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The first step in data analysis is to summarize the important features of the data using **descriptive statistics**

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Definition: The **sample mean**, denoted by \overline{x} , of a set of *observations*

$$x_1, x_2, \ldots, x_n$$

is defined as the sum of the data values, divided by the number of values n.

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The mean is the most commonly used measure of location.

The only shortcoming of the mean is that it can be heavily influenced by a single data value that is very different from the others.

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We will assign the house prices, in thousands, to an array in R using the following statement:

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which seems reasonably representative of this collection of seven numbers:

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This value is not very representative of either the new expensive house, or the original houses.

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Because the sand contained a lot of iron, this particular shellfish was reported as being 4% iron, way out of line with the other specimens where we were dealing with parts per million.