Statistical Description of Data

- Cf. NRiC, Chapter 14.
- Statistics provides tools for understanding data.
 - In the wrong hands these tools can be dangerous!
- Here's a typical data analysis cycle:
 - 1. Apply some formula to data to compute a "statistic".
 - 2. Find where value falls in a probability distribution computed on the basis of some "null hypothesis".
 - 3. If it falls in an unlikely spot (on distribution tail), conclude null hypothesis is *false* for your data set.

Statistics

- Statistics and probability theory are closely related. Statistics can never prove things, only disprove them by ruling out hypotheses.
- Distinguish between *model-independent* statistics (this class, e.g. mean, median, mode) and *model-dependent* statistics (next class, e.g. least-squares fitting).
- Will make use of special functions (e.g. gamma function) described in *NRiC*, Chapter 6.

Moments of a Distribution

- *Cf. NRiC* §14.1.
- The <u>mean</u>, <u>median</u>, and <u>mode</u> of distributions are called *measures of central tendency*.
- The most common description of data involves its *moments*, sums of integer powers of the values.
- The most familiar moment is the <u>mean</u>:

$$\overline{x} = \langle x \rangle = \frac{1}{N} \sum_{i=1}^{N} x_i$$

Variance

• The width of the central value is estimated by its second moment, called the <u>variance</u>:

Var =
$$\frac{1}{N-1} \sum_{i=1}^{N} (x_i - \overline{x})^2$$

or its square root, the standard deviation:

$$\sigma = \sqrt{\mathrm{Var}}$$

Why N -1? If the mean is known *a priori*, i.e. if it's *not* measured from the data, then use N, else N -1. If this matters to you, then N is probably too small!

More on Moments

• A clever way to minimize round-off error when computing the variance is to use the *corrected two-pass algorithm*. First compute *<x>*, then do:

$$\operatorname{Var} = \frac{1}{N-1} \left\{ \sum_{i=1}^{N} (x_i - \overline{x})^2 - \frac{1}{N} \left[\sum_{i=1}^{N} (x_i - \overline{x}) \right]^2 \right\}$$

- The second sum would be zero if <x> were exact, but otherwise it does a good job of correcting RE in Var.
- Higher moments, like skewness (3rd moment) and kurtosis (4th moment) are also sometimes used.

Distribution Functions

- A <u>distribution function</u> (DF) p(x) gives the <u>probability</u> of finding value between x & x + dx.
 - The expected mean data value is:

$$\langle x \rangle = \frac{\int_{-\infty}^{\infty} x p(x) dx}{\int_{-\infty}^{\infty} p(x) dx}$$

- For a discrete DF: $\langle x \rangle = \frac{\sum_{i} x_{i} p_{i}}{\sum_{i} p_{i}}$

• Similar to weighted means, e.g. center of mass.

Median

• The <u>median</u> of a DF is the value x_{med} for which larger & smaller values of *x* are equally probable:

$$\int_{-\infty}^{x_{\text{med}}} p(x) dx = \frac{1}{2} = \int_{x_{\text{med}}}^{\infty} p(x) dx$$

• For discrete values, sort in ascending order, then:

$$x_{\text{med}} = \frac{x_{(N+1)/2}, \qquad N \text{ odd}}{\frac{1}{2} (x_{N/2} + x_{(N/2)+1}), N \text{ even}}$$

Mode

- The <u>mode</u> of a probability DF *p*(*x*) is the value of *x* where the DF takes on a maximum value.
- Most useful when there is a single, sharp max, in which case it estimates the central value.
- Sometimes a distribution will be *bimodal*, with two relative maxima. In this case the mean and median are not very useful since they give only a "compromise" value between the two peaks.

Comparing Distributions

- Often want to know if two distributions have different means or variances (*NRiC* §14.2):
 - 1. Student's *t*-test for significantly different means.
 - a) Find no. of *standard errors* ~ $\sigma/N^{1/2}$ between two means.
 - b) Compute statistic using nasty formula.
 - c) Small numerical value indicates significant difference.
 - 2. *F*-test for significantly different variances.
 - a) Compute $F = Var_1/Var_2$ and plug into nasty formula.
 - b) Small value indicates significant difference.

Comparing Distributions, Cont'd

- Given two sets of data, can generalize to a single question: Are the sets drawn from the same DF?
- Recall can only disprove, not prove.
- May have continuous or binned data.
- May want to compare one data set with known DF, or two unknown data sets with each other.
- Popular technique for binned data is the X² test.
 For continuous data, use the KS test. *NRiC* §14.3.

Chi-Square (χ^2) Test

• Suppose have N_i events in *i*th bin but expect n_i :

$$\chi^2 = \sum_i \frac{(N_i - n_i)^2}{n_i}$$

- Large value of χ^2 indicates unlikely match.
- Compute probability $Q(\chi^2|\nu)$ from *incomplete gamma function*, where ν is # *degrees of freedom*.
- For two binned data sets with events R_i and S_i :

$$\chi^{2} = \sum_{i} \frac{(R_{i} - S_{i})^{2}}{R_{i} + S_{i}}$$

Kolmogorov-Smirnov (KS) Test

- Appropriate for unbinned distributions.
- From <u>sorted</u> list of data points, construct estimate $S_N(x)$ of the *cumulative* DF of the probability DF from which it was drawn...
 - $S_N(x)$ gives fraction of data points to the left of x.
 - Constant between x_i 's, jumps 1/N at each x_i .
 - Note $S_N(x_{\min}) = 0$, $S_N(x_{\max}) = 1$.
 - Behavior between $x_{\min} \& x_{\max}$ distinguishes distributions.

KS Test, Cont'd

- Statistic is maximum value of absolute difference between two cumulative DFs.
- To compare data set to known cumulative DF:

$$D = \max_{x_{\min} \le x \le x_{\max}} \left| S_N(x) - P(x) \right|$$

• To compare two unknown data sets:

$$D = \max_{x_{\min} \le x \le x_{\max}} \left| S_{N_1}(x) - S_{N_2}(x) \right|$$

• Plug *D* and *N* (or $N_e = N_1 N_2 / (N_1 + N_2)$) into nasty formula to get numerical value of significance.