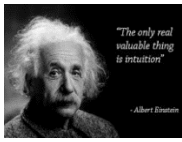


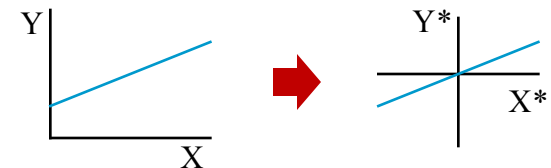


Transformation #6: **Centering**



Centering: Intuition

- There are times when centering one or more variables around their means is useful.
- For example, maybe the **intercept is meaningless** because **x cannot be 0** (e.g., predicting cholesterol based on weight → nobody has 0 weight). Centering the variable shows the effect of x when x is at its mean value (not zero).
- Also, when computing **interaction** variables of **2 continuous** variables x_1 and x_2 , the resulting interaction term $x_1 * x_2$ is **problematic** for a number of reasons:
 - **Scale invariance** – changing the scale of x_1 or x_2 (e.g., from feet to meters) will change the main effect size
 - The product of $x_1 * x_2$ may generate severe **multicollinearity**
- **Centering** x_1 , x_2 and y with respect to their means helps → $x_1^* = x_1 - \bar{x}_1$; $x_2^* = x_2 - \bar{x}_2$
- This is **equivalent** to **shifting** the Y and X **axes** to the Y and X means



Tips

```
Centered.data = data.frame(scale(dataName,  
                                center=TRUE, scale=FALSE))
```

 → Using
the `scale()` function with `center=TRUE` to center an entire data
frame with numeric values

```
dataName$variableName = scale(Boston$medv,  
                                center=TRUE, scale=FALSE)
```

→ To center a specific numeric column in the data set



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