{aligned}$$

4. Find the integral the usual way:

$$\begin{aligned} &\left[ (2 \cdot 4^3 \ln(4)) - (2 \cdot 1^3 \ln(1)) \right] - \left[ (3 \cdot 4^2) - (3 \cdot 1^2) \right] \\ &= 128 \cdot \ln(4) - 45 \\ &\approx 132.446 \end{aligned}$$

## ANTIDERIVATIVE RULES

$$\int a \, dx = ax + C$$

$$\int x^n \, dx = \frac{u'}{2\sqrt{u}}$$

$$\int e^x \, dx = e^x + C$$

$$\int a^x \, dx = \frac{a^x}{\ln(a)} + C$$

$$\int \frac{1}{x} \, dx = \ln|x| + C$$

## 4 STATISTICS

## EXPECTED VALUE

$$E(X) = \mu = \sum_{i=1}^k x_i P(X = x_i)$$

for a discrete random variable with  $k$  possible values.

## GENERAL VARIANCE FORMULA

$$\text{Var}(X) = \sigma^2 = \sum_{j=1}^k (x_j - \mu)^2 P(X = x_j),$$

or, the sum of the squared deviations  $(x_j - \mu)^2$  weighted by the corresponding probabilities  $P(X = x_1), \dots, P(X = x_k)$ .

## GENERAL STANDARD DEVIATION

$$\sigma = \sqrt{\sigma^2} = \sqrt{\text{Var}(X)}$$

## LINEAR COMBINATIONS OF VARIABLES

$$Z = aX + bY$$

is a linear combination of the independent, random variables  $X$  and  $Y$  (often  $a$  and  $b$  are 1 or  $-1$ ).

$$\begin{aligned} E(Z) &= a \times E(X) + b \times E(Y) \\ \text{Var}(Z) &= a^2 \times \text{Var}(X) + b^2 \times \text{Var}(Y) \end{aligned}$$

## PROBABILITY DENSITY FUNCTION (PDF)

$$P(a \leq X \leq b) = \int_a^b f(x) dx$$

is a PDF of  $X$ , for any two numbers  $a$  and  $b$  where  $a \leq b$ . I.e., the probability that  $X$  takes on a value in the interval  $[a, b]$  is the area above this interval and below the graph of the density curve.

- $P(X = c) = 0$  for any constant (bins are infinitesimally small)
- $\sum P(x_i) = 1$

## NORMAL DISTRIBUTION V. STANDARD NORMAL

There is an entire family of distributions that can be called normal, but the prototypical distribution with mean of 0 and standard deviation of 1 is called the standard normal. Formally defined by its PDF as:

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

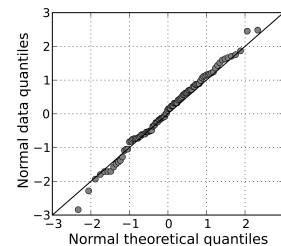
*Properties*

1. Symmetric around mean
2. Mean = mode = median
3. Denser at center than in tails

Consequently,

- 68 percent of distribution is within one standard deviation of the mean
- 95 percent of distribution is within approximately two standard deviations of the mean

## EVALUATING NORMALITY



- A normal probability plot using quantiles can be used to evaluate how closely a given distribution adheres to normality, where the straight line is a perfect normal curve
- As  $N$  increases, the deviation from normality will decrease

## Z SCORES

$$Z = \frac{x - \mu}{\sigma}$$

converts any value from a normal distribution to its corresponding value on the standard normal distribution

- Describes the number of standard deviations a point is from the mean  $\mu$
- Z scores to the left of  $\mu$  are negative, and positive to the right of  $\mu$

#### Z SCORES: PROBABILITIES ON NORMAL DISTRIBUTION

Ex. What is the probability  $X > A$ , given  $X \sim N(\mu = 1500, \sigma = 300)$ ?

$$Z = \frac{x - \mu}{\sigma} = \frac{1630 - 1500}{300} = 0.43$$

This is 0.6664 on Z table, so 66.64 percent of  $X$  is to the left of  $A$  so:

$$1 - 0.6664 = 0.3336$$

The probability  $X > A$  is 33.36 percent.

Ex. Given  $A = 1400$  and  $X \sim N(\mu = 1500, \sigma = 300)$ , what is the percentile corresponding to  $A$ ?

$$Z = \frac{x - \mu}{\sigma} = \frac{1400 - 1500}{300} = -0.33$$

The corresponding value on the Z table is 0.3707, so  $A$  is the 37th percentile.

Ex. Given  $p = .40$  and  $X \sim N(\mu = 70, \sigma = 3.3)$ , what is the value corresponding to percentile  $p$ ?

