

UCLA STAT 251
Statistical Methods for the Life and Health Sciences

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University of California, Los Angeles, Winter 2002
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Continuous Random Variables

- The Normal Distribution
- Sums and differences of random quantities

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Dietary intake of carbohydrate (mg/day) for 5929 people from a variety of work environments. Standardized histogram plot is unimodal but skewed to the right (high values). Vertical scale is (relative freq.)/(interval width) = f_i/m_n . The proportion of the data in $[a : b]$ is the area under the standardized histogram on the range $[a : b]$.

(a) Standardized histogram

(b) Area between $a = 225$ and $b = 375$ shaded

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Superposition of a smooth curve (density function) on the standardized histogram (left panel). Area under the density curve on $[a : b] = [225 : 375]$ is analytically computed to be: 0.486 (right panel), which is close to the empirically obtained estimate of the area under the histogram on the same interval: 0.483 (left panel).

(c) With approximating curve

(d) Area between $a = 225$ and $b = 375$ shaded

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Standardized histograms

For a standardized histogram:

- The vertical scale is $\text{Relative_frequency} / \text{Interval_width}$
- Total area under histogram = 1
- Proportion of the data between a and b is the **area** under histogram between a and b

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Probability and areas

For a continuous X

- the probability a random observation falls between a and b = area under the **density curve** between a and b .

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Sampling from the distribution

[SamplingDistributionApplet.html](#)

(a) Dot plots of 6 sets of 15 random observations

Note the fair amount of **intra- and inter-group variability**. What does that mean? Is that **normal or expected**?

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Two standardized histograms with approximating Normal density curve

Normal density curve has $\mu = 39.8$ in., $\sigma = 2.05$ in.

(a) Chest measurements of Quetelet's Scottish soldiers (in.)

Normal density curve has $\mu = 174$ cm, $\sigma = 6.57$ cm

(b) Heights of the 4294 men in the workforce database (cm)

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The Normal distribution density curve

- Is symmetric about the mean! Bell-shaped and unimodal.
- Mean = Median!

$N(\mu, \sigma)$

50% 50%

Mean μ

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Effects of μ and σ

(a) Changing μ
shifts the curve along the axis

Mean is a measure of ...
central tendency

$\sigma_1 = \sigma_2 = 6$

$\mu_1 = 160$ $\mu_2 = 174$

(b) Increasing σ
increases the spread and flattens the curve

$\sigma_1 = 6$ $\sigma_2 = 12$

Standard deviation is a measure of ...
variability/spread

$\mu_1 = \mu_2 = 170$

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Understanding the standard deviation: σ

Probabilities/areas and numbers of standard deviations for the Normal distribution

Shaded area = 0.683 Shaded area = 0.954 Shaded area = 0.997

$\mu - \sigma$ μ $\mu + \sigma$ $\mu - 2\sigma$ μ $\mu + 2\sigma$ $\mu - 3\sigma$ μ $\mu + 3\sigma$

68% chance of falling between $\mu - \sigma$ and $\mu + \sigma$ 95% chance of falling between $\mu - 2\sigma$ and $\mu + 2\sigma$ 99.7% chance of falling between $\mu - 3\sigma$ and $\mu + 3\sigma$

[NormalCurveInteractive.html](#)

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Basic method for obtaining probabilities

- Sketch a **Normal curve**, marking the mean and other values of interest.
- Shade the area** under the curve that gives the desired probability.
- Devise a way of getting the desired area from **lower-tail areas**.
- Obtain component lower-tail **probabilities from a computer program**

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(a) Computing $\text{pr}(160 < X \leq 180)$

Programs supply

$\text{pr}(X \leq 180)$ and $\text{pr}(X \leq 160)$

We want

$\text{pr}(160 < X \leq 180) = \text{difference}$

$\text{pr}(160 < X \leq 180) = \text{pr}(X \leq 180) - \text{pr}(X \leq 160)$

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Tabular representation of probabilities

(c) More Normal probabilities
(values obtained from Minitab)

b	$\text{pr}(X \leq b)$	a	$\text{pr}(X \leq a)$	$\text{pr}(a < X \leq b) = \text{difference}$
167.6	0.165	152.4	0.001	0.164
177.8	0.718	167.6	0.165	0.553
177.8	0.718	152.4	0.001	0.717
182.9	0.912	167.6	0.165	0.747

Note: 152.4cm = 5ft, 167.6cm = 5ft 6in., 177.8cm = 5ft 10in., 182.9cm = 6ft

From Chance Encounters by C.J. Wild and G.A.F. Seber, © John Wiley & Sons, 2000.

