

# Repeated Measures (RM) ANOVAs

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October 23, 2019

# ANOVA Family

## One way ANOVA

One Categorical  
(Independent) Variable  
(AKA one factor)

One continuous  
(dependent)  
variable  
measured in  
different  
groups

One-way Repeated  
Measures:  
One continuous  
variable measured  
>2 times across time  
intervals in same  
person

Different  
participants

Same  
participants

## Factorial ANOVA

More than One  
Categorical Variable

Two  
way

Three  
way

Four  
way

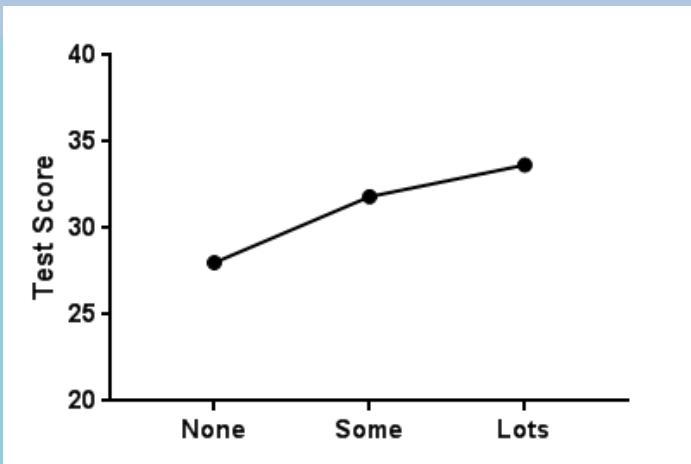
Two-way Repeated  
Measures:  
One continuous variable  
measured by two factors

# Objectives

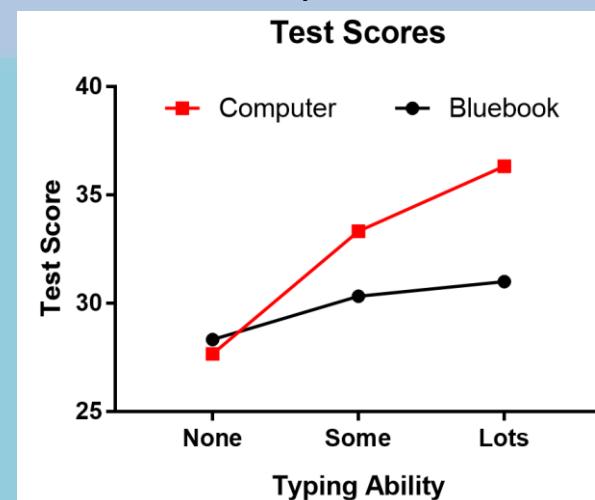
Finish Monday's lecture on the two-way ANOVA  
Understand what is meant by repeated measures  
Be able to enter data in required format  
Carry out a RM ANOVA in Prism  
Interpret the output

# What is the main question you are trying to answer about your data?

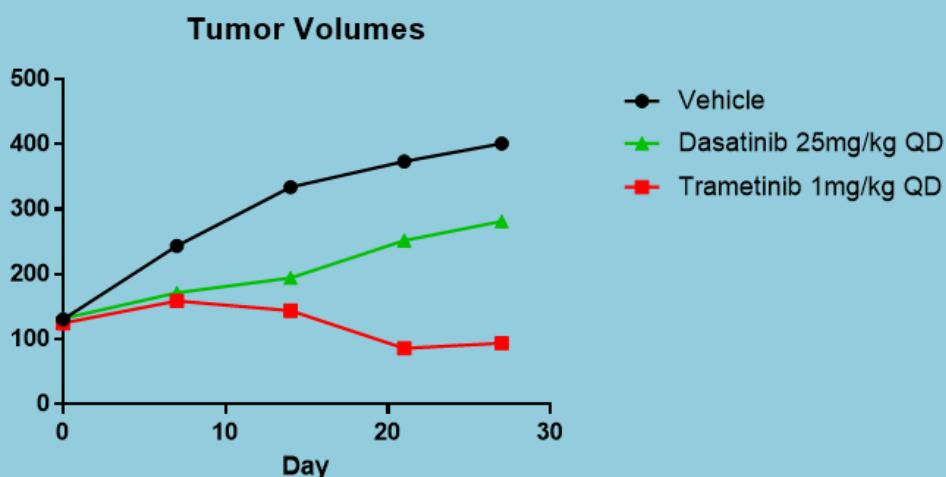
One-Way ANOVA



Two-Way ANOVA



RM ANOVA



Measuring the same variable in the same subject (cell, rat, patient) >2 times

If =2 times, use a paired t-test

## What is meant by 'Repeated Measures'?

When there are multiple measurements of the same characteristic from the same individual (or mouse, cell line, etc.)

Subject	Sex	Fed	Fasted 10 days	Fasted 20 days
1	Male	42.8	42.4	38.9
2	Male	43.1	42.2	40.3
3	Male	40.0	40.8	37.5
4	Male	46.6	45.9	42.9
5	Female	42.2	42.4	39.7
6	Female	38.7	38.1	35.8
7	Female	35.3	34.3	32.3
8	Female	40.5	40.1	37.3

Four male and four female turtles had their plasma protein measured when well fed and 10 and 20 days after fasting

## Why ‘Repeated Measures’?

Increases statistical power (*i.e.*, increases the probability of rejecting a false null hypothesis)

A subject’s repeated measurements will have less variability compared to measurements in multiple subjects

- Leads to a decrease in sampling error

- Each subject acts as their own control

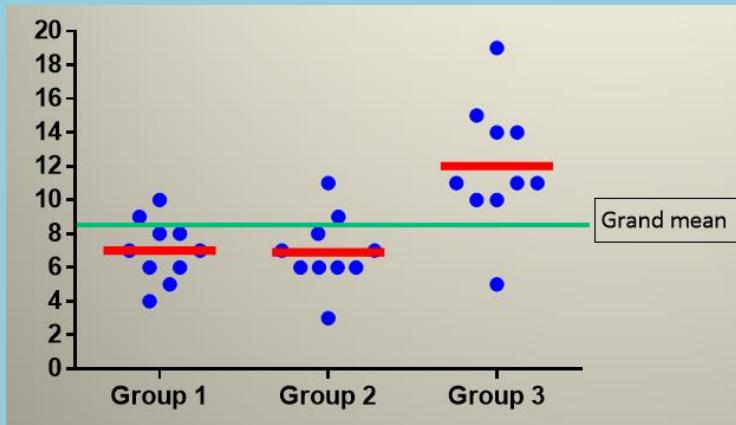
Standard one-way ANOVA assumes independence

To learn something about how the variables change over time or across situations

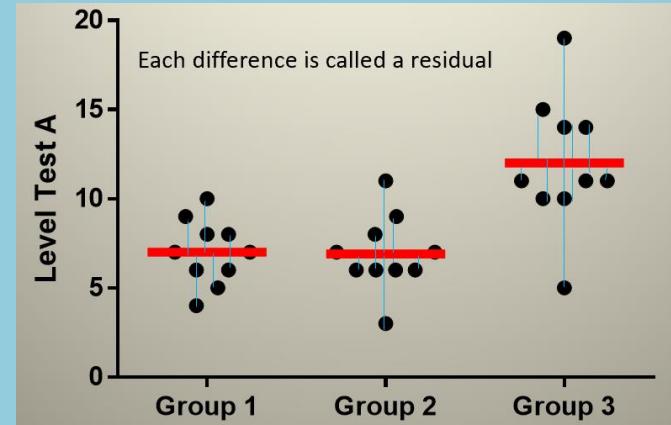
# One-way ANOVA review

One-way ANOVA partitions the total variation,  
i.e. total sum of squares, into 2 parts

SS\_between



SS\_within (SS\_Residuals)



$$F = \frac{\frac{SS_{\text{between}}}{df (\# \text{groups} - 1)}}{\frac{SS_{\text{within}}}{df (n - \# \text{groups})}} = \frac{\text{Mean of Squares}_{\text{between}}}{\text{Mean of Squares}_{\text{within}}}$$

# What happens to F-ratio if repeated measures are used?

Numerator (SS\_between) includes:

Systematic variation due to difference in treatment

Some unsystematic random variation

Denominator (SS\_within) includes:

Variation due to different individuals

Removed by the RM ANOVA  
since we have the same  
individual within each group

Denominator for RM ANOVA is smaller than the SS\_within for a regular one-way ANOVA

When denominator decreases then F increases leading to higher chance to reject ( $p<0.05$ ) the null hypothesis of equal means.

Turtle	sex	fed	fasted 10 days	fasted 20 days	SD by turtle
1	male	42.8	42.4	38.9	2.15
2	male	43.1	42.2	40.3	1.43
3	male	40.0	40.8	37.5	1.72
4	male	46.6	45.9	42.9	1.97
5	female	42.2	42.4	39.7	1.50
6	female	38.7	38.1	35.8	1.53
7	female	35.3	34.3	32.3	1.53
8	female	40.5	40.1	37.3	1.74
	SD by group	3.4	3.4	3.2	

## Limitations of RM ANOVA

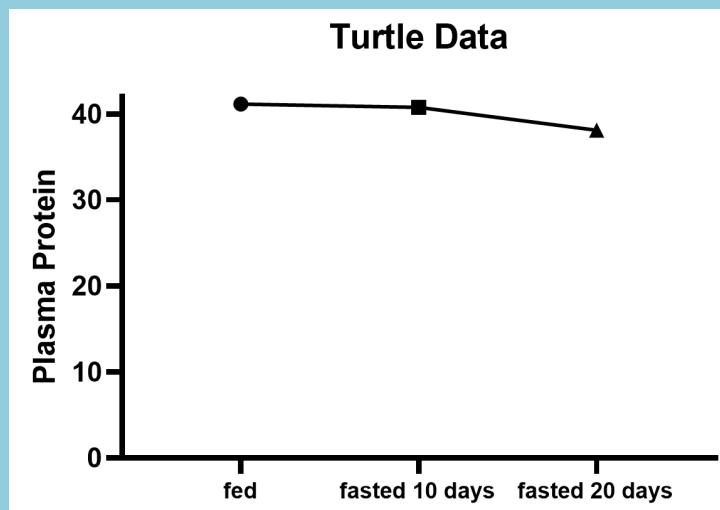
Missing data is a problem

Often tough to complete a study with no missing data.

Involves important assumption of ‘Sphericity’ that Prism cannot test.