Lookup  $p$  on Z table, getting a  $Z = -0.25$ . Work backwards:

$$-0.25 = Z = \frac{x - \mu}{\sigma} = \frac{x - 70}{3.3}$$

and solve for  $x = 69.18$ .

Ex. What is the probability  $X$  is between  $A$  and  $B$ , given  $X \sim N(\mu, \sigma)$ ?

Using Z-scores method, find the area to the left of  $A$  and to the right of  $B$ , then  $A - B = 1 - \text{area left of } A - \text{area to right of } B$ .

#### BERNOULLI DISTRIBUTION

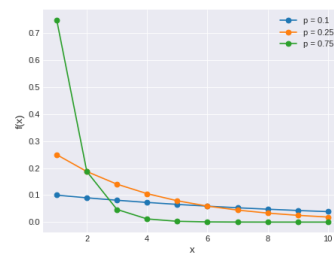
$$P(X = x) = \begin{cases} p & \text{for } x = 1 \\ 1 - p & \text{for } x = 0 \end{cases}$$

describes the distribution of individual trials with two possible outcomes, success or failure, described by proportion of successes  $0 \leq p \leq 1$ :

$$\begin{aligned} \hat{p} &= \frac{|\text{successes}|}{|\text{failures}|} \\ \mu &= p \\ \sigma^2 &= p(1 - p) \end{aligned}$$

- The probability of success after  $n$  trials is  $(1 - p)^{n-1} \times p$

#### BERNOULLI: GEOMETRIC DISTRIBUTION

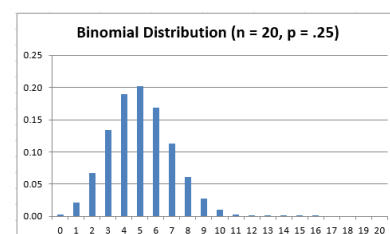


describes the wait time until a success for *independent* Bernoulli random variables; or, the probability of observing the  $k$ -th success by the  $n$ -th trial

$$\begin{aligned} \mu &= \frac{1}{p} \\ \sigma^2 &= \frac{1 - p}{p^2} \end{aligned}$$

- Higher  $p$  means fewer trials until success
- Can never be approximated by a normal distribution

#### BINOMIAL DISTRIBUTION



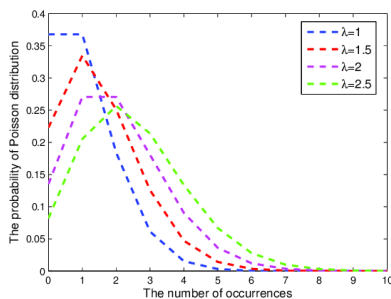
describes the probability of having exactly  $k$  successes in  $n$  independent Bernoulli trials (with probability of success  $p$ ):

$$\begin{aligned} P(X = k | n, \mu, \sigma) &= \binom{n}{k} p^k (1-p)^{n-k} \\ &= \frac{n!}{k!(n-k)!} p^k (1-p)^{n-k} \end{aligned}$$

Parameters, can be used to approximate to normal when  $n$  is sufficiently large and  $np$  and  $n(1-p)$  are both greater than or equal to 10:

$$\begin{aligned} \mu &= np \\ \sigma^2 &= np(1-p) \end{aligned}$$

#### POISSON DISTRIBUTION



$$P(X = x | \lambda) = \frac{\lambda^x e^{-\lambda}}{x!}$$

Describes the number of events in a larger population over a unit of time with rate  $\lambda$ :

$$\begin{aligned} \mu &= \lambda \\ \sigma^2 &= \lambda \end{aligned}$$

#### INFERENCE STATISTICS

The body of thought governing the inferences of populations from samples, and how these sample statistics can vary.

#### STANDARD ERROR

$$SE = \frac{\sigma}{\sqrt{n}}$$

is the standard deviation of distributions of sample statistics, when population  $\sigma$  is known. If it is unknown, and if  $n > 30$ , substitute sample standard deviation  $s$

- SE decreases as  $n$  increases
- SE decreases as  $\sigma$  (or  $s$ ) decreases

#### CONFIDENCE INTERVALS

$$\bar{x} \pm z \times SE$$

- $\bar{x}$  is the sample statistic, such as sample mean
- $z \times SE$  is the *margin of error*
- $z$  is the desired confidence level, e.g.,  $z = 1.96$  for a 95 percent confidence interval

*Interpretation.* "We are  $Z$  percent confident the true population *statistic* is between  $A$  and  $B$ "; or, " $Z$  percent of samples will have a *sample statistic* between  $A$  and  $B$ ."

