Testing for Significant Differences Between Distributions

A Crib Sheet

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The overall question

If you have done an experiment and collected some sets of measures from two or more experimental setups, a common question is:

Q: Is there a significant difference between the distributions represented by the points making up my data sets?

This set of slides gives you some advice on how to proceed to answer this question.

More detailed answers can be found in excellent online sources. My two favourites are these - they both give excellent explanations:

McDonald, J.H. (2014): Handbook of Biological Statistics (3rd ed.).

Sparky House Publishing, Baltimore, Maryland.

http://www.biostathandbook.com/index.html

NIST/SEMATECH (2013): e-Handbook of Statistical Methods,

http://www.itl.nist.gov/div898/handbook/

Some questions will depend on the answers to one or both of these questions, as applied to each of your sets of points:

• Q: Is this set of points normally distributed?

^{*}Razali & Wah (2011): Power comparisons of Shapiro-Wilk, Kolmogorov-Smirnov, Lilliefors and Anderson-Darling Tests Journal of Statistical Modelling and Analytics 2(1):21–33. ISBN: 978-967-363-157-5 Link via ResearchGate (Apr 2021)

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 - Anderson-Darling,
 - Kolmogorov-Smirnov
 - a Q-Q plot (quartile-quartile plot)

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- Q: Are the variances between two of my sets of points different? (a.k.a. Q: Is my data homoscedastic?)

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- Q: Are the variances between two of my sets of points different? (a.k.a. Q: Is my data homoscedastic?)
 - F-test (of equality of variances)

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Statistics: Independent two-sample *t*-test

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 you can use the (independent two-sample) t-test to answer your question.

 \mathcal{Q} : Is there a significant difference between the distributions represented by the points making up my data sets?

If all of the following are true:

- you have more than two distributions of data (sets of points),
- and they are normally distributed
- and your observations are independent
- and your variances are equal (a.k.a. homoscedasticity)
- and your question is "Q: do these samples come from distributions with the same mean?"

then

Statistics: ANOVA

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- this only tells you that <u>one or more</u> of the means of your distributions is different from the others, and doesn't tell you which one.
- You run a post hoc test to determine that.

- you collected your points in paired observations from a set
 of subjects or entities, so that you have two measures from
 the same subject under two conditions, and you want to
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Statistics: Paired sample *t*-test

Here is something that people frequently get mixed up for the paired t-test ...

The Normality requirement only applies to the differences

Note that only the **differences** calculated between the pairs need to be normally distributed – the distributions of the unpaired points does not matter at all.

This is because the paired *t*-test **converts the set of differences into a single distribution**, and uses a *t*-test to determine if **this distribution** is non-zero.

If you have point sets \mathcal{A} , \mathcal{B} , \mathcal{C} (or more), and want to make multiple comparisons, such as \mathcal{A} : \mathcal{B} and \mathcal{B} : \mathcal{C} , then use a correction such as the **Bonferroni** correction:

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Why do we do this?

This decreases our likelihood of saying that any test has produced a statistically significant result — because we are attempting to control for the likelihood of finding such a result by random chance.

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- If your data are NOT normally distributed
 - or your variances do not conform to your constraints
- but your observations are independent

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Note that KW test functions like a **one-way ANOVA**, and to determine which pairs differ, you can use MWW <u>without</u> the **Bonferroni correction**.

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If your data are **NOT** independent then

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- if you have two distributions (and ... would have considered Mann-Whitney-Wilcoxon) then use the Wilcoxon signed-rank test and
- if you have more than two distributions (and would have considered Kruskal-Wallis), then use the Friedman test instead.

 \mathcal{Q} : Is there a significant difference between the distributions represented by the points making up my data sets?

Independence	Normality	N dist.	\rightarrow	Use	
	Yes	2			
Yes	ies	3 or more			
165	No	2			‡
	INO	3 or more			

⁴Of differences only. See note on page 22.

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Yes	Yes⁴	Pairs	paired <i>t</i> -test

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