## Example

Subject	Sex	Fed	Fasted 10 days	Fasted 20 days
1	Male	42.8	42.4	38.9
2	Male	43.1	42.2	40.3
3	Male	40.0	40.8	37.5
4	Male	46.6	45.9	42.9
5	Female	42.2	42.4	39.7
6	Female	38.7	38.1	35.8
7	Female	35.3	34.3	32.3
8	Female	40.5	40.1	37.3



Eight turtles had their plasma protein measured when well fed and 10 and 20 days after fasting

$$H_0: \mu_{Fed} = \mu_{Fast\ 10\ days} = \mu_{Fast\ 20\ days}$$

A significant finding indicates that at least one group of feeding and fasting is significantly differs from at least one other

## RM ANOVA assumptions

Dependent variable should be continuous (**plasma protein is continuous**)

Independent variable(s) should be categorical with >2 levels (**3 fasting times**)

Time intervals for repeated measures should be equal (**0, 10 20 days**)

Subjects are independent (**8 different turtles**)

Data should be normally distributed and with no outliers by group (see next slide)

Homoscedasticity by group (see next slide)

The data should show ‘sphericity’ across groups (see following slides)

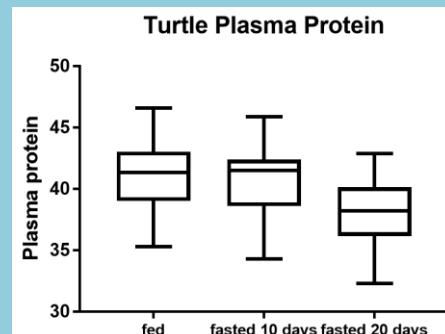
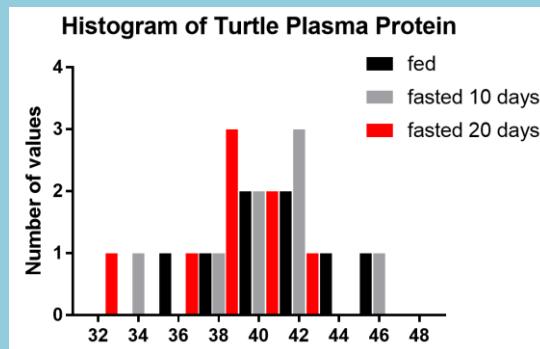
There should be no missing data (**OK**)

# Normality and Homoscedasticity

Not as important when sample size is large unless deviations from normality and homoscedasticity are large

Design has to be balanced for a RM ANOVA

The test is more robust against deviations from normality and homoscedasticity when balanced



	Col. stats	A	B	C
		fed	fasted 10 days	fasted 20 days
1	Number of values	8	8	8
2		Y	Y	Y
3	Minimum	35.3	34.3	32.3
4	25% Percentile	39.03	38.6	36.18
5	Median	41.35	41.5	38.2
6	75% Percentile	43.03	42.4	40.15
7	Maximum	46.6	45.9	42.9
8				
9	Mean	41.15	40.78	38.09
10	Std. Deviation	3.366	3.446	3.185
11	Std. Error of Mean	1.19	1.218	1.126
12				
13	Skewness	-0.2086	-0.6743	-0.4687
14	Kurtosis	0.7489	1.221	0.822
15				

## Missing data

Prism does not allow missing data for RM ANOVA

If you have any missing data, you should impute it.

Use classic ‘last observation carried forward’ strategy for simplicity, but other more complicated imputation strategies may be more appropriate (see your friendly statistician)

Can use a mixed models approach (again see your friendly statistician)

## RM ANOVA Key Assumption: Sphericity

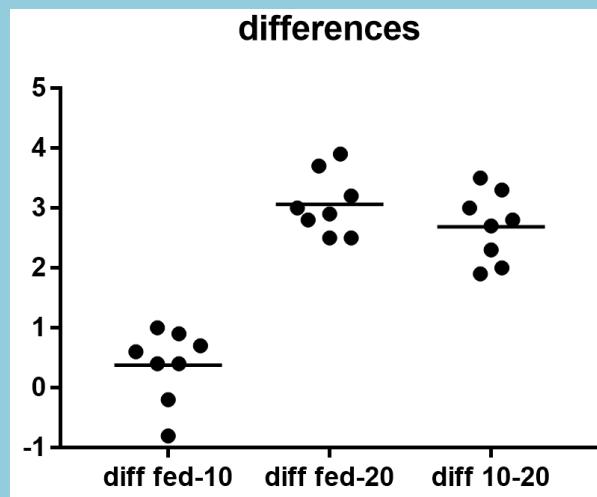
Refers to the equality of variances of the *differences* between the various time groups

Variances of the difference must be the same for each pair of time points

$$\text{Variance}_{t1-t2} \approx \text{Variance}_{t1-t3} \approx \text{Variance}_{t2-t3}$$

Correlation between pairs of time point groups is the same

$$\text{corr}_{t1 \text{ to } t2} \approx \text{corr}_{t2 \text{ to } t3} \approx \text{corr}_{t1 \text{ to } t3}$$



Sphericity (denoted by  $\varepsilon$ —epsilon - and sometimes referred to as *circularity*) is a type of *compound symmetry*.

$\varepsilon$  indicates the degree to which sphericity has been violated

# Sphericity

The violation of sphericity is serious for a RM ANOVA model

Causes the test to become too prone to Type I error (false positive)

Even more important when sample size is small

Determining if sphericity is violated is important

If a violation occurs, corrections can be made to produce a more valid critical  $F$ -value (i.e., reduce the increase in Type I error rate).

Estimating the degree to which sphericity has been violated and apply a correction factor to the df of the  $F$ -distribution

Common corrections: Geisser-Greenhouse (Prism ) and Huynh-Feldt

## How to test sphericity?

Mauchly's Test of Sphericity is a formal way of testing the assumption of sphericity

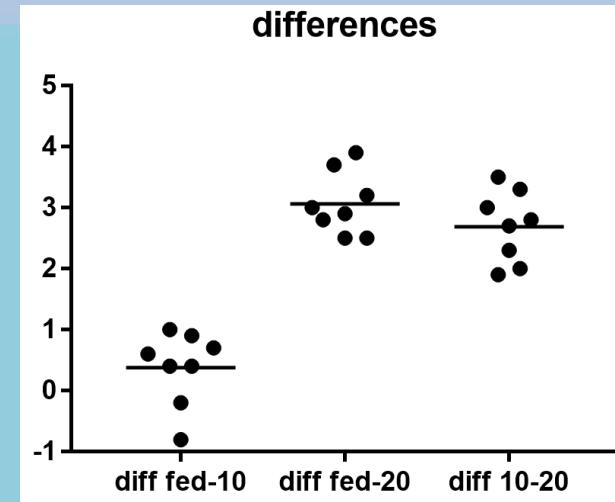
But fails to detect departures from sphericity in small samples and over-detects them in large samples

Also, Prism does not offer this test

Prism can run the test either with a correction (for lack of sphericity) or with no correction (assumes sphericity exists)

# Eyeball for sphericity assumptions : Turtle Plasma Protein Data

	Col. stats	A	B	C
		diff fed-10	diff fed-20	diff 10-20
1	Number of values	8	8	8
2		Y	Y	Y
3	Minimum	-0.8	2.5	1.9
4	25% Percentile	-0.05	2.575	2.075
5	Median	0.5	2.95	2.75
6	75% Percentile	0.85	3.575	3.225
7	Maximum	1	3.9	3.5
8				
9	Mean	0.375	3.063	2.688
10	Std. Deviation	0.6018	0.5153	0.5842
11	Std. Error of Mean	0.2128	0.1822	0.2065
12				



Compound symmetry and sphericity: Look at the correlations:

	Fed	Fasted 10	Fasted 20
Fed	1	0.985	0.989
Fasted 10	0.985	1	0.988
Fasted 20	0.989	0.988	1

Compound symmetry holds if correlations in the blue cells (off diagonal) are similar

Compound symmetry implies sphericity (reverse not true)

## An example without compound symmetry

	time1	time2	time3	time4
time1	1.00000	0.94035	-0.14150	0.28445
time2	0.94035	1.00000	-0.02819	0.26921
time3	-0.14150	-0.02819	1.00000	0.27844
time4	0.28445	0.26921	0.27844	1.00000

Certainly do NOT have equal correlations. Time 1 and time 2 are highly correlated ( $r=0.94$ ), but time 1 and time 3 are *inversely* correlated.

CONCLUSION: Probably no sphericity

When in doubt, assume sphericity assumption is violated and run the test with the correction

## Other Things to Consider in a RM ANOVA

### Spacing of time intervals

RM ANOVA require that all subjects are measured at equal time intervals

i.e.,      days 0, 5, 10, 15, 20  
NOT      days 0 10 15 21 30

### Missing Data

Prism will not perform a RM ANOVA if data are missing

# What happens if you don't check the assumptions?

Prism will fail to run an analysis if there is missing data

Inaccurate results

Increased Type-I error – false positive (e.g. if sphericity is assumed but not satisfied)

Decreased power (i.e. increased Type-II error – false negative)

Journals won't accept the paper

Statistics Instructor will be unhappy

# One-way RM ANOVA Example: Turtles, fasting and plasma protein

Use a “Columns” table

	Group A	Group B	Group C
	fed	fasted 10 days	fasted 20 days
1	42.8	42.4	38.9
2	43.1	42.2	40.3
3	40.0	40.8	37.5
4	46.6	45.9	42.9
5	42.2	42.4	39.7
6	38.7	38.1	35.8
7	35.3	34.3	32.3
8	40.5	40.1	37.3
9			

Analyze Data

Built-in analysis

Which analysis?

- Transform, Normalize...**
  - Transform
  - Transform concentrations (X)
  - Normalize
  - Prune rows
  - Remove baseline and column math
  - Transpose X and Y
  - Fraction of total
- XY analyses**
- Column analyses**
  - t tests (and nonparametric tests)
  - One-way ANOVA (and nonparametric or mixed)**
  - One sample t and Wilcoxon test
  - Descriptive statistics
  - Normality and Lognormality Tests
  - Frequency distribution
  - ROC Curve
  - Bland-Altman method comparison
  - Identify outliers
  - Analyze a stack of P values
- Grouped analyses**
- Contingency table analyses**
- Survival analyses**
- Parts of whole analyses**

Analyze which data sets?