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Obtaining an upper-tail probability

We want

$\text{pr}(X > 25) = ??$

Programs supply

$\text{pr}(X \leq 25) = 0.2874$

Since total area under curve = 1, $\text{pr}(X > 25) = 1 - \text{pr}(X \leq 25)$

Generally, $\text{pr}(X > x) = 1 - \text{pr}(X \leq x)$

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The inverse problem – Percentiles/quantiles

(a) p-Quantile

Programs supply x_p
 x -value for which $\text{pr}(X \leq x_p) = p$

prob = p

$x_p = ??$

(b) 80th percentile (0.8-quantile) of women's heights

Normal ($\mu = 162.7$)

prob = 0.8

$\mu = 162.7$, $x_{0.8} = ??$

Program returns
Thus 80% lie below

(c) Further percentiles of women's heights

Percent	1%	5%	10%	20%	30%	70%	80%	90%	95%
Probn	0.01	0.05	0.1	0.2	0.3	0.7	0.8	0.9	0.95
Percentile (for quantile)	148.3	152.5	154.8	157.5	159.4	166.0	167.9	170.6	172.5
(cm)	410"	50"	50"	52"	52"	55"	56"	57"	58"

The inverse problem is what is the height for the 80th percentile/quantile? So far we studied given the height value what's the corresponding percentile?

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Standard Normal Curve

- The standard normal curve is described by the equation:

$$y = \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}}$$

Where remember, the natural number $e \sim 2.7182\dots$
We say: $X \sim \text{Normal}(\mu, \sigma)$, or simply $X \sim N(\mu, \sigma)$

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Standard Normal Approximation

- The standard normal curve can be used to estimate the percentage of entries in an interval for any process. Here is the protocol for this approximation:
 - Convert the interval (we need to assess the percentage of entries in to standard units). We saw the algorithm already.
 - Find the corresponding area under the normal curve (from tables or online databases);

Report back %

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General Normal Curve

- The **general normal curve** is defined by:
 - Where μ is the **average** of (the symmetric) normal curve, and σ is the **standard deviation** (spread of the distribution).
- Why worry about a **standard** and **general normal curves**?
- How to convert between the two curves?

$$y = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$

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Areas under Standard Normal Curve – Normal Approximation

- Protocol:**
 - Convert the interval (we need to assess the percentage of entries in) to **Standard units**. Actually convert the end points in Standard units.
 - In general, the transformation $X \rightarrow (X-\mu)/\sigma$, **standardizes** the observed value X , where μ and σ are the **average** and the **standard deviation** of the distribution X is drawn from.
 - Find the corresponding area under the normal curve (from tables or online databases);
 - Sketch the normal curve and shade the area of interest
 - Separate your area into individually computable sections
 - Check the Normal Table and extract the areas of every sub-section
 - Add/compute the areas of all sub-sections to get the total area.

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Obtaining central range for symmetric distributions

What values contain the **central 50%**?

What does that say about the **lower tails**?

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The z-score

- The **z-score** of x is the number of standard deviations x is from the mean. (Body-Mass-Index, BMI)

X	z -score = $(x - \mu) / \sigma$	Interpretation
Male BMI values (kg/m ²)		
25	$(25-27.3)/4.1 = -0.56$	25 kg/m ² is 0.56 sd's below the mean
35	$(35-27.3)/4.1 = 1.88$	35 kg/m ² is 1.88 sd's above the mean
Female heights (cm)		
155	$(155-162.7)/6.2 = -1.24$	155cm is 1.24 sd's below the mean
180	$(180-162.7)/6.2 = 2.79$	180cm is 2.79 sd's above the mean

Male BMI-values: $\mu=27.3, \sigma=4.1$ Females heights: $\mu=162.7, \sigma=6.2$

- Which ones of these are **unusually** large/small/away from the mean?