#### CENTRAL LIMIT THEOREM

Given a population with a finite mean  $\mu$  and a finite non-zero variance  $\sigma^2$ , the sampling distribution of the mean approaches a normal distribution with a mean of  $\mu$  and a variance of  $\frac{\sigma^2}{N}$ , as  $N$ , the sample size, increases—regardless of the shape of the parent population.

#### HYPOTHESIS TESTING

One-Sided	Two-Sided
$H_0 : x = A$	$H_0 : x = A$
$H_A : x > / < A$	$H_A : x \neq A$

The process of comparing two point estimates, to determine if any difference between them is "real" or the result of natural variance in samples.

- *Type I* errors, or false positives, occur when  $H_0$  is true, but rejected
- *Type II* errors, or false negatives, occur when  $H_A$  is true, and  $H_0$  is not rejected

### Quantifying Risk

- The risk of Type I errors is quantified by  $\alpha$ , i.e., the probability the point estimate is more than  $z^*$  standard deviations away from the true population parameter
- The p-value is the probability of observing data at least as favorable to the alternative hypothesis, i.e., as “extreme,” as the present data set, if  $H_0$  is actually true
- If the p-value is less than the chosen  $\alpha$ , data is sufficient to reject  $H_0$

### P-VALUE CALCULATIONS

#### One-Sided

#### Two-Sided

### SAMPLE PROPORTIONS

Population parameter  $\pi$  is sampled:

$$\mu = \hat{p} = \frac{\sum_{i=1}^n x_i}{n}$$

$$SE_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$$

where  $0 \leq \hat{p} \leq 1$  and  $x_i = \{0, 1\}$

### SAMPLE PROPORTIONS: CONFIDENCE INTERVALS

1. Assess normality:
  - At least 10 observations for each  $\{0, 1\}$
  - Sample is less than 10 percent of population and observations are independent
2. Calculate standard error  $SE_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$
3. Determine  $z^*$ , e.g., 1.96
4. Put together point estimate and margin of error:

$$\hat{p} \pm z^* \times SE_{\hat{p}}$$

### SAMPLE PROPORTIONS: HYPOTHESIS TESTS

### FIX

$$H_0 : \hat{p} = 0.5$$

$$H_A : \hat{p} > / \neq 0.5$$

1. Evaluate normality
2. Compute  $SE_{\hat{p}}$  using null hypothesis:

$$SE_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}},$$

$$\text{often } \sqrt{\frac{0.5(1-0.5)}{n}}$$

3. Calculate Z-score using hypotheses:

$$\frac{\hat{p} - \hat{p}_0}{SE_{\hat{p}}}$$

4. Convert Z to p-value and decide whether to reject the null or fail to

### SAMPLE PROPORTIONS: SAMPLE SIZE

### DIFFERENCE OF PROPORTIONS

### DIFFERENCE OF PROPORTIONS: CONFIDENCE INTERVALS

### DIFFERENCE OF PROPORTIONS: HYPOTHESIS TESTS

### DIFFERENCE OF PROPORTIONS: POOLED PROPORTION

### $\chi^2$ GOODNESS OF FIT

$$\chi^2 = \sum_{k=1}^N \frac{(\text{observed}_k - \text{expected}_k)^2}{\text{expected}_k}$$

- k mutually exclusive classes
- n observations of  $x_i$
- one parameter, degrees of freedom df
- follows the chi-square distribution if null hypothesis is true

Summarizes how strongly observed count data deviates from the expected, or null, counts—larger values of  $\chi^2$  indicate stronger deviation

Does a statistical model fit this sample?

1. Develop hypotheses:
  - a)  $H_0$ : Sample follows distribution D
  - b)  $H_A$ : Sample does not follow distribution D
2. Check assumptions
  - a) Each expected count must be at least 5
  - b) Can use binning to get around this
3. Establish expected counts (expected proportion of total count in each bin):
 
$$E_k = \text{expected}_k \times n$$
4. Compute  $\chi^2$  statistic
5. Validate assumptions hold to apply  $\chi^2$  to  $\chi^2$  distribution
6. Using  $k - 1$  degrees of freedom, use  $\chi^2$  table to compute a p-value
7. Decide to reject or fail to reject  $H_0$

$\chi^2$ : p-VALUE

TWO-WAY TABLES: INDEPENDENCE

SAMPLE STATISTICS: MEAN AND VARIANCE

- $\mu_M = \mu$  is the mean of the sampling distribution of means
- $\sigma_M^2 = \frac{\sigma^2}{N}$  is the variance of the sampling distribution of the mean
- $\sigma_M = \frac{\sigma}{\sqrt{N}}$  is the standard error of the sampling distribution of the mean