- A:fed
- B:fasted 10 days
- C:fasted 20 days

Select All   Deselect All

Help   Cancel   OK

# One-way RM ANOVA using prism: Not assuming sphericity (Prism recommended)

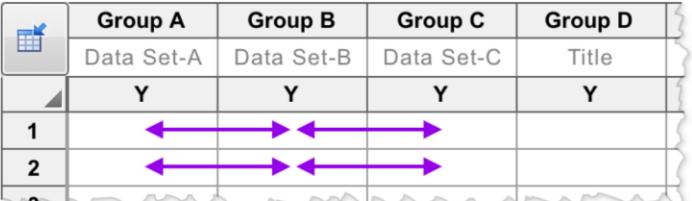
Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design   Repeated Measures   Multiple Comparisons   Options   Residuals

**Experimental design**

No matching or pairing

Each row represents matched, or repeated measures, data



**Assume Gaussian distribution of residuals?**

Yes. Use ANOVA.

No. Use nonparametric test.

**Assume sphericity (equal variability of differences)?**

Yes. No correction.

No. Use the Geisser-Greenhouse correction. Recommended.

Based on your choices (on all tabs), Prism will perform:

- RM one-way ANOVA, with the Geisser-Greenhouse correction.

Learn   Cancel   OK

Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design   Repeated Measures   Multiple Comparisons   Options   Residuals

Analyses of repeated measures data can be reported in two ways.

- ANOVA (partition sum-of-squares). This is the same as the general linear model (GLM).
- Mixed-effects model. This uses the restricted maximum likelihood method.

If there are no missing values, the two approaches give the same main results (F and P values). But the methods are very different, so the other reported results differ.

**Analyze using which method**

Repeated measures ANOVA (based on GLM).  
Same as Prism 7 and earlier.  
Requires balanced data (no missing values).

Mixed-effects model.  
Results are presented in a format different than ANOVA.  
Works fine with missing values.

It depends.  
Use ANOVA if there are no missing values.  
Use mixed-effects model if there are missing values.

**What to do if a random effect is zero (or negative)?**

Remove term(s) from model and fit a simpler model (recommended).

Fit the full model anyway (corresponds to NOBOUND parameter in SAS).

Make these choices the default for future ANOVAs (One-, Two- and Three-way).

Learn   Cancel   OK

Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design   Repeated Measures   Multiple Comparisons   Options   Residuals

### Followup tests

- None.
- Compare the mean of each column with the mean of every other column.
- Compare the mean of each column with the mean of a control column.  
Control column: Column A : fed
- Compare the means of preselected pairs of columns.  
Selected pairs:
- Test for linear trend between column mean and left-to-right column order.

### Which test?

Use choices on the Options tab to choose the test, and to set the defaults for future ANOVAs.

Learn   Cancel   OK

Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design   Repeated Measures   Multiple Comparisons   Options   Residuals

### Multiple comparisons test

- Correct for multiple comparisons using statistical hypothesis testing. Recommended.  
Test: Tukey (recommended)
- Correct for multiple comparisons by controlling the False Discovery Rate.  
Test: Two-stage step-up method of Benjamini, Krieger and Yekutieli (recommended)
- Don't correct for multiple comparisons. Each comparison stands alone.  
Test: Fisher's LSD test

### Multiple comparisons options

- Swap direction of comparisons (A-B) vs. (B-A).
- Report multiplicity adjusted P value for each comparison.  
Each P value is adjusted to account for multiple comparisons.

Family-wise significance and confidence level: 0.05 (95% confidence interval)

### Graphing

- Graph confidence intervals.
- Graph ranks (nonparametric).
- Graph differences (repeated measures).

### Additional results

- Descriptive statistics for each data set.
- Report comparison of models using AICc.
- Report goodness of fit.

### Output

Show this many significant digits (for everything except P values): 4

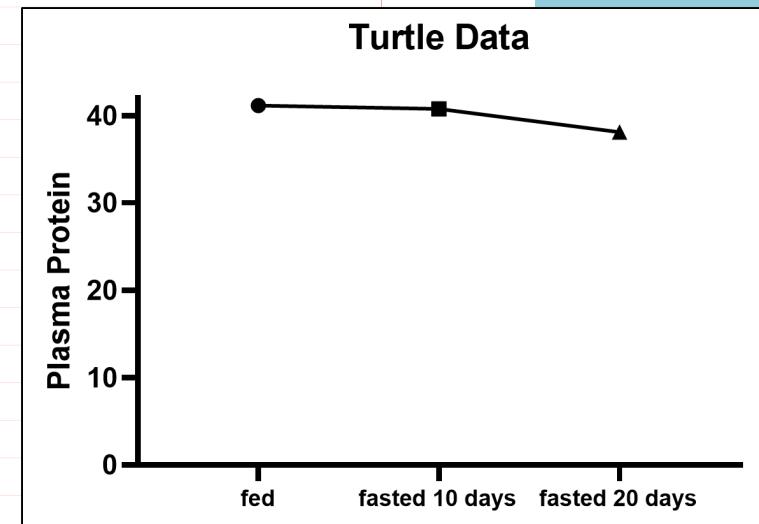
P value style: GP: 0.1234 (ns), 0.0332 (\*), 0.0021 (\*\*), 0.00C N = 6

Make options on this tab be the default for future One-Way ANOVAs.

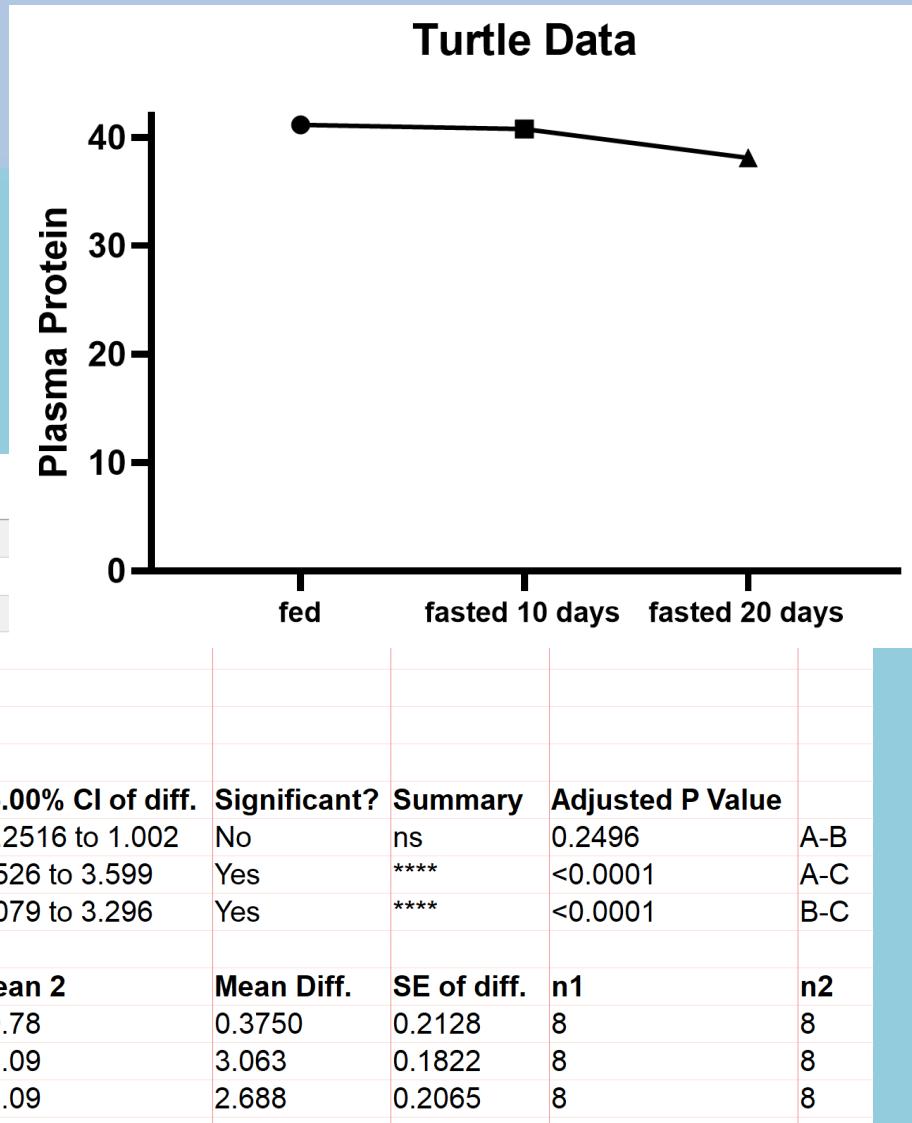
Learn   Cancel   OK

	Group A	Group B	Group C
	fed	fasted 10 days	fasted 20 days
	Y	Y	Y
1	42.8	42.4	38.9
2	43.1	42.2	40.3
3	40.0	40.8	37.5
4	46.6	45.9	42.9
5	42.2	42.4	39.7
6	38.7	38.1	35.8
7	35.3	34.3	32.3
8	40.5	40.1	37.3
9			

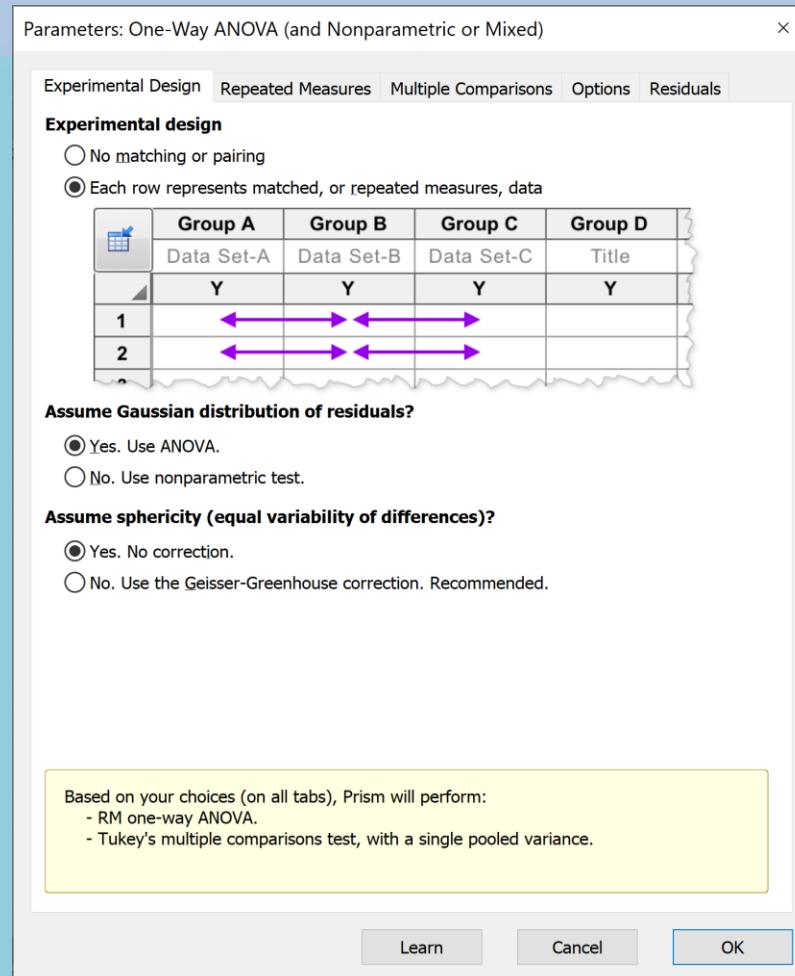
	ANOVA results	Multiple comparisons				
<b>RM one-way ANOVA</b>						
ANOVA results						
1	Table Analyzed	Turtle Data				
2						
3	<b>Repeated measures ANOVA summary</b>					
4	Assume sphericity?	No				
5	F	138.2				
6	P value	<0.0001				
7	P value summary	****				
8	Statistically significant ( $P < 0.05$ )?	Yes				
9	Geisser-Greenhouse's epsilon	0.9680				
10	R square	0.9518				
11						
12	<b>Was the matching effective?</b>					
13	F	204.5				
14	P value	<0.0001				
15	P value summary	****				
16	Is there significant matching ( $P < 0.05$ )?	Yes				
17	R square	0.8313				
18						
19	<b>ANOVA table</b>					
20	Treatment (between columns)	SS 44.65	DF 2	MS 22.32	F (DFn, DFd) $F(1.936, 13.55) = 138.2$	P value $P < 0.0001$
21	Individual (between rows)	231.2	7	33.02	$F(7, 14) = 204.5$	$P < 0.0001$
22	Residual (random)	2.261	14	0.1615		
23	Total	278.1	23			



