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ZNZ MD/PhD Neuroscience Course, Module BIO628  
Thu 01.06.2017

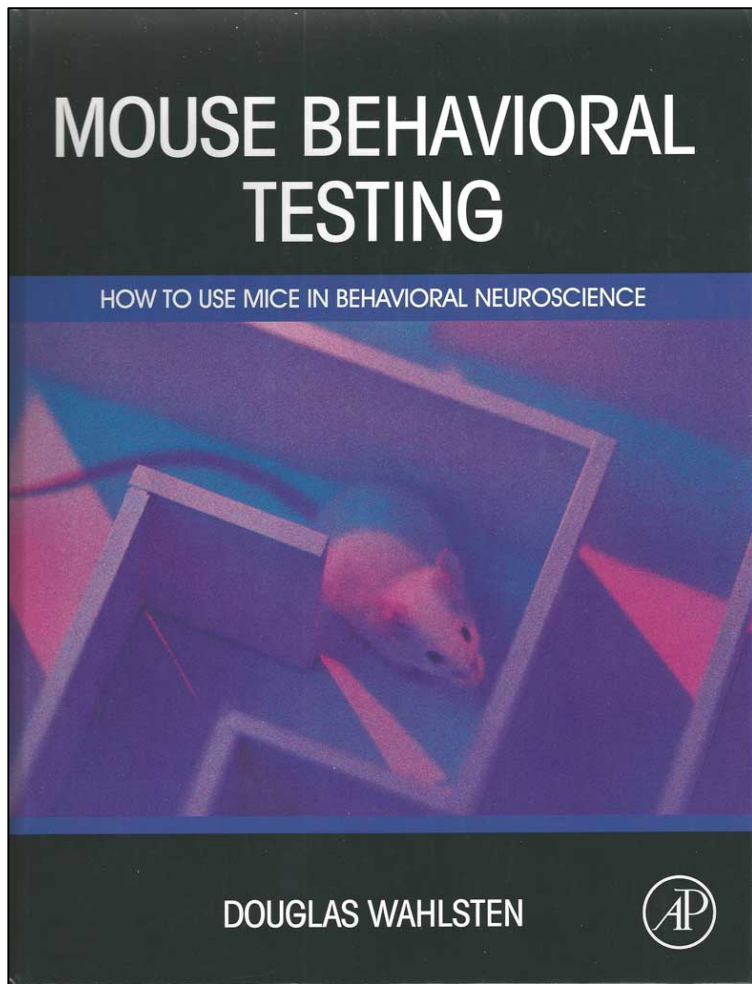
# Data analysis and presentations: Examples from basic statistics

