## Class 10. Statistics and the K-S Test

## Statistical Description of Data

- Cf. NRiC §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 that 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 §6.


## Moments of a Distribution

- The mean, median, and mode 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 mean:

$$
\bar{x}=\langle x\rangle=\frac{1}{N} \sum_{i=1}^{N} x_{i} .
$$

## Variance

- The width of the central value is estimated by its second moment, called the variance,

$$
\operatorname{Var}=\frac{1}{N-1} \sum_{i=1}^{N}\left(x_{i}-\bar{x}\right)^{2},
$$

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!
- A clever way to minimize round-off error when computing the variance is to use the corrected two-pass algorithm. First compute $\bar{x}$, then do:

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

- The second sum would be zero if $\bar{x}$ were exact, but otherwise it does a good job of correcting RE in Var. Proof: EFTS (hint: set $\bar{x} \rightarrow \bar{x}+\epsilon$ ).


## Other moments

- Higher moments, like skewness ( $3^{\text {rd }}$ moment) and kurtosis ( $4^{\text {th }}$ moment) are also sometimes used, but can be unreliable.
- Cf. NRiC §14.1.


## Distribution Functions

- A distribution function (DF) $p(x)$ gives the probability of finding a value between $x$ and $x+d x$, e.g., the familiar "normal" (Gaussian) distribution $p(x) d x=\frac{1}{\sqrt{2 \pi}} e^{-x^{2} / 2} d x$.
- The expected mean data value is:

$$
<x>=\frac{\int_{-\infty}^{\infty} x p(x) d x}{\int_{-\infty}^{\infty} p(x) d x} .
$$

- For a discrete DF:

$$
<x>=\frac{\sum_{i} x_{i} p_{i}}{\sum_{i} p_{i}} .
$$

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


## Median

- The median of a DF is the value $x_{\text {med }}$ for which larger and smaller values of $x$ are equally probable:

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

- For discrete values, sort in ascending order $(i=1,2, \ldots, N)$, then:

$$
x_{\mathrm{med}}= \begin{cases}x_{(N+1) / 2}, & \text { if } N \text { is odd } \\ \frac{1}{2}\left(x_{N / 2}+x_{N / 2+1}\right), & \text { if } N \text { is even }\end{cases}
$$

## Mode

- The mode 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 number of standard errors $\sim \sigma / N^{1 / 2}$ between two means.
(b) Compute statistic using nasty formula: probability that the two means are different by chance.
(c) Small numerical value indicates significant difference.
2. $F$-test for significantly different variances.
(a) Compute $F=\operatorname{Var}_{1} / \operatorname{Var}_{2}$ and plug into nasty formula (the distribution of $F$ in the case that the variances are the same - the null hypothesis - is related to the incomplete beta function).
(b) Small value indicates significant difference.

- Given two sets of data, can generalize to a single question: Are the sets drawn from the same DF? E.g., are stars distributed uniformly in the sky? Do two brands of lightbulbs have the same distribution of burn-out times?
- Recall can only disprove (to a certain confidence level), 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 $\chi^{2}$ test. For continuous data, use the KS test. Cf. NRiC §14.3.


## Chi-square ( $\chi^{2}$ ) test

- Suppose have $N_{i}$ events in $i$ th bin but expect $n_{i}$ :

$$
\chi^{2}=\sum_{i} \frac{\left(N_{i}-n_{i}\right)^{2}}{n_{i}} .
$$

- Large value of $\chi^{2}$ indicates unlikely match (i.e., $N_{i}$ 's probably not drawn from population represented by $n_{i}{ }^{\prime}$ s).
- Compute probability $Q\left(\chi^{2} \mid \nu\right)$ from incomplete gamma function, where $\nu$ is the number of degrees of freedom.
* Typically $\nu=N_{B}$, where $N_{B}$ is the number of bins, or $N_{B}-1$, if the $n_{i}$ 's are normalized such that $\sum_{i} n_{i}=\sum_{i} N_{i}$.
* Null hypothesis assumes differences $N_{i}-n_{i}$ are standard normal random variables of unit variance and zero mean.
- For two binned data sets with events $R_{i}$ and $S_{i}$ :

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

- Have sum in denominator, rather than average, because variance of difference of two normal quantities is sum of individual variances.


## Kolmogorov-Smirnov (KS) test

- Appropriate for unbinned distributions of single independent variable.
- From sorted 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 $S_{N}\left(x_{\min }\right)=0, S_{N}\left(x_{\max }\right)=1$.
* Behaviour between $x_{\min }$ and $x_{\max }$ distinguishes distributions.
- Cf. NRiC Fig. 14.3.1.
- Statistic is maximum value of absolute difference between two cumulative DFs.
- To compare data set to known cumulative DF:

$$
D=\max _{x_{\min } \leq x \leq x_{\max }}\left|S_{N}(x)-P(x)\right| .
$$

- To compare two unknown data sets:

$$
D=\max _{x_{\min } \leq x \leq x_{\max }}\left|S_{N_{1}}(x)-S_{N_{2}}(x)\right| .
$$

- Plug $D$ and $N\left(\right.$ or $\left.N_{e}=N_{1} N_{2} /\left(N_{1}+N_{2}\right)\right)$ into nasty formula to get numerical value of significance. As usual, a small value indicates a significant difference.