	ANOVA results		Multiple comparisons			
	RM one-way ANOVA		Multiple comparisons			
1	Number of families	1				
2	Number of comparisons per family	3				
3	Alpha	0.05				
4						
5	<b>Tukey's multiple comparisons test</b>	<b>Mean Diff.</b>	<b>95.00% CI of diff.</b>	<b>Significant?</b>	<b>Summary</b>	<b>Adjusted P Value</b>
6	fed vs. fasted 10 days	0.3750	-0.2516 to 1.002	No	ns	0.2496
7	fed vs. fasted 20 days	3.063	2.526 to 3.599	Yes	****	<0.0001
8	fasted 10 days vs. fasted 20 days	2.688	2.079 to 3.296	Yes	****	<0.0001
9						
10	<b>Test details</b>	<b>Mean 1</b>	<b>Mean 2</b>	<b>Mean Diff.</b>	<b>SE of diff.</b>	<b>n1</b>
11	fed vs. fasted 10 days	41.15	40.78	0.3750	0.2128	8
12	fed vs. fasted 20 days	41.15	38.09	3.063	0.1822	8
13	fasted 10 days vs. fasted 20 days	40.78	38.09	2.688	0.2065	8



# Now assuming sphericity: what is different?



RM one-way ANOVA ANOVA results	
1	Table Analyzed
2	Turtle Data
3	<b>Repeated measures ANOVA summary</b>
4	Assume sphericity?
5	No
6	F 138.2
7	P value <0.0001
8	P value summary ****
9	Statistically significant ( $P < 0.05$ )? Yes
10	Geisser-Greenhouse's epsilon 0.9680
11	R square 0.9518
12	<b>Was the matching effective?</b>
13	F 204.5
14	P value <0.0001
15	P value summary ****
16	Is there significant matching ( $P < 0.05$ )? Yes
17	R square 0.8313
18	
19	<b>ANOVA table</b>
20	Treatment (between columns) SS 44.65 DF 2 MS 22.32 F (DFn, DFd) $F(1.936, 13.55) = 138.2$ P value $P < 0.0001$
21	Individual (between rows) 231.2 7 33.02 $F(7, 14) = 204.5$ $P < 0.0001$
22	Residual (random) 2.261 14 0.1615
23	Total 278.1 23

RM one-way ANOVA ANOVA results	
1	Table Analyzed
2	Turtle Data
3	<b>Repeated measures ANOVA summary</b>
4	Assume sphericity? Yes
5	F 138.2
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7	P value summary ****
8	Statistically significant ( $P < 0.05$ )? Yes
9	R square 0.9518
10	
11	<b>Was the matching effective?</b>
12	F 204.5
13	P value <0.0001
14	P value summary ****
15	Is there significant matching ( $P < 0.05$ )? Yes
16	R square 0.8313
17	
18	<b>ANOVA table</b>
19	Treatment (between columns) SS 44.65 DF 2 MS 22.32 F (DFn, DFd) $F(2, 14) = 138.2$ P value $P < 0.0001$
20	Individual (between rows) 231.2 7 33.02 $F(7, 14) = 204.5$ $P < 0.0001$
21	Residual (random) 2.261 14 0.1615
22	Total 278.1 23

Decrease the df and the critical value increases;  
 F statistics must be larger to reach  $p \leq 0.05$

		F - Distribution ( $\alpha = 0.05$ in the Right Tail)								
Denominator Degrees of Freedom <i>df<sub>2</sub></i>	Numerator Degrees of Freedom <i>df<sub>1</sub></i>	Numerator Degrees of Freedom								
		1	2	3	4	5	6	7	8	9
1	161.45	199.50	215.71	224.58	230.16	233.99	236.77	238.88	240.54	
2	18.513	19.000	19.164	19.247	19.296	19.330	19.353	19.371	19.385	
3	10.128	9.5521	9.2766	9.1172	9.0135	8.9406	8.8867	8.8452	8.8123	
4	7.7086	9.9443	6.5914	6.3882	6.2561	6.1631	6.0942	6.0410	6.9988	
5	6.6079	5.7861	5.4095	5.1922	5.0503	4.9503	4.8759	4.8183	4.7725	
6	5.9874	5.1433	4.7571	4.5337	4.3874	4.2839	4.2067	4.1468	4.0990	
7	5.5914	4.7374	4.3468	4.1203	3.9715	3.8660	3.7870	3.7257	3.6767	
8	5.3177	4.4590	4.0662	3.8379	3.6875	3.5806	3.5005	3.4381	3.3881	
9	5.1174	4.2565	3.8625	3.6331	3.4817	3.3738	3.2927	3.2296	3.1789	
10	4.9646	4.1028	3.7083	3.4780	3.3258	3.2172	3.1355	3.0717	3.0204	
11	4.8443	3.9823	3.5874	3.3567	3.2039	3.0946	3.0123	2.9480	2.8962	
12	4.7472	3.8853	3.4903	3.2592	3.1059	2.9961	2.9134	2.8486	2.7964	
13	4.6672	3.8056	3.4105	3.1791	3.0254	2.9153	2.8321	2.7669	2.7144	
14	4.6001	3.7389	3.3439	3.1122	2.9582	2.8477	2.7642	2.6987	2.6458	
15	4.5431	3.6823	3.2874	3.0556	2.9013	2.7905	2.7066	2.6408	2.5876	
16	4.4940	3.6337	3.2389	3.0069	2.8524	2.7413	2.6572	2.5911	2.5377	
17	4.4513	3.5915	3.1968	2.9647	2.8100	2.6987	2.6143	2.5480	2.4943	
18	4.4139	3.5546	3.1599	2.9277	2.7729	2.6613	2.5767	2.5102	2.4563	
19	4.3807	3.5219	3.1274	2.8951	2.7401	2.6283	2.5435	2.4768	2.4227	
20	4.3512	3.4928	3.0984	2.8661	2.7109	2.5990	2.5140	2.4471	2.3928	
21	4.3248	3.4668	3.0725	2.8401	2.6848	2.5727	2.4876	2.4205	2.3660	
22	4.3009	3.4434	3.0491	2.8167	2.6613	2.5491	2.4638	2.3965	2.3419	
23	4.2793	3.4221	3.0280	2.7955	2.6400	2.5277	2.4422	2.3748	2.3201	
24	4.2597	3.4028	3.0088	2.7763	2.6207	2.5082	2.4226	2.3551	2.3002	
25	4.2417	3.3852	2.9912	2.7587	2.6030	2.4904	2.4047	2.3371	2.2821	
26	4.2252	3.3690	2.9752	2.7426	2.5868	2.4741	2.3883	2.3205	2.2655	
27	4.2100	3.3541	2.9604	2.7278	2.5719	2.4591	2.3732	2.3053	2.2501	
28	4.1960	3.3404	2.9467	2.7141	2.5581	2.4453	2.3593	2.2913	2.2360	
29	4.1830	3.3277	2.9340	2.7014	2.5454	2.4324	2.3463	2.2783	2.2229	
30	4.1709	3.3158	2.9223	2.6896	2.5336	2.4205	2.3343	2.2662	2.2107	
40	4.0847	3.2317	2.8387	2.6060	2.4495	2.3359	2.2490	2.1802	2.1240	
60	4.0012	3.1504	2.7581	2.5252	2.3683	2.2541	2.1665	2.0970	2.0401	
120	3.9201	3.0718	2.6802	2.4472	2.2899	2.1750	2.0868	2.0164	1.9588	
$\infty$	3.8415	2.9957	2.6049	2.3719	2.2141	2.0986	2.0096	1.9384	1.8799	

Let's now do a standard one-way ANOVA  
We will pretend the data are not paired

Notice the difference

### One-way ANOVA

ANOVA table	SS	DF	MS	F (DFn, DFd)	P value
Treatment (between columns)	44.65	2	22.32	F (2, 21) = 2.008	P=0.1592
Residual (within columns)	233.4	21	11.12		
Total	278.1	23			

### One-way RM ANOVA

ANOVA table	SS	DF	MS	F (DFn, DFd)	P value
Treatment (between columns)	44.65	2	22.32	F (2, 14) = 138.2	P<0.0001
Individual (between rows)	231.2	7	33.02	F (7, 14) = 204.5	P<0.0001
Residual (random)	2.261	14	0.1615		
Total	278.1	23			

The residual SS is much smaller for RM ANOVA  
The smaller the residual SS, the larger the F-statistic

# Another data format

New Data Table and Graph

**New table & graph**

- XY
- Column
- Grouped
- Contingency
- Survival
- Parts of whole
- Multiple variables
- Nested

**Existing file**

Clone a graph

Prism Tips

**Column tables have one grouping variable, with each group defined by a column**

	A	B
Control		Treated
Y		Y
1		
2		

[Learn more](#)

**Data table:**

- Enter or import data into a new table
- Start with sample data to follow a tutorial

**Options:**

- Enter replicate values, stacked into columns
- Enter paired or repeated measures data - each subject on a separate row
- Enter and plot error values already calculated elsewhere

Enter: Mean, SD, N

Table format:		Group A	Group B	Group C
Column		fed	fasted 10 days	fasted 20 days
1	1	42.8	42.4	38.9
2	2	43.1	42.2	40.3
3	3	40.0	40.8	37.5
4	4	46.6	45.9	42.9
5	5	42.2	42.4	39.7
6	6	38.7	38.1	35.8
7	7	35.3	34.3	32.3
8	8	40.5	40.1	37.3
9				

## Analyze Data

## Built-in analysis

## Which analysis?

 Transform, Normalize...

