## Sampling Distributions and the Central Limit Theorem

**Recall**: A (univariate) random variable is defined to be a function from a sample space S into the real numbers,  $\mathbb{R}$ . We've talked a lot about distributions of random variables (including multivariate random variables), and we've used these distributions to find probabilities and other quantities like expected values that can be used to make decisions.

Often, we are not interested in the random variable(s) themselves, but instead in functions of the random variable(s), such as the **sample mean**.

Specifically, we can focus on functions of the variables  $Y_1, Y_2, ..., Y_n$  **observed in a random sample** selected from the population of interest.

In such scenarios, we need the probability distribution of these functions of random variables in order to make inferences about population parameters, and, in turn, make decisions.

Definition: (Random Sample)  $Y_1$  can be Discrete or Continuous.  $Y_1, Y_2, ..., Y_n$  are a random sample of size n from a population with distribution (CDF)  $Y_1, Y_2, ..., Y_n$  are:

- independent Durclated
- identically distributed (have the same probability distribution)

This is also called an iid (independent and identically distributed) sequence.

**Example 1**: Let Y be the number of calls received per hour at a customer service center. The observed values of Y in randomly selected 8 hours are given below:

 $y_1 = 3$ ,  $y_2 = 1$ ,  $y_3 = 2$ ,  $y_4 = 1$ ,  $y_5 = 10$ ,  $y_6 = 3$ ,  $y_7 = 0$ ,  $y_8 = 12$ Here,  $y_2 = 1$  is the number of calls received per hour and it can take any value from  $y_1 = 1$  if  $y_2 = 1$  is a Poisson distribution with parameter  $y_1 = 1$  which is the average number of calls per hour. Since these  $y_1 = 1$  hours are randomly selected  $y_1 = 1$  are inclependent here.