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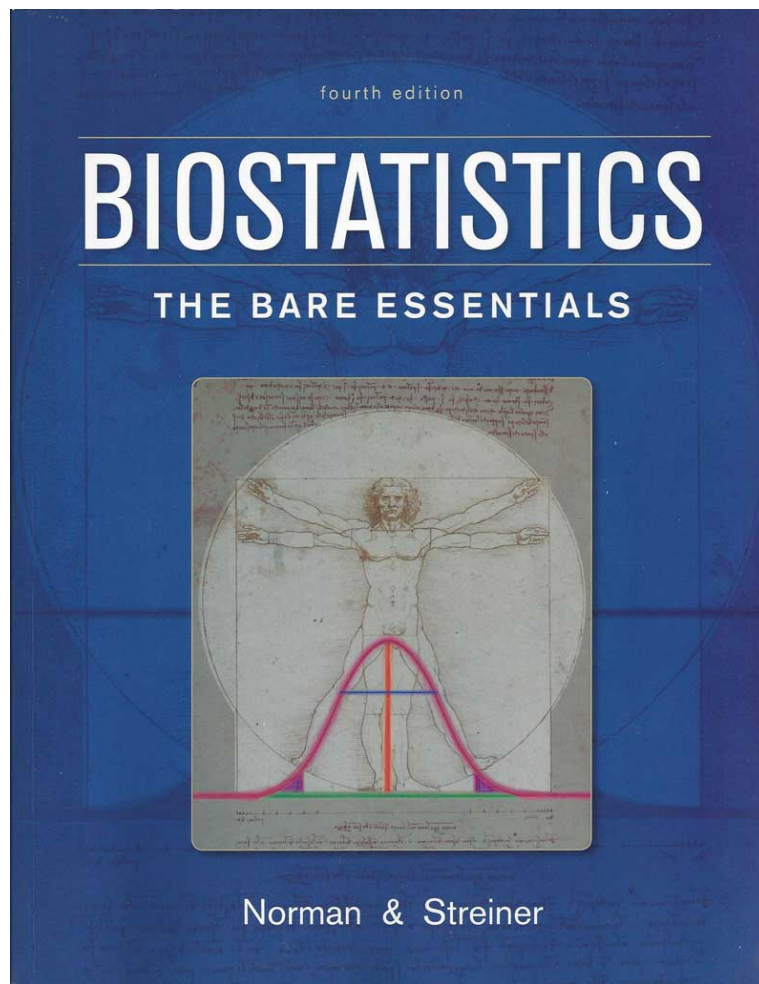
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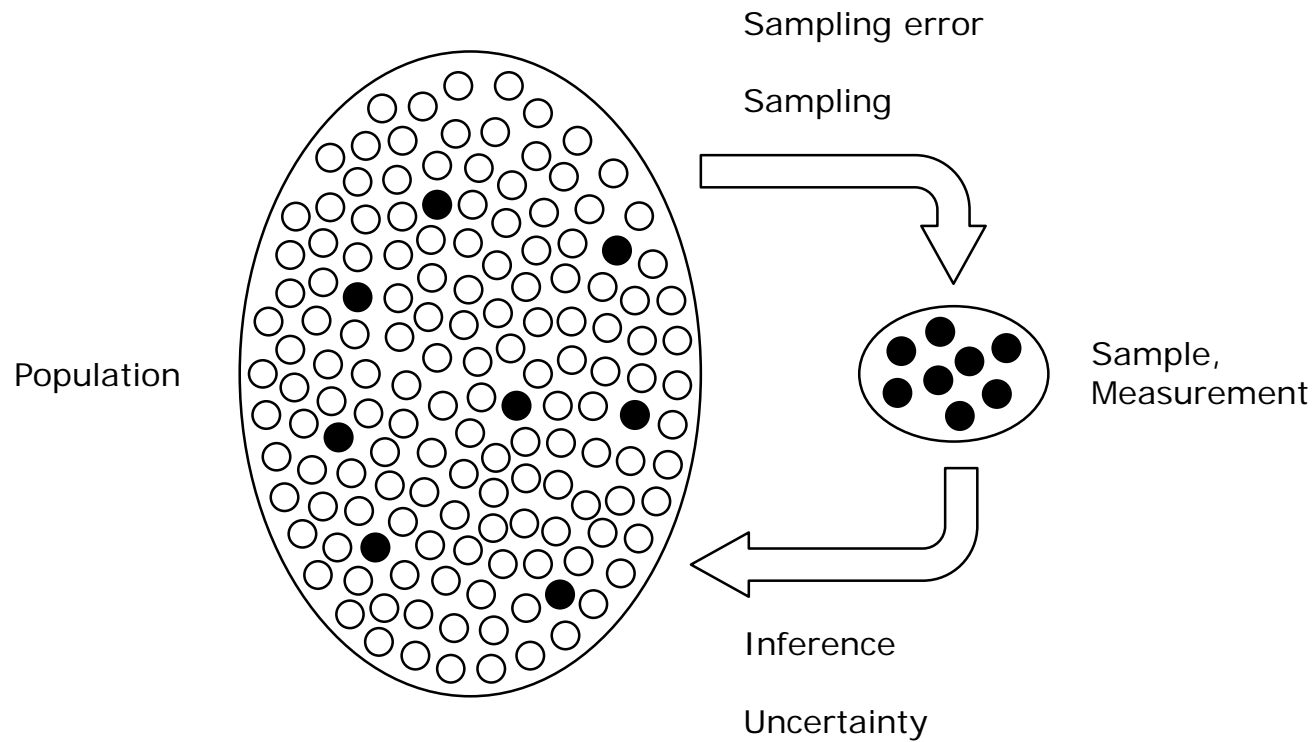
Wahlsten D  
Mouse Behavioral Testing  
Academic Press, 1. edition, 2011



Norman GR, Streiner DL  
Biostatistics, the bare essentials  
BC Decker, 4. edition, 2014

# Sampling

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# Mean, variance and standard deviation

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Sample

Population

$$M = \frac{\sum (X)}{n}$$

mean

$$\mu = \frac{\sum (X)}{N}$$

$$S^2 = \frac{\sum (X-M)^2}{n-1}$$

variance

$$\sigma^2 = \frac{\sum (X-\mu)^2}{N}$$

S

standard deviation

$\sigma$

$$\frac{S}{\sqrt{n}}$$

standard error of mean

# Effect size

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true  
effect size,  
population

estimated  
effect size,  
sample

t-test

$$\delta = \Delta\mu/\sigma$$

0.2 = small  
0.5 = medium  
0.8 = large

$$d = \Delta M/S_{\text{pooled}}$$

ANOVA

$$\omega^2 = \frac{\sigma^2 \text{ between groups}}{\sigma^2 \text{ total}}$$

5% = small  
20% = large

$$\eta^2 = \frac{S^2 \text{ between groups}}{S^2 \text{ total}}$$

# Hypothesis testing 1

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	Ho true	Ho false
reject Ho	$\alpha$ P(reject Ho   Ho true) Type I error false positive	$1-\beta$ P(reject Ho   Ho false) power true positive
retain Ho	$1-\alpha$ P(retain Ho   Ho true) true negative	$\beta$ P(retain Ho   Ho false) Type II error false negative

## Type-I error

- important if we **reject** Ho
- calculated from experimental samples by statistical tests