Transform

Transform concentrations (X)

Normalize

Prune rows

Remove baseline and column math

Transpose X and Y

Fraction of total

 XY analyses Column analyses

t tests (and nonparametric tests)

 One-way ANOVA (and nonparametric or mixed)

One sample t and Wilcoxon test

Descriptive statistics

Normality and Lognormality Tests

Frequency distribution

ROC Curve

Bland-Altman method comparison

Identify outliers

Analyze a stack of P values

 Grouped analyses Contingency table analyses Survival analyses Parts of whole analyses

## Analyze which data sets?

- A:fed  
 B:fasted 10 days  
 C:fasted 20 days

Select All

Deselect All

## Parameters: One-Way ANOVA (and Nonparametric or Mixed)

## Experimental Design

## Repeated Measures

## Multiple Comparisons

## Options

## Residuals

## Experimental design

 No matching or pairing Each row represents matched, or repeated measures, data

	Group A	Group B	Group C	Group D	
	Data Set-A	Data Set-B	Data Set-C	Title	
1	Y	Y			
2		Y	Y		
3			Y	Y	

Automatically chooses paired

## Assume Gaussian distribution of residuals?

 Yes. Use ANOVA. No. Use nonparametric test.

## Assume sphericity (equal variability of differences)?

 Yes. No correction. No. Use the Geisser-Greenhouse correction. Recommended.

Based on your choices (on all tabs), Prism will perform:  
 - RM one-way ANOVA, with the Geisser-Greenhouse correction.

Learn

Cancel

OK

# Post Hoc tests

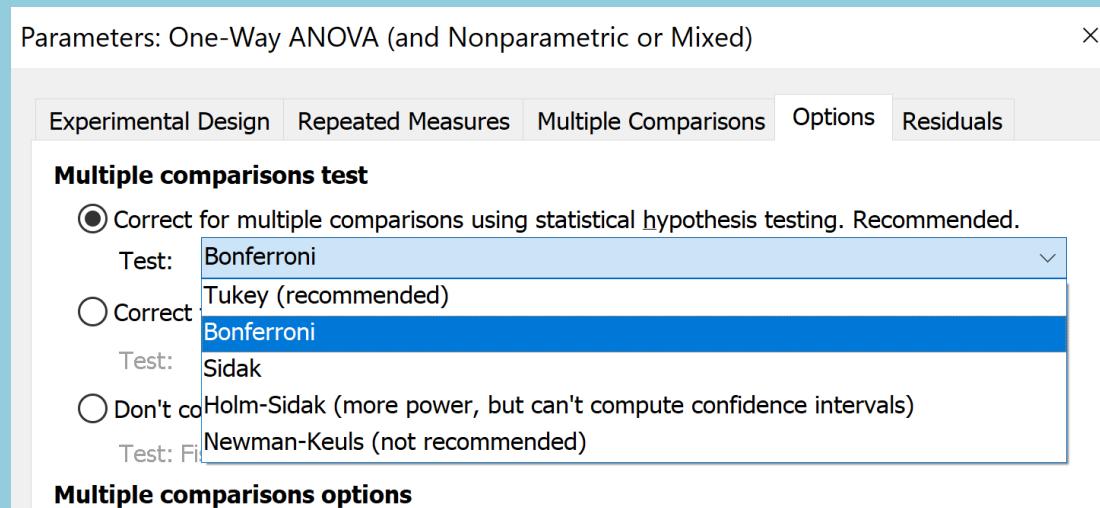
Global RM ANOVA  $\leq 0.05$

Need Post-Hoc Multiple Comparisons  
which groups are different?

If there is sphericity, use standard methods (e.g., Tukey's)

Prism will recommend Tukey's with or without using the correction

Without sphericity better to use conservative approach: Bonferroni test from drop down menu of choice of tests for multiple comparisons



## Post hoc tests in Prism

### When sphericity is assumed

Tukey's multiple comparisons test	Mean Diff.	95.00% CI of diff.	Significant?	Summary	Adjusted P Value	
fed vs. fasted 10 days	0.375	-0.1509 to 0.9009	No	ns	0.1850	A-B
fed vs. fasted 20 days	3.063	2.537 to 3.588	Yes	****	<0.0001	A-C
fasted 10 days vs. fasted 20 days	2.688	2.162 to 3.213	Yes	****	<0.0001	B-C

### When sphericity is not assumed

Tukey's multiple comparisons test	Mean Diff.	95.00% CI of diff.	Significant?	Summary	Adjusted P Value	
fed vs. fasted 10 days	0.375	-0.2516 to 1.002	No	ns	0.2496	A-B
fed vs. fasted 20 days	3.063	2.526 to 3.599	Yes	****	<0.0001	A-C
fasted 10 days vs. fasted 20 days	2.688	2.079 to 3.296	Yes	****	<0.0001	B-C

## Write up results (we assume sphericity is present)

The data passed the assumptions for a one-way RM ANOVA (no missing data, independent subjects, normality, homoscedasticity, sphericity, equal time points). With  $p < 0.0001$  (two-tailed hypothesis  $F(2,14)=138.2$ ,  $\alpha=0.05$ ), we reject the null hypothesis of no difference in population means by group and conclude there is at least one group different from one other. Tukey post-hoc tests showed that 'Fasted 20' (mean=38.1) was significantly different from 'Fed' (41.2) and 'Fasted 10' (40.8). There was no significant difference between 'Fed' and 'Fasted 10'. We conclude that plasma protein levels decrease with fasting.

	Col. stats	A	B	C
		fed	fasted 10 days	fasted 20 days
1	Number of values	8	8	8
2				
3	Minimum	35.3	34.3	32.3
4	25% Percentile	39.03	38.6	36.18
5	Median	41.35	41.5	38.2
6	75% Percentile	43.03	42.4	40.15
7	Maximum	46.6	45.9	42.9
8				
9	Mean	41.15	40.78	38.09
10	Std. Deviation	3.366	3.446	3.185
11	Std. Error of Mean	1.19	1.218	1.126

## How you might write it if sphericity is not present

The data passed most of the assumptions for a one-way RM ANOVA (no missing data, independent subjects, normality, homoscedasticity by groups, equal time points), except for sphericity. Because of this we will use the one-way RM ANOVA with the Greenhouse-Geisser correction. With  $p < 0.0001$  etc...

## Other options if assumptions for RM ANOVA are not satisfied

Use mixed models

Mixed models are advanced modelling methods (New in Prism 8)

Can deal with missing values, uneven time points, and lack of sphericity

Otherwise use the non-parametric Friedman Test

# Non-parametric alternative to the one-way RM ANOVA: Friedman Test

Analyze Data

Built-in analysis

Which analysis?

- Transform, Normalize...**
  - Transform
  - Transform concentrations (X)
  - Normalize
  - Prune rows
  - Remove baseline and column math
  - Transpose X and Y
  - Fraction of total
- XY analyses**
- Column analyses**
  - t tests (and nonparametric tests)
  - One-way ANOVA (and nonparametric or mixed)**
  - One sample t and Wilcoxon test
  - Descriptive statistics
  - Normality and Lognormality Tests
  - Frequency distribution
  - ROC Curve
  - Bland-Altman method comparison
  - Identify outliers
  - Analyze a stack of P values
- Grouped analyses**
- Contingency table analyses**
- Survival analyses**
- Parts of whole analyses**

Analyze which data sets?

- A:fed
- B:fasted 10 days
- C:fasted 20 days

Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design    Repeated Measures    Multiple Comparisons    Options    Residuals

**Experimental design**

No matching or pairing  
 Each row represents matched, or repeated measures, data

	Group A	Group B	Group C	Group D
	Data Set-A	Data Set-B	Data Set-C	Title
Y	Y	Y	Y	Y
1	↔↔↔↔			
2	↔↔↔↔			

**Assume Gaussian distribution of residuals?**

Yes. Use ANOVA.  
 No. Use nonparametric test.

Based on your choices (on all tabs), Prism will perform:

- Friedman test.
- Dunn's multiple comparisons test.

Learn    Cancel    OK

Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design   Repeated Measures   Multiple Comparisons   Options   Residuals

Analyses of repeated measures data can be reported in two ways.

- ANOVA (partition sum-of-squares). This is the same as the general linear model (GLM).
- Mixed-effects model. This uses the restricted maximum likelihood method.

If there are no missing values, the two approaches give the same main results (F and P values). But the methods are very different, so the other reported results differ.

Analyze using which method

Repeated measures ANOVA (based on GLM).  
Same as Prism 7 and earlier.  
Requires balanced data (no missing values).

Mixed-effects model.  
Results are presented in a format different than ANOVA.  
Works fine with missing values.

It depends.  
Use ANOVA if there are no missing values.  
Use mixed-effects model if there are missing values.

What to do if a random effect is zero (or negative)?

Remove term(s) from model and fit a simpler model (recommended).

Fit the full model anyway (corresponds to NOBOUND parameter in SAS).

These choices are available only when you choose a repeated measures design on the first (Experimental Design) tab.

Make these choices the default for future ANOVAs (One-, Two- and Three-way).

Learn   Cancel   OK

Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design   Repeated Measures   Multiple Comparisons   Options   Residuals

**Followup tests**

None.

Compare the mean rank of each column with the mean rank of every other column.

Compare the mean rank of each column with the mean rank of a control column.  
Control column: Column A : fed

Compare the mean ranks of preselected pairs of columns.  
Selected pairs:

Test for linear trend between column mean and left-to-right column order.

**Which test?**

Use choices on the Options tab to choose the test, and to set the defaults for future ANOVAs.

