

Brief note on degrees of freedom.

Degrees of freedom = # indep samples - numbers of parameters in stat. to be est.

$$t\text{-stat. } \frac{\bar{x} - \mu}{\frac{s}{\sqrt{N-1}}} \quad \mu \text{ est. (1 param)}$$

$$\chi^2 \quad \chi^2 = \frac{(N-1) s^2}{\sigma^2} \quad \sigma^2 \text{ estimated (1 param)}$$

## CAVEAT IN DETERMINING DEGREES OF FREEDOM!!

- The samples N must be "independent"
- Independence means samples are not related, or auto correlated, in time and space.

More on this later - - -

## ~~Hypothesis~~ testing

What are we trying to do?

Evaluate whether something satisfies a null hypothesis ( $H_0$ ) or an alternative ( $H_1$ )

$H_0$  = ~~is~~ statistically insignificant

$H_1$  = statistically significant (i.e interesting)

In statistics, the word "significant" means that the alternative hypothesis is satisfied - a very specific definition!!

Typical significance levels = 90%, 95%, 99%

Type I error: Reject hypothesis and it turns out to be correct

Type II error: Accept hypothesis and it's wrong.

## How to do hypothesis testing:

Lect. 5

- 1) State the desired significance level
- 2) Formulate the null hypothesis and the alternative hypothesis.
- 3) Choose the statistic (e.g. t-stat)
- 4) Define critical region
- 5) Evaluate the statistic and state the conclusion

" $H_0$  is rejected with  $\alpha \geq 0.05$  certainty."

Hartmann's simple example: Dilbert's ski vacation:

- Dilbert skis 10 times per season
- Temperature avg. on days there is  $35^{\circ}\text{F}$
- Climatological mean is  $32^{\circ}\text{F}$
- Standard deviation =  $5^{\circ}\text{F}$

- 1) Use a significance level of 95%

- 2) Null hypothesis: The mean for Dilbert's 10 days is no different than climatology  
Alternative: Days were warmer than climatology

- 3) Use t-statistic

$$t = \frac{\bar{x} - \mu}{\sigma / \sqrt{n-1}} = \frac{35 - 32}{5 / \sqrt{9}}$$

- 4)  $t = 1.80$  with 9 deg. of freedom  
 $t_{\text{critical}} = t_{0.05} = 1.83$  (one-sided)

∴ t-test is not satisfied.

5) Conclusion : Dilbert did not experience conditions that were statistically ~~significantly~~ significantly different from the mean climate. (ie. null hypothesis is not rejected).

Compositing or superimposed epoch analysis.

Idea : Isolate signal of an event from the background by averaging data in relation to the event.

Advantage : ~~Simple~~ Simple and powerful technique. Statistics are very straightforward to interpret.

Disadvantage : Can be easily misused. ~~How to~~ What criteria are used to select composite ~~is selected~~ is at the discretion of the user.

Steps in compositing

- 1) Determine categories
- 2) Compute mean, variance, etc. for each category
- 3) Statistical test
- 4) Display results with significance.

## Significance of Composite results

- ↗ Quantitative  
Well talk about ways to account for this later. after we talk about correlation.
- 1) Only use the statistical tests if you expect the result "a priori" - or in other words there is not "large" relationship in time and space between in the data (i.e. autocorrelation)
  - 2) You should subdivide the sample to make sure data are independent (more later..)
  - 3) Composites should be based on a good theory that has <sup>objective</sup> physical basis (e.g. ENSO, diurnal cycle, etc..) That is a good "a priori"

If there is no a priori expectation in constructing the composite, the relationship was found after "fishing."

Some examples of Compositing - and their misuse (Hartmann's notes)

### ENSO Compositing example.

- 1) Composite ~~all~~ all the months in a 50 year record according to Niño 3 index. = 600 months.  
→ 100 months in a positive composite (El Niño)  
100 months in a negative comp. (La Niña)
- 2) Compute mean and standard deviation monthly rainfall of each composite, then use a

~~difference of means test~~ to evaluate whether the El Niño and La Niña composites are statistically different or not.

3) ~~Get the result that~~

Question: Should we trust the results from this statistical analysis? Why or why not?

This example makes two errors

- 1) There is autocorrelation in the ENSO data. That is, SSTAs in the tropical Pacific are not going to change that much over a month. So the true number of independent samples should take this into account is actually much less.  $\rightarrow$  "A posteriori problem"
- 2) The formulation of the hypothesis is not based on a very good physical basis - do we assume ENSO relationships and physical mechanisms of rainfall are constant all year  $\nexists$  in AZ?  
 $\rightarrow$  "A priori problem"