
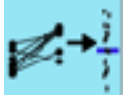
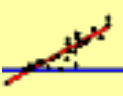





# Common statistical tests are linear models

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See worked examples and more details at the accompanying notebook: <https://github.com/eigenfoo/tests-as-linear>

	Common name	Function in scipy.stats	Equivalent linear model in smf.ols	Exact?	The linear model in words	Icon
Simple Regression: ( $y \sim 1 + x$ )	<b>y is independent of x</b> P: One-sample t-test N: Wilcoxon signed-rank	scipy.stats.ttest_1samp(y) scipy.stats.wilcoxon(y)	smf.ols("y ~ 1", data) smf.ols("y ~ 1", signed_rank(data))	✓ <a href="#">for N &gt; 14</a>	One number (intercept, i.e., the mean) predicts <b>y</b> . - (Same, but it predicts the <i>signed rank</i> of <b>y</b> .)	
	P: Paired-sample t-test N: Wilcoxon matched pairs	scipy.stats.ttest_rel(y1, y2) scipy.stats.wilcoxon(y1, y2)	smf.ols("y2_sub_y1 ~ 1", data) smf.ols("y2_sub_y1 ~ 1", signed_rank(data))	✓ <a href="#">for N &gt; 14</a>	One intercept predicts the pairwise <b>y<sub>2</sub>-y<sub>1</sub></b> differences. - (Same, but it predicts the <i>signed rank</i> of <b>y<sub>2</sub>-y<sub>1</sub></b> .)	
	<b>y ~ continuous x</b> P: Pearson correlation N: Spearman correlation	scipy.stats.pearsonr(x, y) scipy.stats.spearmanr(x, y)	smf.ols("y ~ 1 + x", data) smf.ols("y ~ 1 + x", rank(data))	✓ <a href="#">for N &gt; 10</a>	One intercept plus <b>x</b> multiplied by a number (slope) predicts <b>y</b> . - (Same, but with <i>ranked x</i> and <b>y</b> )	
	<b>y ~ discrete x</b> P: Two-sample t-test P: Welch's t-test N: Mann-Whitney U	scipy.stats.ttest_ind(y1, y2) N/A scipy.stats.mannwhitneyu(y1, y1)	smf.ols("y ~ 1 + group", data) <sup>A</sup> N/A smf.ols("y ~ 1 + group", signed_rank(data)) <sup>A</sup>	✓ ✓ <a href="#">for N &gt; 11</a>	An intercept for <b>group 1</b> (plus a difference if <b>group 2</b> ) predicts <b>y</b> . - (Same, but with one variance <i>per group</i> instead of one common.) - (Same, but it predicts the <i>signed rank</i> of <b>y</b> .)	
Multiple regression: ( $y \sim 1 + x_1 + x_2 + \dots$ )	P: One-way ANOVA N: Kruskal-Wallis	scipy.stats.f_oneway(a, b, c) scipy.stats.kruskal(a, b, c)	smf.ols(y ~ 1 + G <sub>2</sub> + G <sub>3</sub> + ... + G <sub>N</sub> ) <sup>A</sup> smf.ols(rank(y) ~ 1 + G <sub>2</sub> + G <sub>3</sub> + ... + G <sub>N</sub> ) <sup>A</sup>	✓ <a href="#">for N &gt; 11</a>	An intercept for <b>group 1</b> (plus a difference if group ≠ 1) predicts <b>y</b> . - (Same, but it predicts the <i>rank</i> of <b>y</b> .)	
	P: One-way ANCOVA	N/A	smf.ols("y ~ 1 + G <sub>2</sub> + G <sub>3</sub> + ... + G <sub>N</sub> + x", data) <sup>A</sup>	✓	- (Same, but plus a slope on <b>x</b> .) <i>Note: this is discrete AND continuous. ANCOVAs are ANOVAs with a continuous x.</i>	
	P: Two-way ANOVA	N/A	smf.ols("y ~ 1 + G <sub>2</sub> + G <sub>3</sub> + ... + G <sub>N</sub> + S <sub>2</sub> + S <sub>3</sub> + ... + S <sub>K</sub> + G <sub>2</sub> *S <sub>2</sub> +G <sub>3</sub> *S <sub>3</sub> +...+G <sub>N</sub> *S <sub>K</sub> ", data)	✓	Interaction term: changing <b>sex</b> changes the <b>y ~ group</b> parameters. <i>Note: G<sub>2</sub> to N is an indicator (0 or 1) for each non-intercept levels of the group variable. Similarly for S<sub>2</sub> to K for sex. The first line (with G<sub>i</sub>) is main effect of group, the second (with S<sub>j</sub>) for sex and the third is the group × sex interaction. For two levels (e.g. male/female), line 2 would just be "S<sub>2</sub>" and line 3 would be S<sub>2</sub> multiplied with each G<sub>i</sub>.</i>	[Coming]
	<b>Counts ~ discrete x</b> N: Chi-square test	scipy.stats.chisquare(data)	sm.GLM(y ~ 1 + G <sub>2</sub> + G <sub>3</sub> + ... + G <sub>N</sub> + S <sub>2</sub> + S <sub>3</sub> + ... + S <sub>K</sub> + G <sub>2</sub> *S <sub>2</sub> +G <sub>3</sub> *S <sub>3</sub> +...+G <sub>N</sub> *S <sub>K</sub> , family=...) <sup>A</sup>	✓	Interaction term: (Same as Two-way ANOVA.) <i>Note: Run glm using the following arguments: glm(model, family=poisson()) As linear-model, the Chi-square test is log(y<sub>i</sub>) = log(N) + log(a<sub>i</sub>) + log(β<sub>j</sub>) + log(a<sub>i</sub>β<sub>j</sub>) where a<sub>i</sub> and β<sub>j</sub> are proportions. See more info in the accompanying notebook.</i>	Same as Two-way ANOVA
	N: Goodness of fit	scipy.stats.chi2_contingency(data)	sm.GLM(y ~ 1 + G <sub>2</sub> + G <sub>3</sub> + ... + G <sub>N</sub> , family=...) <sup>A</sup>	✓	(Same as One-way ANOVA and see Chi-Square note.)	1W-ANOVA

List of common parametric (P) non-parametric (N) tests and equivalent linear models. The notation  $y \sim 1 + x$  is R shorthand for  $y = 1 \cdot b + a \cdot x$  which most of us learned in school. Models in similar colors are highly similar, but really, notice how similar they *all* are across colors! For non-parametric models, the linear models are reasonable approximations for non-small sample sizes (see "Exact" column and click links to see simulations). Other less accurate approximations exist, e.g., Wilcoxon for the sign test and Goodness-of-fit for the binomial test. The signed rank function is `signed_rank(df) = np.sign(df) * df.rank()`. The variables  $G_i$  and  $S_i$  are "[dummy coded](#)" [indicator variables](#) (either 0 or 1) exploiting the fact that when  $\Delta x = 1$  between categories the difference equals the slope. Subscripts (e.g.,  $G_2$  or  $y_1$ ) indicate different columns in data. Im requires long-format data for all non-continuous models. All of this is exposed in greater detail and worked examples at <https://eigenfoo.xyz/tests-as-linear/>.

<sup>A</sup> See the note to the two-way ANOVA for explanation of the notation.



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