STAT 518 --- Nonparametric Density Estimation

• The <u>probability density function</u> (or <u>density</u>) of a continuous random variable *X* describes its probability distribution.

- We denote the density as
- Note that if F(x) is the c.d.f. of X, then

Two important properties of density functions (1) They are always _____:

(2) The total area under a density curve is always _____.

• In real data analysis, we do not know the true density, so we can estimate it using sample data $X_1, X_2, ..., X_n$.

<u>Parametric approach</u>: Assume a specific functional form (e.g., normal, gamma, etc.) for the density and use the sample data to estimate certain ______.

Example: Could assume the density is <u>normal</u> and get sample estimates of _____ and _____.

• The <u>nonparametric</u> approach is to make very <u>few</u> assumptions about the functional form of the density.

Histograms

• A simple density estimator is a <u>histogram</u>.

• If we rescale the heights of each bar so that the <u>total</u> <u>combined area</u> within all the bars is 1, we have a <u>histogram density estimate</u>.

• Assume there are *K* bins, each of width *h*:

Picture (*K* = 5, *h* = 2):

• In general, this histogram is:

where

• The total combined area within all bars is

• The R function hist produces such histograms.

• The choice of bin width *h* determines the number of bins, which can affect the appearance of the estimate.

• A simple rule of thumb for choosing *h* is derived from a normal density:

Let

where

• Note: the sample standard deviation *s* is a consistent estimator of σ , as is *IQR* / 1.34 when the true density is normal.

• In reality, this provides a good initial choice of *h*, which may then be adjusted by trial and error.

• Choosing *h* too small produces <u>many bins</u> and a density estimate that is too _____.

• Choosing *h* too large produces <u>few bins</u> and a density estimate that is ______.

Example 1:

Example 2:

• We could also let the bin width vary across bins, choosing a ______ width in regions where we expect the density to be <u>flatter</u> and a ______ width in regions where we expect the density to be <u>spiky</u>.

Kernel Density Estimation

• An obvious drawback to the histogram density estimate is that it is not _____.

• A <u>kernel density estimate</u> (k.d.e.) produces a smooth estimate and works similarly to the kernel regression method.

• As $n \to \infty$, the k.d.e. will approach the true density f(x) more quickly than the histogram will.

Recall:

- Plug in the e.d.f. for $F(\cdot)$ to obtain:
- This is exactly the same as

with K(u) =

 \rightarrow a kernel estimate with a _____ kernel function.

• However, with the _____ kernel, the resulting density estimate is not smooth.

• Better choices of kernel function $K(\cdot)$ include:

• Let $K(\cdot)$ in the above k.d.e. formula be a standard normal kernel function.

• Then for, say, *h* = 1:

• We see at each point <i>x</i> , the k.d.e. of normal densities, centered at	is the average
• Sample values near <i>x</i> will contribute	
• Sample values far from <i>x</i> will	
Role of the Bandwidth	h
• If h increases, these normal densities h and more \rightarrow	Decome
 If <i>h</i> decreases, these normal densities and → 	become

• Rule of thumb for choosing *h* (again based on the true density being normal):

Let

where

• In reality, this provides a good initial choice of h, which may then be adjusted by trial and error.

• The density function in R produces a kernel density estimate.

Example 1:

Example 2:

• As with kernel regression, kernel density estimators tend to be biased at the left and right edges:

• The k.d.e. also has a tendency to be too flat (not rise or dip enough) in the peaks and valleys of the density.

• An option is to use a bandwidth that <u>varies</u> over the region (being _______ where the density is expected to be flat and ______ where the density is expected to have bumps).