Learn   Cancel   OK

## Friedman Test

ANOVA results		Multiple comparisons	
<b>Friedman test</b>			
ANOVA results			
1	Table Analyzed	Turtle Data Paired Entry	
2			
3	<b>Friedman test</b>		
4	P value	0.0003	
5	Exact or approximate P value?	Exact	
6	P value summary	***	
7	Are means signif. different? (P < 0.05)	Yes	
8	Number of groups	3	
9	Friedman statistic	13.00	
10			
11	<b>Data summary</b>		
12	Number of treatments (columns)	3	
13	Number of subjects (rows)	8	
14			

## One-way RM ANOVA

ANOVA table	SS	DF	MS	F (DFn, DFd)	P value
Treatment (between columns)	44.65	2	22.32	F (2, 14) = 138.2	P<0.0001
Individual (between rows)	231.2	7	33.02	F (7, 14) = 204.5	P<0.0001
Residual (random)	2.261	14	0.1615		
Total	278.1	23			

# Friedman Test

Friedman test					
Multiple comparisons					
1	Number of families	1			
2	Number of comparisons per family	3			
3	Alpha	0.05			
4					
5	Dunn's multiple comparisons test	Rank sum diff.	Significant?	Summary	Adjusted P Value
6	fed vs. fasted 10 days	4.000	No	ns	0.9519
7	fed vs. fasted 20 days	14.00	Yes	**	0.0014
8	fasted 10 days vs. fasted 20 days	10.00	Yes	*	0.0373
9					
10	Test details	Rank sum 1	Rank sum 2	Rank sum diff.	n1
11	fed vs. fasted 10 days	22.00	18.00	4.000	8
12	fed vs. fasted 20 days	22.00	8.000	14.00	8
13	fasted 10 days vs. fasted 20 days	18.00	8.000	10.00	8
					n2

# One-way RM ANOVA

RM one-way ANOVA					
Multiple comparisons					
1	Number of families	1			
2	Number of comparisons per family	3			
3	Alpha	0.05			
4					
5	Tukey's multiple comparisons test	Mean Diff.	95.00% CI of diff.	Significant?	Summary
6	fed vs. fasted 10 days	0.3750	-0.2516 to 1.002	No	ns
7	fed vs. fasted 20 days	3.063	2.526 to 3.599	Yes	****
8	fasted 10 days vs. fasted 20 days	2.688	2.079 to 3.296	Yes	****
9					
10	Test details	Mean 1	Mean 2	Mean Diff.	SE of diff.
11	fed vs. fasted 10 days	41.15	40.78	0.3750	0.2128
12	fed vs. fasted 20 days	41.15	38.09	3.063	0.1822
13	fasted 10 days vs. fasted 20 days	40.78	38.09	2.688	0.2065
					n1
					n2

An example of how you might write up the results of the Friedman test

The data failed many of the assumptions of a one-way RM ANOVA (normality, homoscedasticity, sphericity) but met the assumptions of independence of groups, paired/correlated data, no missing data, equally spaced time measurements, we will use the non-parametric Friedman test. With  $p=0.0003$  (two-sided hypothesis,  $\alpha=0.05$ ), we reject the null hypothesis of equal medians and conclude that at least one median is different. Dunn's post-hoc test 'Fasted 20' (median=38.2) was significantly different from 'Fed' (41.35) and 'Fasted 10' (41.5). There was no significant difference between 'Fed' and 'Fasted 10'. We conclude that plasma protein levels decrease with fasting.

# What about a mixed model?

Parameters: One-Way ANOVA (and Nonparametric or Mixed)

Experimental Design   Repeated Measures   Multiple Comparisons   Options   Residuals

Analyses of repeated measures data can be reported in two ways.

- ANOVA (partition sum-of-squares). This is the same as the general linear model (GLM).
- Mixed-effects model. This uses the restricted maximum likelihood method.

If there are no missing values, the two approaches give the same main results (F and P values). But the methods are very different, so the other reported results differ.

**Analyze using which method**

Repeated measures ANOVA (based on GLM).  
Same as Prism 7 and earlier.  
Requires balanced data (no missing values).

Mixed-effects model.  
Results are presented in a format different than ANOVA.  
Works fine with missing values.

It depends.  
Use ANOVA if there are no missing values.  
Use mixed-effects model if there are missing values.

**What to do if a random effect is zero (or negative)?**

Remove term(s) from model and fit a simpler model (recommended).

Fit the full model anyway (corresponds to NOBOUND parameter in SAS).

Make these choices the default for future ANOVAs (One-, Two- and Three-way).

Learn   Cancel   OK

Tabular results    Multiple comparisons

Mixed-effects analysis  
Tabular results

1	Table Analyzed	Turtle Data
2		
3	<b>Mixed-effects model (REML)</b>	Matching: Across row
4	Assume sphericity?	No
5	Alpha	0.05
6		
7	<b>Fixed effect (type III)</b>	<b>P value</b>
8	Treatment (between columns)	<0.0001
9		<b>P value summary</b>
		****

THERE IS A BIG BUT....

Tabular results    Multiple comparisons

Mixed-effects analysis  
Multiple comparisons

1	Number of families	1				
2	Number of comparisons per family	3				
3	Alpha	0.05				
4						
5	<b>Tukey's multiple comparisons test</b>	<b>Mean Diff.</b>	<b>95.00% CI of diff.</b>	<b>Significant?</b>	<b>Summary</b>	<b>Adjusted P Value</b>
6	fed vs. fasted 10 days	0.3750	-0.2516 to 1.002	No	ns	0.2496
7	fed vs. fasted 20 days	3.063	2.526 to 3.599	Yes	****	<0.0001
8	fasted 10 days vs. fasted 20 days	2.688	2.079 to 3.296	Yes	****	<0.0001
9						
10	<b>Test details</b>	<b>Mean 1</b>	<b>Mean 2</b>	<b>Mean Diff.</b>	<b>SE of diff.</b>	<b>n1</b>
11	fed vs. fasted 10 days	41.15	40.78	0.3750	0.2128	8
12	fed vs. fasted 20 days	41.15	38.09	3.063	0.1822	8
13	fasted 10 days vs. fasted 20 days	40.78	38.09	2.688	0.2065	8
14						

*Running a mixed model is actually rather complex – seek professional help*

SAS:

It is very important to explore different variance-covariance structures when running a mixed model on repeated measure.

The output contains fit statistics indicating how well each model with different variance-covariance structures fits the data compared to other models.

Different variance-covariance structures

- Compound symmetry
- Auto-regressive
- Unstructured
- etc.

We can do a two-way RM ANOVA to test if there is a difference in plasma protein by fasting *and* turtle sex

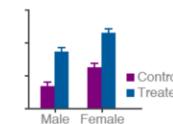
Subject	Sex	Fed	Fasted 10 days	Fasted 20 days
1	Male	42.8	42.4	38.9
2	Male	43.1	42.2	40.3
3	Male	40.0	40.8	37.5
4	Male	46.6	45.9	42.9
5	Female	42.2	42.4	39.7
6	Female	38.7	38.1	35.8
7	Female	35.3	34.3	32.3
8	Female	40.5	40.1	37.3

- XY
- Column
- Grouped
- Contingency
- Survival
- Parts of whole
- Multiple variables
- Nested
  
- Existing file
- Open a file
- LabArchives
- Clone a graph
- Graph portfolio

Grouped tables have two grouping variables, one defined by columns and the other defined by rows

Table format

Grouped	A	B
	Control	Treated
1	A:Y1	A:Y2
2	A:Y3	B:Y1
	B:Y2	B:Y3



[Learn more](#)

Data table:

- Enter or import data into a new table  
 Start with sample data to follow a tutorial

Options:

- Enter and plot a single Y value for each point  
 Enter  replicate values in side-by-side subcolumns  
 Enter and plot error values already calculated elsewhere

Enter: Mean, SD, N

Table format:		Group A				Group B			
Grouped		Males				Females			
		A:1	A:2	A:3	A:4	B:1	B:2	B:3	B:4
1	fed	42.8	43.1	40.0	46.6	42.2	38.7	35.3	40.5
2	fasted 10 days	42.4	42.2	40.8	45.9	42.4	38.1	34.3	40.1
3	fasted 20 days	38.9	40.3	37.5	42.9	39.7	35.8	32.3	37.3

### Analyze Data

#### Built-in analysis

Which analysis?

**Transform, Normalize...**

- Transform
- Transform concentrations (X)
- Normalize
- Prune rows
- Remove baseline and column math
- Transpose X and Y
- Fraction of total

**XY analyses**

**Column analyses**

**Grouped analyses**

**Two-way ANOVA (or mixed model)**

Three-way ANOVA (or mixed model)

Row means with SD or SEM

Multiple t tests - one per row

**Contingency table analyses**

**Survival analyses**

**Parts of whole analyses**

**Multiple variable analyses**

**Nested analyses**

**Generate curve**

**Simulate data**

**Recently used**

Analyze which data sets?

- A:Males  
 B:Females

Select All

Deselect All

### Parameters: Two-Way ANOVA (or Mixed Model)

RM Design RM Analysis Factor names Multiple Comparisons Options Residuals

**Data arrangement**

	Group A		Group B		Group C	
	Title		Title		Title	
	A:Y1	A:Y2	B:Y1	B:Y2	C:Y1	C:Y2
1	Time1					
2	Time2					
3	Time3					
4	Time4					

**Matching by which factor(s)?**

- Each column represents a different time point, so matched values are spread across a row.  
 Each row represents a different time point, so matched values are stacked into a subcolumn.

**Assume sphericity (equal variability of differences)?**

- No. Use the Geisser-Greenhouse correction. Recommended.  
 Yes. No correction.

Based on your choices (on all tabs), Prism will perform:

- RM two-way ANOVA with the Geisser-Greenhouse correction, matched values are stacked into a subcolumn.

Learn

Cancel

OK

## Parameters: Two-Way ANOVA (or Mixed Model)

RM Design RM Analysis Factor names Multiple Comparisons Options Residuals

Analyses of repeated measures data can be reported in two ways.

- ANOVA (partition sum-of-squares). This is the same as the general linear model (GLM).
- Mixed-effects model. This uses the restricted maximum likelihood method.

If there are no missing values, the two approaches give the same main results (F and P values). But the methods are very different, so the other reported results differ.

### Analyze using which method

Repeated measures ANOVA (based on GLM).

Same as Prism 7 and earlier.

Requires balanced data (no missing values).

Mixed-effects model.

Results are presented in a format different than ANOVA.

Works fine with missing values.

It depends.

Use ANOVA if there are no missing values.

Use mixed-effects model if there are missing values.

### What to do if a random effect is zero (or negative)?

Remove term(s) from model and fit a simpler model (recommended).

Fit the full model anyway (corresponds to NOBOUND parameter in SAS).

Make these choices the default for future ANOVAs (One-, Two- and Three-way).

