AUA CS108, Statistics, Fall 2020 Lecture 25

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Estimation

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We can, of course, calculate the Sample Mean and the Sample Variance of our Dataset. Or, we can plot the Histogram or KDE. But will this give an info about the Population or the process generating the Dataset? Well, no, in general.

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then we think about this as a Sample, and we realize that if we would make another Sampling, we'd get other numbers. Even with the same cars!

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Example: If we consider the weights (in Kg) of 10 persons:

then we make the following model: let X_1 be the weight of the first person (say, the first person we will meet when performing the experiment), X_2 be the weight of the second person,..., X_{10} be the weight of the 10-th person.

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Our Dataset of weights is just one of the possible realizations of $X_1, ..., X_{10}$.

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So, again, having a Dataset $x_1, ..., x_n$, statisticians work with a r.v.s $X_1, X_2, ..., X_n$ to work not only with a particular Sample, but with all possible samples from the Distribution (Process) behind the phenomenon.

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Now, if we have a Random Sample $X_1, ..., X_n$, then, because they are IID, we will have that all X_k -s are coming from the same Distribution:

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In Parametric Statistics, we assume that we have a Random Sample

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and ${\mathcal F}$ is a member of the Parametric Familiy of Distributions:

$$\mathcal{F} = \mathcal{F}_{\theta}, \qquad \theta \in \Theta \subset \mathbb{R}^m.$$

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We will consider one of the main Problems of the Parametric Statistics: Using the observations from our Random Sample, estimate the value of the Parameter θ .

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In this problem, p is **fixed, but unknown**. And our aim will be to estimate p, using our observations $x_1, ..., x_n$.