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The standard Normal distribution

Standard Normal distribution:

mean(μ) = 0, SD(σ) = 1

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Working in standard units

What values contain the **central 90%**?

What does that say about the **lower tail**?

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Review

What values contain the **central 90%**?

What does that say about the lower tail?

Obtain z from program
[Program returns 1.6449]

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Working in standard units (# of SD's)

TABLE 6.3.2 Central Ranges

Percentage	z	Male BMI values		Female heights	
		$\mu - z\sigma$	$\mu + z\sigma$	$\mu - z\sigma$	$\mu + z\sigma$
80%	1.2816	22.05	32.55	154.8	170.6
90%	1.6449	20.56	34.04	152.5	172.9
95%	1.9600	19.26	35.34	150.5	174.9
99%	2.5758	16.74	37.86	146.7	178.7
99.9%	3.2905	13.81	40.79	142.3	183.1

Male BMI-values: $\mu=27.3, \sigma=4.1$
Females heights: $\mu=162.7, \sigma=6.2$

Standardizing $Z = (X - \mu) / \sigma$
Inverting $X = Z\sigma + \mu$

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TABLE 6.3.3 Using z-score tables

As an example, we shall find $\text{pr}(Z \leq 1.1357)$ using part of the table given in Appendix A4 (reproduced below).

Step 1: Correct the z -value to two decimal places, that is, use $z = 1.14$.

Step 2: Look down the z column until you find 1.1. This tells you which row to look in.

Step 3: The second decimal place, here 4, tells you which column to look in.

Step 4: The entry in the table corresponding to that row and column is $\text{pr}(Z \leq 1.14) = 0.873$

z	0	1	2	3	4	5	6	7	8	9
1.0	.841	.844	.846	.848	.851	.853	.855	.858	.860	.862
1.1	.864	.867	.869	.871	.873	.875	.877	.879	.881	.883
1.2	.885	.887	.889	.891	.893	.894	.896	.898	.900	.901
1.3	.903	.905	.907	.908	.910	.911	.913	.915	.916	.918
1.4	.919	.921	.922	.924	.925	.926	.928	.929	.931	.932

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Summary

QuincunxApplet.html

Show Sampling Distribution Simulation Applet
<file:///C:/Ivo dir/UCLA Classes/Winter2002/AdditionalInstructorAids/SamplingDistributionApplet.html>

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The Normal distribution

$X \sim \text{Normal}(\mu_x = \mu, \sigma_x = \sigma)$

Features of the Normal density curve:

- The curve is a symmetric bell-shape centered at μ .
- The standard deviation σ governs the spread.
 - 68.3% of the probability lies within 1 standard deviation of the mean
 - 95.4% within 2 standard deviations
 - 99.7% within 3 standard deviations

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Probabilities

- Computer programs provide lower-tail (or cumulative) probabilities of the form $\text{pr}(X \leq x)$
 - We give the program the x -value; it gives us the probability.
- Computer programs also provide inverse lower-tail probabilities (or quantiles)
 - We give the program the probability; it gives us the x -value.
- When calculating probabilities, we shade the desired area under the curve and then devise a way of obtaining it via lower-tail probabilities.

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Standard Units

The z-score of a value a is

- the number of standard deviations a is away from the mean
- positive if a is above the mean and negative if a is below the mean.

The *standard Normal* distribution has $\mu = 0$ and $\sigma = 1$.

- We usually use Z to represent a random variable with a standard Normal distribution.

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Ranges, extremes and z-scores

Central ranges:

- $P(-z \leq Z \leq z)$ is the same as the probability that a random observation from an arbitrary Normal distribution falls within z SD's either side of the mean.

Extremes:

- $P(Z \geq z)$ is the same as the probability that a random observation from an arbitrary Normal distribution falls more than z standard deviations above the mean.
- $P(Z \leq -z)$ is the same as the probability that a random observation from an arbitrary Normal distribution falls more than z standard deviations below the mean.