## Type-II error

- important if we **retain** Ho
- depends on:
  - type-I-error threshold
  - effect size, sample size

# Sample size for t-test

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	<b>2 groups: n &amp; Power to detect <math>\delta &gt; 0</math> when Null is <math>\delta = 0</math></b>												
2	Using Wahlsten's (1991) eqn (5) $n = 2C_{\alpha,\beta} / \delta^2 + 2$												
3					$\delta =$	<b>1.000</b>					<b>Note: <math>\delta = (\mu_1 - \mu_2) / \sigma</math></b>		
4	<b>For one-tailed test of null hypothesis using t test</b>						<b>General result</b>						
5	Type I error ( $\alpha$ )	Type II error ( $\beta$ )	Power (%)	$Z_{\alpha}$ 1-tail	$Z_{1-\beta}$	$2C_{\alpha,\beta}$	n per group	Rounded up				To do a quick calculation of effect size from means:	
6	0.05	0.05	95	-1.645	1.645	21.644	23.644	<b>24</b>				$\mu_1 =$	<b>15.000</b>
7	0.05	0.1	90	-1.645	1.282	17.128	19.128	<b>20</b>				$\mu_2 =$	<b>20.000</b>
8	0.05	0.2	80	-1.645	0.842	12.365	14.365	<b>15</b>				$\sigma =$	<b>4.500</b>
9	0.01	0.05	95	-2.326	1.645	31.541	33.541	<b>34</b>				$\delta =$	<b>-1.111</b>
10	0.01	0.1	90	-2.326	1.282	26.034	28.034	<b>29</b>					
11	0.01	0.2	80	-2.326	0.842	20.072	22.072	<b>23</b>					
12	0.001	0.05	95	-3.090	1.645	44.842	46.842	<b>47</b>					
13	0.001	0.1	90	-3.090	1.282	38.225	40.225	<b>41</b>					
14	0.001	0.2	80	-3.090	0.842	30.919	32.919	<b>33</b>					
15													
16	<b>For two-tailed test of null hypothesis using t test</b>						<b>General result</b>						
17	Type I error ( $\alpha$ )	Type II error ( $\beta$ )	Power (%)	$Z_{\alpha/2}$ 2-tail	$Z_{1-\beta}$	$2C_{\alpha,\beta}$	n per group	Rounded up					
18	0.05	0.05	95	-1.960	1.645	25.989	27.989	<b>28</b>					
19	0.05	0.1	90	-1.960	1.282	21.015	23.015	<b>24</b>					
20	0.05	0.2	80	-1.960	0.842	15.698	17.698	<b>18</b>					
21	0.01	0.05	95	-2.576	1.645	35.628	37.628	<b>38</b>					
22	0.01	0.1	90	-2.576	1.282	29.759	31.759	<b>32</b>					
23	0.01	0.2	80	-2.576	0.842	23.358	25.358	<b>26</b>					
24	0.001	0.05	95	-3.291	1.645	48.716	50.716	<b>51</b>					
25	0.001	0.1	90	-3.291	1.282	41.808	43.808	<b>44</b>					
26	0.001	0.2	80	-3.291	0.842	34.149	36.149	<b>37</b>					
27													

Sample size 2 2x2 Jgroups P.xls

<http://www.elsevierdirect.com/v2/companion.jsp?ISBN=9780123756749>

# 3 bad consequences of low power

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- 1 high probability to miss true effects
- 2 tendency to overestimate effect size (selection effect)
- 3 lower positive predictive value

Nat Rev Neurosci 14:365, 2013  
*Power failure: why small sample size  
undermines the reliability of neuroscience*



# Hypothesis testing 2

	Ho true	Ho false
reject Ho	$\alpha$ P(reject Ho   Ho true) Type I error false positive	$1-\beta$ P(reject Ho   Ho false) power true positive
retain Ho	$1-\alpha$ P(retain Ho   Ho true) true negative	$\beta$ P(retain Ho   Ho false) Type II error false negative

P(Ho false | Ho rejected)  
positive predictive value

$$PPV = \frac{\text{true positives}}{\text{true} + \text{false positives}} = \frac{(1-\beta)R}{(1-\beta)R + \alpha}$$

$R$  = pre-study odds

## 4 caveats regarding the interpretation of $\alpha$

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1

$\alpha$  is not a measure of effect size.

2

If  $\alpha$  is larger than the rejection threshold, this is no proof that  $H_0$  is true. Under low power conditions, negative test results are inconclusive! special statistics are needed to demonstrate equivalence.

3

$\alpha$  does not tell us how likely it is that the effect is real if the test tells us to reject  $H_0$ . For this we need the positive predictive value (PPV).

4

$\alpha$  is only valid under the assumption that only one test is done to test  $H_0$ : with  $N$  tests true type-I error probability =  $1-(1-\alpha)^N$   
Example: 2 Tests at 5%  $1-0.95^2 = 9.8\%$

Nat Rev Neurosci 14:365, 2013  
*Power failure: why small sample size undermines the reliability of neuroscience*

Nature 506:150,2014  
*Statistical errors*