

MA_615_HW_Formating Exercise

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Extract From:

Bradley Efron and Trevor Hastie

Computer Age Statistical Inference: Algorithms, Evidence and Data Science

Cambridge University Press, 2016

https://web.stanford.edu/~hastie/CASI_files/PDF/casi.pdf

Modern Bayesian practice uses various strategies to construct an appropriate “prior” $g(\mu)$ in the absense of prior experience, leaving many statisticians unconvinced by the resulting Bayesian inferences. Our second example illustrates the difficulty.

Table 3.1 Scores from two tests taken by 22 students, **mechanics** and **vectors**

| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 |
|------------------|----|----|----|----|----|----|----|----|----|----|----|
| mechanics | 7 | 44 | 49 | 59 | 34 | 46 | 0 | 32 | 49 | 52 | 44 |
| vectors | 51 | 69 | 41 | 70 | 42 | 40 | 40 | 45 | 57 | 64 | 61 |

| | 12 | 13 | 14 | 15 | 16 | 17 | 18 | 19 | 20 | 21 | 22 |
|------------------|----|----|----|----|----|----|----|----|----|----|----|
| mechanics | 36 | 42 | 5 | 22 | 18 | 41 | 48 | 31 | 42 | 46 | 63 |
| vectors | 59 | 60 | 30 | 58 | 51 | 63 | 38 | 42 | 69 | 49 | 63 |

Table 3.1 shows the scores on two tests, **mechanics** and **vectors**, achieved by $n = 22$ students. The sample correlation coefficient between the two scores is $\hat{\theta} = 0.498$,

$$\hat{\theta} = \frac{\sum_{i=1}^{22} (m_i - \bar{m})(v_i - \bar{v})}{\left[\sum_{i=1}^{22} (m_i - \bar{m})^2 \sum_{i=1}^{22} (v_i - \bar{v})^2 \right]^{1/2}}$$

with m and v short for **mechanics** and **vectors**, \bar{m} and \bar{v} their average ages.