Learn

Cancel

OK

## Parameters: Two-Way ANOVA (or Mixed Model)

RM Design RM Analysis Factor names Multiple Comparisons Options Residuals

### Data arrangement

Table format:		Group A		Group B		Group C	
		Title		Title		Title	
		A:Y1	A:Y2	B:Y1	B:Y2	C:Y1	C:Y2
1	Time1						
2	Time2						
3	Time3						
4	Time4						

### Factor names

Name the factor that defines the columns:

Sex

Name the factor that defines the rows:

Fasting time

Name of matched set (i.e. person or block):

Turtle

Learn

Cancel

OK

## Parameters: Two-Way ANOVA (or Mixed Model)

RM Design RM Analysis Factor names Multiple Comparisons Options Residuals

### What kind of comparison?

Within each column, compare rows (simple effects within columns) < >

	Group A		Group B		Group C	
	Data Set-A		Data Set-B		Data Set-C	
	A:Y1	A:Y2	B:Y1	B:Y2	C:Y1	C:Y2
1	Mean		Mean		Mean	
2	Mean		Mean		Mean	
3	Mean		Mean		Mean	

### How many comparisons?

- Compare each cell mean with every other cell mean on that column.
- Compare each cell mean with the control cell mean on that column.

Control row: Row 1 : fed < >

### How many families?

One family per column (recommended) < >

### Which test?

Use choices on the Options tab to choose the test, and to set the defaults for future ANOVAs.

Learn

Cancel

OK

## Parameters: Two-Way ANOVA (or Mixed Model)

RM Design RM Analysis Factor names Multiple Comparisons Options Residuals

### What kind of comparison?

Compare each cell mean with the other cell mean in that row < >

	Group A		Group B	
	Data Set-A		Data Set-B	
	A:Y1	A:Y2	B:Y1	B:Y2
1	Mean		Mean	
2	Mean		Mean	
3	Mean		Mean	

### How many comparisons?

- Compare each column mean with every other column mean.
- Compare each column mean with the control column mean.

Control column: Group A : Males < >

### Which test?

Use choices on the Options tab to choose the test, and to set the defaults for future ANOVAs.

Learn

Cancel

OK

Parameters: Two-Way ANOVA (or Mixed Model)

RM Design RM Analysis Factor names Multiple Comparisons Options Residuals

**Multiple comparisons test**

Correct for multiple comparisons using statistical hypothesis testing. Recommended.  
Test: Sidak (more power, recommended)

Correct for multiple comparisons by controlling the False Discovery Rate.  
Test: Two-stage step-up method of Benjamini, Krieger and Yekutieli (recommended)

Don't correct for multiple comparisons. Each comparison stands alone.  
Test: Fisher's LSD test

**Multiple comparisons options**

Swap direction of comparisons (A-B) vs. (B-A).

Report multiplicity adjusted P value for each comparison.  
Each P value is adjusted to account for multiple comparisons.

Family-wise significance and confidence level: 0.05 (95% confidence interval)

**Graphing options**

Graph confidence intervals.

**Additional results**

Narrative results.

Show cell/row/column/grand means.

Report goodness of fit.

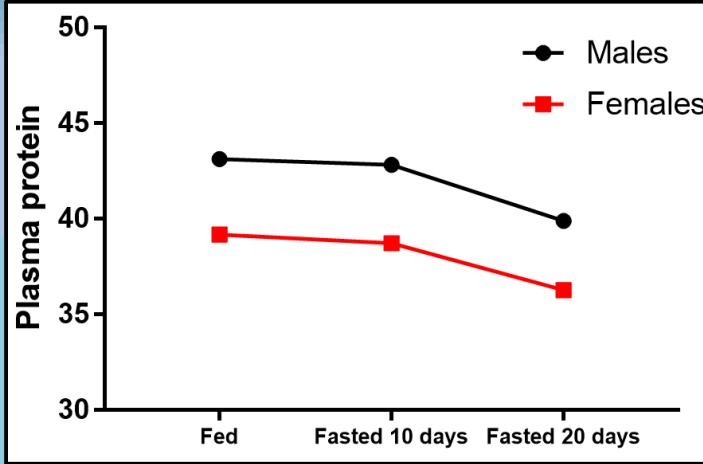
**Output**

Show this many significant digits (for everything except P values): 4

P value style: GP: 0.1234 (ns), 0.0332 (\*), 0.0021 (\*\*), 0.000 N = 6

Make options on this tab be the default for future Two-Way ANOVAs.

Learn Cancel OK



## Evidence of interaction and main effects?

### Evidence of main effect by fasting time?

Are there differences between fasting time points, i.e., are values increasing or decreasing over time (combining males and females)?

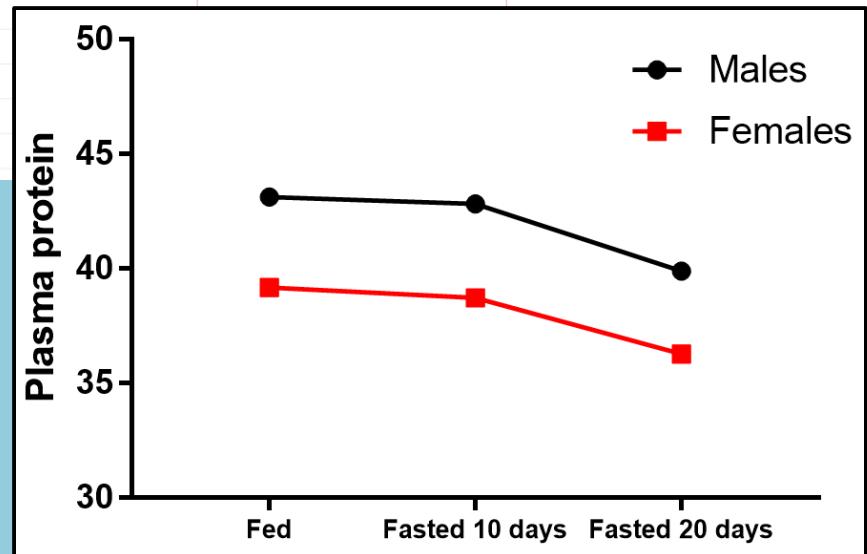
### Evidence of main effect by fasting time?

Do males and females differ at any time points, i.e., is there a main effect where one sex group has higher values compared to the other group?

### Evidence of interaction between sex and fasting time?

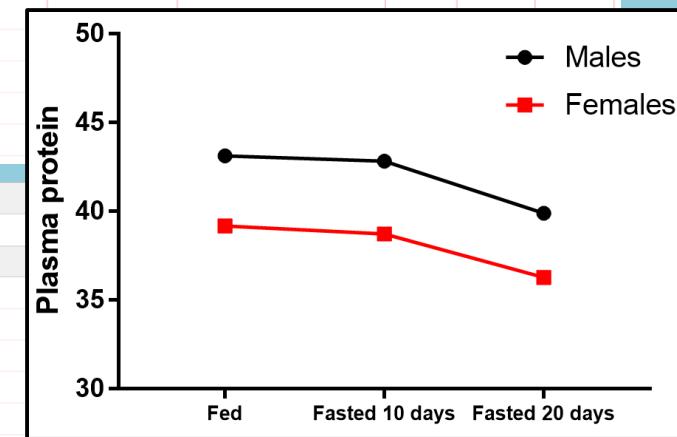
Do the sex groups differ in their responses over fasting time, i.e., is there a time\*group interaction (where one sex group changes differently over time compared to the other)?

2way ANOVA	
ANOVA results	
3 Two-way RM ANOVA	Matching: Stacked
4 Assume sphericity?	No
5 Alpha	0.05
6	
7 Source of Variation	% of total variation
8 Fasting time x Sex	0.08481
9 Fasting time	16.06
10 Sex	32.68
11 Turtle	50.45
12	
13 ANOVA table	SS
14 Fasting time x Sex	0.2358
15 Fasting time	44.65
16 Sex	90.87
17 Turtle	140.3
18 Residual	2.025
19	
20 Difference between column means	
21 Mean of Males	41.95
22 Mean of Females	38.06
23 Difference between means	3.892
24 SE of difference	1.974
25 95% CI of difference	-0.9387 to 8.722



**2way ANOVA**  
Multiple comparisons

1	Compare each cell mean with the other cell mean in that row						
2							
3	Number of families	1					
4	Number of comparisons per family	3					
5	Alpha	0.05					
6							
7	<b>Sidak's multiple comparisons test</b>		<b>Mean Diff.</b>	<b>95.00% CI of diff.</b>	<b>Significant?</b>	<b>Summary</b>	<b>Adjusted P Value</b>
8							
9	Females - Males						
10	fed	-3.950	-10.52 to 2.622	No	ns	0.2620	
11	fasted 10 days	-4.100	-11.20 to 3.004	No	ns	0.2680	
12	fasted 20 days	-3.625	-10.13 to 2.885	No	ns	0.3035	
13							
14							
15	<b>Test details</b>		<b>Mean 1</b>	<b>Mean 2</b>	<b>Mean Diff.</b>	<b>SE of diff.</b>	<b>N1</b>
16							
17	Females - Males						
18	fed	39.18	43.13	-3.950			
19	fasted 10 days	38.73	42.83	-4.100			
20	fasted 20 days	36.28	39.90	-3.625			



**2way ANOVA**  
Multiple comparisons

1	Within each column, compare rows (simple effects within columns)						
2							
3	Number of families	2					
4	Number of comparisons per family	3					
5	Alpha	0.05					
6							
7	<b>Tukey's multiple comparisons test</b>		<b>Mean Diff.</b>	<b>95.00% CI of diff.</b>	<b>Significant?</b>	<b>Summary</b>	<b>Adjusted P Value</b>
8							
9	Males						
10	fed vs. fasted 10 days	0.3000	-1.291 to 1.891	No	ns	0.7351	
11	fed vs. fasted 20 days	3.225	1.804 to 4.646	Yes	**	0.0051	
12	fasted 10 days vs. fasted 20 days	2.925	1.434 to 4.416	Yes	**	0.0078	
13							
14	Females						
15	fed vs. fasted 10 days	0.4500	-0.5947 to 1.495	No	ns	0.3096	
16	fed vs. fasted 20 days	2.900	2.285 to 3.515	Yes	***	0.0006	
17	fasted 10 days vs. fasted 20 days	2.450	1.678 to 3.222	Yes	**	0.0019	
18							

## How you might write this up

*<you checked all assumptions and you conclude that you can use the two-way RM ANOVA>*

We used a two-way RM ANOVA to study differences in plasma protein by time fasting and sex of turtle. Interaction between gender and time fasting was not significant (we fail to reject the null hypothesis of no interaction (two sided hypothesis,  $F(2,12)=0.70$ ,  $p=0.53$ ,  $\alpha=0.05$ ). Gender was also not significant ( $F(2,12)=3.886$ ,  $p=0.10$ ). With  $p<0.0001$  (two-sided hypothesis,  $F(2,12)=132.3$ ,  $\alpha=0.05$ ), we reject the null hypothesis that mean plasma protein levels are the same across groups of time fasting and conclude that at least one is different. Tukey's post hoc tests showed that 'Fasted 20' (mean=38.1) was significantly different from 'Fed' (41.2) and 'Fasted 10' (40.8). There was no significant difference between 'Fed' and 'Fasted 10'. We conclude that plasma protein levels decrease with fasting and that gender has no effect on this finding.