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Combining Random Quantities

Variation and independence:

- No two animals, organisms, natural or man-made objects are ever identical.
- There is always variation. The only question is whether it is large enough to have a practical impact on what you are trying to achieve.
- Variation in component parts leads to even greater variation in the whole.

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Independence

We model variables as being independent

- if we think they relate to physically independent processes
- and if we have no data that suggests they are related.

Both sums and differences of independent random variables are more variable than any of the component random variables

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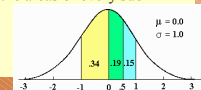
Formulas

- For a constant number a , $E(aX) = aE(X)$
and $SD(aX) = |a| SD(X)$.
- Means of sums and differences of random variables act in an obvious way
 - the mean of the sum is the sum of the means
 - the mean of the difference is the difference in the means
- For independent random variables, (cf. Pythagorean theorem),
 $SD(X_1 + X_2) = SD(X_1 - X_2) = \sqrt{SD(X_1)^2 + SD(X_2)^2}$
 $E(X_1 + X_2) = E(X_1) + E(X_2)$
[ASIDE: Sums and differences of independent Normally distributed random variables are also Normally distributed]

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Areas under Standard Normal Curve – Normal Approximation

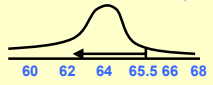
- Protocol:
 - Convert the interval (we need to assess the percentage of entries in) to **Standard units**. Actually convert the end points in Standard units.
 - In general, the transformation $X \rightarrow (X-\mu)/\sigma$, **standardizes** the observed value X , where μ and σ are the **average** and the **standard deviation** of the distribution X is drawn from.
 - Find the corresponding area under the normal curve (from tables or online databases);
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Areas under Standard Normal Curve – Normal Approximation, Scottish Army Recruits

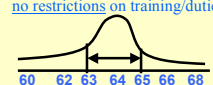
- The mean height is 64 in and the standard deviation is 2 in.
 - Only recruits shorter than 65.5 in will be trained for tank operation. What percentage of the incoming recruits will be trained to operate armored combat vehicles (tanks)?




$X \rightarrow (X-64)/2$
 $65.5 \rightarrow (65.5-64)/2 = 1/4$
 Percentage is 77.34%



- Recruits within 1/2 standard deviations of the mean will have no restrictions on duties. About what percentage of the recruits will have no restrictions on training/duties?



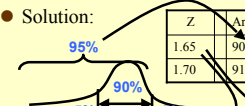
$X \rightarrow (X-64)/2$
 $65 \rightarrow (65-64)/2 = 1/2$
 $63 \rightarrow (63-64)/2 = -1/2$
 Percentage is 38.30%



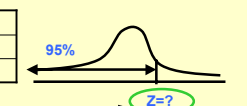
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Percentiles for Standard Normal Curve

- When the histogram of the observed process follows the normal curve Normal Tables (of any type, as described before) may be used to estimate percentiles. The N-th percentile of a distribution is P is N% of the population observations are less than or equal to P.
- Example, suppose the Math-part SAT scores of newly admitted freshmen at UCLA averaged 535 (out of [200:800]) and the SD was 100. Estimate the 95 percentile for the score distribution.
- Solution:



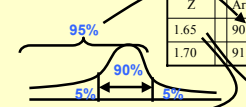
Z	Area
1.65	90.11
1.70	91.09



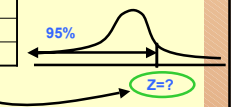
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Percentiles for Standard Normal Curve

- Example, suppose the Math-part SAT scores of newly admitted freshmen at UCLA averaged 535 (out of [200:800]) and the SD was 100. Estimate the 95 percentile for the score distribution.
- Solution:



Z	Area
1.65	90.11
1.70	91.09



- Z=1.65 (std. Units) \rightarrow 700 (data units), since
- $X \rightarrow (X - \mu)/\sigma$, converts data to standard units and
- $X \rightarrow \sigma X + \mu$, converts standard to data units!
- $\sigma = 100; \mu = 535; 100 \times 1.65 + 535 = 700$.

