Diamond Plot

for Comparing Group Means and Variability

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Comparing group differences for examining treatment effectiveness is a common practice in research and evaluation. Parametric procedures such as t-tests and F-tests are widely used for this purpose. However, those procedures are not very informative because the conclusion is nothing more than rejecting or failing to reject the null hypothesis.

APA Task Force on Statistical Inference (Wilkinson, 1996) endorsed the use of confidence intervals (CI) as a supplement to conventional p value. By using CI, the researcher can look at the group differences by means and variability. As the sample size increases, the variability decreases, and the CI gets narrower. Why should we judge the quality of a CI by its narrowness? Take this scenario as a metaphor: You ask me to guess your age, I reply, "from 16 to 60." I am 95% confident that your actual age would fall within this range, but is it a useful estimation? Probably not. If I say "from 18-21" instead, it is definitely a much better answer.

SAS/JMP provides a powerful tool named diamond plot to visualize CI and it is very easy to obtain the result. In JMP you don't even need to know the name of the procedure. As long as you know what your dependent and independent variables are, you can simply choose **Fit Y by X** from the **Analyze** menu, as shown in the following:

Fit Y by X - Contextual - JMP Pro						
Distribution of Y for each X. Modeling types determine analysis.						
- Select Columns	elect Columns Cast Selected Columns into Roles Action					
Group	Y, Response	Score	ОК			
Score		optional	Cancel			
Oneway	X, Factor	📕 Group				
		optional				
<u>•∕-</u>	Block	optional	Remove			
Bivariate Oneway	Weight	optional numeric	Recall			
	Freq	optional numeric	Help			
Logistic Contingency	Ву	optional				
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JMP provides the user with a contextual menu system and thus you would not be overwhelmed by too many options. In the next screen only the options that are applicable to the data structure are available to you. At this stage, you can select **Quantiles** to display the box plot and **Means/Anova** to display the diamond plot.

(On	Oneway Analysis of Score By Group				
ΛĴ	Quantiles				
	Means/Anova				
	Means and Std Dev				
	Analysis of Means Methods				
	Compare Means				
	Nonparametric				
	Unequal Variances				
	Equivalence Test				
	Power				
	Set α Level				
	Normal Quantile Plot				
	CDF Plot				
	Densities •				
	Matching Column				
	Save +				
	Display Options				
	Script +				

The result is shown in the following figure. It condenses a lot of important information:



• Grand sample mean: it is represented by a horizontal black line

- Group means: the horizontal line inside each diamond is the group means
- **Confidence intervals**: The diamond is the CI for each group. Because the population parameter is unknown, there is always some uncertainty in estimation. Thus, we need to bracket the estimation. Take photography as an analogy. If the photographer is not sure whether the exposure is correct, he would take at least one over-exposed photo (upper bound), one under-exposed photo (lower bound), and one in the middle. In the JMP output, the top of the diamond is the upper bound (best case scenario) while the bottom is the lower bound (worst case scenario).
- **Quantile**: In addition to CI, JMP also provides the option of overlaying a boxplot showing quantile information



In this hypothetical example, Professor Yu taught three classes in different modes: Conventional classroom, online class, and hybrid class. He wants to know which method could yield better exam scores. It is obvious that the performance gap between classroom group and the two others is significant, because even the upper bound of the classroom group is worse than the lower bound of the other two. However, it seems that the difference between the hybrid group and the online group is not substantive at all because there is a lot of overlapping between the two groups. If you need to report formal statistics, you can extract the appropriate information below the graphic.



When I was a graduate student, I took a course in multiple comparison procedures (MPC) as a post hoc step after ANOVA. At most the F test of ANOVA could tell you whether one of the means differ from one of the other means. In order to test which pairwise difference is significant but control the Type I error rate at the same time, different MPCs are needed.

The course required the learners to memorize the pros and cons of 10-15 tests, such as LSA, Bonferroni, Ryan, Tukey, Duncan, Gabriel...etc.. To tell you the truth, today I forgot most of the information. The following is a screenshot of MPCs offered by SPSS. You can tell how confusing it is. In my opinions, the diamond pot is a much quicker and easier way for group comparison.

Cone-Way ANOVA: Post Hoc Multiple Comparisons						
_ Equal Variances A	ssumed		1			
LSD	<u>В-N-К</u>	─ Waller-Duncan				
Bonferroni	Tukey	Type I/Type II Error Ratio: 100				
🔲 S <u>i</u> dak	🔲 Tu <u>k</u> ey's-b	Dunn <u>e</u> tt				
Scheffe	🔲 <u>D</u> uncan	Control Category : Last				
🗏 <u>R</u> -E-G-W F	E Hochberg's GT2	2 Test				
🗏 R-E-G-W <u>Q</u>	Cabriel	O 2-sided $O < Control O > Control$				
Equal Variances Not Assumed						
Tamhane's T2 Dunnett's T3 Games-Howell Dunnett's C						
Significance level: 0.05						
Continue Cancel Help						

However, Payton, Greenstone and Schenker (2003) warned researchers that inferring from non-overlapping CIs to significant mean differences is a dangerous practice, because the error rate associated with this comparison is quite large. The probability of overlap is a function of the standard error. As the standard errors become less homogeneous, the probability of overlap decreases. Simulations result showed that when the standard errors are approximately equal, using 83% or 84% size for the intervals will give an approximate alpha = 0.05 test, but using 95% confidence intervals, which is a common practice, will give very conservative results. Thus, researchers are encouraged to use both CI and hypothesis testing.

References

Payton, M. E., Greenstone, M. H., & Schenker, N. (2003). Overlapping confidence intervals or standard error intervals: What do they mean statistical significance? *Journal of Insect Science*, *3*(34). Retrieved April 21, 2008 from <u>http://insectscience.org/3.34</u>

Wilkinson, L, & the task Force on Statistical Inference. (1996). Stataistical methods in psychology journals; Guidelines and explanations. Retrieved from <u>http://www.apa.org</u>/science/leadership/bsa/statistical/tfsi-followup-report.pdf

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