Note: because gender was not significant, report the results for time fasting for males and females combined.

# Using a one-way ANOVA when a RM ANOVA would be better

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PLoS One 2014 Apr 23;9(4):e95964



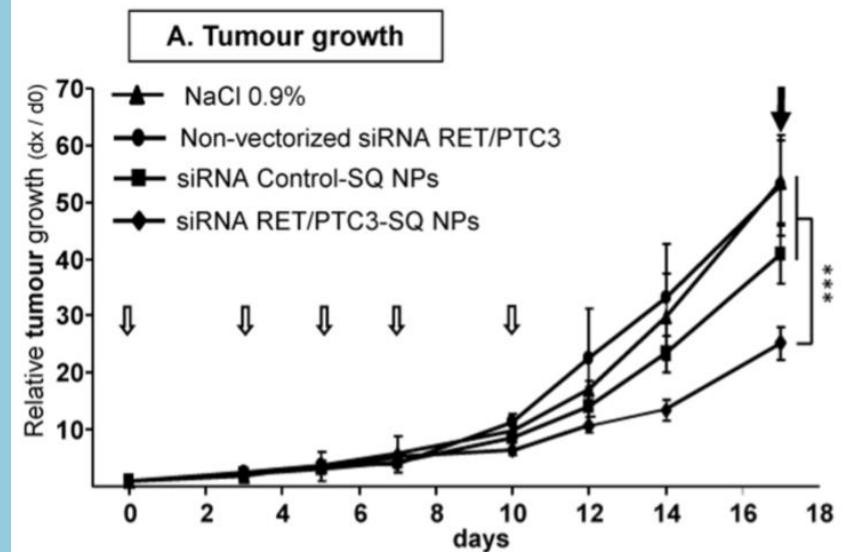
## Effects of siRNA on RET/PTC3 Junction Oncogene in Papillary Thyroid Carcinoma: From Molecular and Cellular Studies to Preclinical Investigations

Hafiz Muhammad Ali<sup>1,2,3</sup>, Giorgia Urbinati<sup>1,2,3</sup>, Hubert Chapuis<sup>4</sup>, Didier Desmaele<sup>4</sup>, Jean-Rémi Bertrand<sup>1,2,3</sup>, Patrick Couvreur<sup>4</sup>, Liliane Massaad-Massade<sup>1,2,3\*</sup>

**1** Université Paris-Sud 11, Laboratoire de Vectorologie et Thérapeutiques Anticancéreuses, UMR 8203, Villejuif, France, **2** CNRS, Villejuif, Laboratoire de Vectorologie et Thérapeutiques Anticancéreuses, UMR 8203, Villejuif, France, **3** Gustave Roussy, Laboratoire de Vectorologie et Thérapeutiques Anticancéreuses, UMR 8203, Villejuif, France, **4** Institut Galien, UMR CNRS 8612, Université Paris-Sud 11, Faculté de pharmacie, Châtenay-Malabry, France

### Statistical analysis

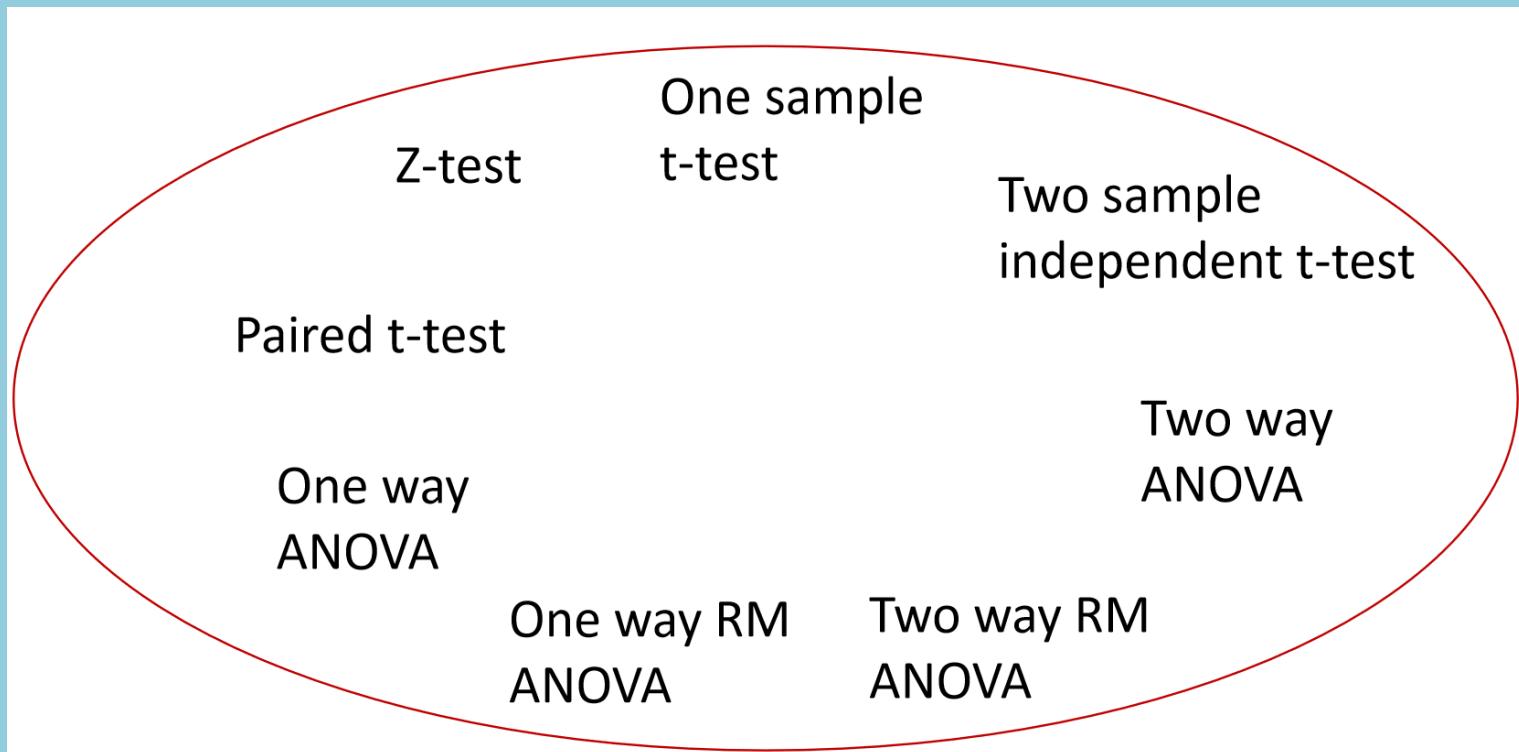
All data are presented as mean  $\pm$  SD (Standard deviation). One-way analysis of variance (ANOVA) was employed to compare multiple treatments. Using a linear regression model, doubling time was calculated by GraphPad Prism 4 software. While all pair wise comparisons between different treatment groups were done by least significant difference (LSD) Post-hoc test by using “InVivoStat” software.  $p<0.05$  was considered as statistically significant level.



To compare difference between treatments, a statistical analysis was performed by using ANOVA followed by LSD Post-hoc test ( $***=p<0.001$ )

# Applied statistics so far: Which test should you use?

You have learned many statistical tools. Given a real problem, can you decide which one to use?



## Example 1

A principal at a certain school claims that the students in his school are above average intelligence. A random sample of thirty students IQ scores have a mean score of 112. Is there sufficient evidence to support the principal's claim? It is known that the mean population IQ is 100 with a standard deviation of 15.

## Example 2

A study was designed to compare the effect of two soporific drugs. 100 volunteers were chosen and they were randomly assigned either drug A or drug B. Number of hours of increased sleep was recorded for each patient. Use the data to test if Drug A is more effective than Drug B at increasing sleep.

## Example 3

A sample of 50 students were given a diagnostic test before studying a particular module and then again after completing the module. Analyze the data to test if studying the module leads to improvements in students' knowledge/skills (i.e. test scores).

## Example 4

A study was designed to combine the effect of two soporific drugs. 150 volunteers were chosen and they were randomly assigned either drug A, drug B or drug A+B. Each drug was administered to 50 participants. Number of hours of increased sleep was recorded for each patient. Use the data to test if there is any difference among effectiveness of the drugs to increase sleep.

## Example 5

A researcher was interested in whether an individual's interest in politics was influenced by their level of education and gender. In particular, the researcher wanted to know if the effect of level of education on interest in politics is different for males and females? To answer this question, a random sample of 60 participants were recruited to take part in the study – 30 males and 30 females – equally split by level of education: school, college and university (i.e., 10 participants in each group). Each participant in the study completed a questionnaire that scored their interest in politics on a scale of 0 to 100, with higher scores indicating a greater interest in politics.

How can these be used to answer the research questions?

## Example 6

This is a study to test the effect of a specific diet program proposed to reduce weight. Ten people in need of losing weight agree to try the diet for two months. Their weight was measured in Kilograms at the start of the diet and then after 1 month and 2 months. Did the diet work?

Before Diet	After 1 Month	After 2 Months
63.75	65.38	81.34
62.98	66.24	69.31
65.98	67.70	77.89
107.27	102.72	91.33
66.58	69.45	72.87
120.46	119.96	114.26
62.01	66.09	68.01
71.87	73.62	55.43
83.01	75.81	71.63
76.62	67.66	68.60

## Example 7

Severe atheroma of the aortic arch has now been established as an important risk factor and mechanism for stroke. Two drugs (A) a statin (B) an anticoagulant were given to stroke patients and followed up every 6 months for 3 years. At every follow up, the thickness of plaque was measured. The scientific questions are:

- (i) Does the thickness of the plaque change over time post medication?
- (ii) Is there any difference between the change in the thickness of plaque between patients who received drug A and patients who received drug B?