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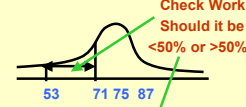
Summary

- The Standard Normal curve is symmetric w.r.t. the origin (0,0) and the total area under the curve is 100% (1 unit)
- Std units indicate how many SD's is a value below (-)/above (+) the mean
- Many histograms have roughly the shape of the normal curve (bell-shape)
- If a list of numbers follows the normal curve the percentage of entries falling within each interval is estimated by: 1. Converting the interval to StdUnits and, 2. Computing the corresponding area under the normal curve (Normal approximation)
- A histogram which follows the normal curve may be reconstructed just from (μ, σ^2) , mean and variance=std_dev²
- Any histogram can be summarized using percentiles
- $E(aX+b)=aE(X)+b$, $Var(aX+b)=a^2Var(X)$, where E(Y) the mean of Y and Var(Y) is the square of the StdDev(Y),

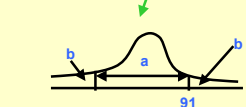
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Example – work out in your notebooks

- Compute the chance a random observation from a distribution (symmetric, bell-shaped, unimodal) with $m=75$ and $SD=12$ falls within the range [53 : 71].



Check Work
Should it be
<50% or >50%?

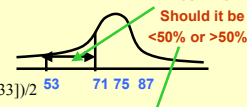


91

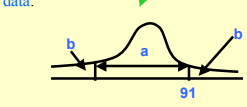
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Example – work out in your notebooks

- Compute the chance a random observation from a distribution (symmetric, bell-shaped, unimodal) with $m=75$ and $SD=12$ falls within the range [53 : 71].
- $(53-75)/12 = -11/6 = -1.83$ Std unit
- $(71-75)/12 = -0.333(3)$ Std units
- Area[53:71] =
- $(SN_area[-1.83:1.83] - SN_area[-0.33:0.33])/2$
- $= (93\% - 25\%)/2 = 34\%$
- Compute the 90th percentile for the same data:
- $b+a+b=100\% \quad a=80\% \quad \rightarrow A=0.8$
- $a+b=90\% \quad b=10\% \quad \rightarrow Z=1.3$ SU
- $90\% P = \sigma 1.3 + \mu = 12 \times 1.3 + 75 = 90.6$



Check Work
Should it be
<50% or >50%?



91

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General Normal Curve

- The **general normal curve** is defined by:
 - Where μ is the **average** of (the symmetric) normal curve, and σ is the **standard deviation** (spread of the distribution).

$$y = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$

- Why worry about a **standard** and **general normal curves**?
- How to convert between the two curves?

$\mu = 0.0$
 $\sigma = 1.0$

$\mu = 52$
 $\sigma = 6$

$\mu = 52$
 $\sigma = 12$

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Areas under Standard Normal Curve

- Many histograms are similar in shape to the **standard normal curve**. For example, persons height. The height of all incoming female army recruits is measured for custom training and assignment purposes (e.g., very tall people are inappropriate for constricted space positions, and very short people may be disadvantages in certain other situations). The mean height is computed to be 64 in and the standard deviation is 2 in. Only recruits shorter than 65.5 in will be trained for tank operation and recruits within $\frac{1}{2}$ standard deviations of the mean will have no restrictions on duties.
 - What percentage of the incoming recruits will be trained to operate armored combat vehicles (tanks)?
 - About what percentage of the recruits will have no restrictions on training/duties?

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Standard Normal Curve – Table differences

- There are different tables and computer packages for representing the area under the **standard normal curve**. But the results are always **interchangeable**.

Area under Normal curve on $[-z : z]$

Z	Area
0.50	38.29
1.0	68.27

Area under Normal curve on $[-\infty : z]$

Z	Area
0.50	69.15
1.0	84.13

Area under Normal curve on $[z : \infty]$

Z	Area
0.50	30.85
1.0	15.87

Area under Normal curve on $[0 : z]$

Z	Area
0.50	30.85
1.0	15.87

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0.50	38.29
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Area under Normal curve on $[-\infty : z]$

Z	Area
0.50	69.15
1.0	84.13

Area under Normal curve on $[z : \infty]$

Z	Area
0.50	30.85
1.0	15.87

Area under Normal curve on $[0 : z]$

Z	Area
0.50	30.85
1.0	15.87

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