As  $N$  increases, variance of sample mean decreases

SAMPLE STATISTICS: DIFFERENCE IN MEAN

Two samples from a population the size  $n_1$  and  $n_2$ , calculate the means  $M_1$  and  $M_2$ , and the difference is  $M_1 - M_2$

$$\begin{aligned}\mu_{M_1-M_2} &= M_1 - M_2 \\ \sigma_{M_1-M_2}^2 &= \sigma_{M_1}^2 + \sigma_{M_2}^2 \\ \sigma_{M_1-M_2} &= \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}\end{aligned}$$

When variance and sample size are the same, standard error becomes:

$$\sigma_{M_1-M_2} = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}} = \sqrt{\frac{\sigma^2}{n} + \frac{\sigma^2}{n}} = \sqrt{\frac{2\sigma^2}{n}}$$

If  $n_1 \neq n_2$  then variance becomes:

$$\sigma_{M_1-M_2}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

What is the probability that the mean of sample 1 will exceed that of sample 2 by  $N$  or more?

1. Find mean:  $\mu_{M_1-M_2} = M_1 - M_2$
2. Find standard error:  $\sigma_{M_1-M_2}$
3. Find area underneath distribution of sample 1 to the right of the mean of sample 2 plus  $N$

SAMPLE STATISTICS:  $r$  AND  $\rho$

- Not normally distributed—right-skewed—because correlation cannot exceed 1
- As  $\rho$  increases, the more right-skewed the distribution

SAMPLE STATISTICS: PROPORTION  $\pi$

Sampling proportion is closely related to the binomial distribution—the total number of successes—where  $p$  is the distribution of the mean number of successes

$$\begin{aligned}\mu_p &= \pi \\ \sigma_p &= \frac{\sqrt{N\pi(1-\pi)}}{N} = \sqrt{\frac{\pi(1-\pi)}{N}}\end{aligned}$$

Find probability  $p$  is greater than  $A$

Given  $N$  and population proportion  $\pi$ :

1. Find mean of  $p = \pi$
2. Calculate standard error as above
3. Conduct as normal distribution given  $N$  is sufficiently large and  $\pi$  is not too close to 0 or 1



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ESTIMATION

The process of estimating population parameters from sample statistics. Usually results in a point estimate as well as interval estimates called confidence intervals.

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DEGREES OF FREEDOM

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## 5 REGRESSION: OLS

## FORM

$$y = \beta_0 + \beta_1 x$$

describes the true, unobserved model, while

$$\hat{y} = b_0 + b_1 x$$

describes the estimated model. The estimate  $\hat{y}$  describes the average value around which subjects where  $x = x_i$  will cluster.

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## RESIDUALS

$$e_i = y_i - \hat{y}_i$$

is the residual of the  $i$ th observations  $(x_i, y_i)$ , the differences between the observed response  $(y_i)$  and the prediction  $\hat{y}$

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## CORRELATION

$$r = \frac{1}{n-1} \sum_{i=1}^n \frac{x_i - \bar{x}}{s_x} \frac{y_i - \bar{y}}{s_y}$$

describes the strength of the linear relationship between two variables  $x$  and  $y$ , where  $0 \leq r \leq 1$ , and  $s$  is sample standard deviation

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## LEAST SQUARES CRITERION

$$\arg \min \sum_{i=1}^n e_i^2 \equiv e_1^2 + e_2^2 + \dots + e_n^2$$

describes the best fitting line, the *least squares line*, i.e., minimizes the sum of squared residuals. To calculate:

$$b_1 = \frac{s_y}{s_x} r$$

then use the fact that the point  $(\bar{x}, \bar{y})$  is on the least squares line to set  $x_0 = \bar{x}$  and  $y_0 = \bar{y}$  along with the slope  $b_1$ , solve for  $x$  in:

$$y - \hat{y} = b_1(x - \hat{x})$$

*Assumptions*

1. *Linearity.* Data must show a linear trend

2. *Near normal residuals*
  3. *Constant variability.* The variability of points around the least-squares line must be constant, e.g., the scale of  $e$  cannot increase as  $x$  increases producing a fanning pattern
  4. *Independent observations.*
- 

$r^2$

$$r^2 = 1 - \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y})^2} = 1 - \frac{SSE}{SST}$$

- SSE is summed squares of residuals:
  - SSTO is total sum of squares
- 

#### REGRESSION: T-TEST

Tests to determine if null hypothesis  $b_1 = 0$  is to be rejected

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