

Bayes Theorem – Conceptual Example

An AI model you developed allows you to select potential winning funds based on certain properties. Published parameters for this model include:

- Sensitivity, which is nothing but the ability to flag winning funds from all funds that have beaten the benchmark. Sensitivity for the model is quite high: 85% [$P(W_{AI}|W_A)$].
- False positives: The model detects about 15% of the losers as winners [$P(W_{AI}|\sim W_A)$].

You know one critical thing from published data: Over a one-year period, just 65% [$P(W_A)$] of active funds beat their benchmark. Given these numbers, what is the probability that the next fund categorized by the AI software is a winner?

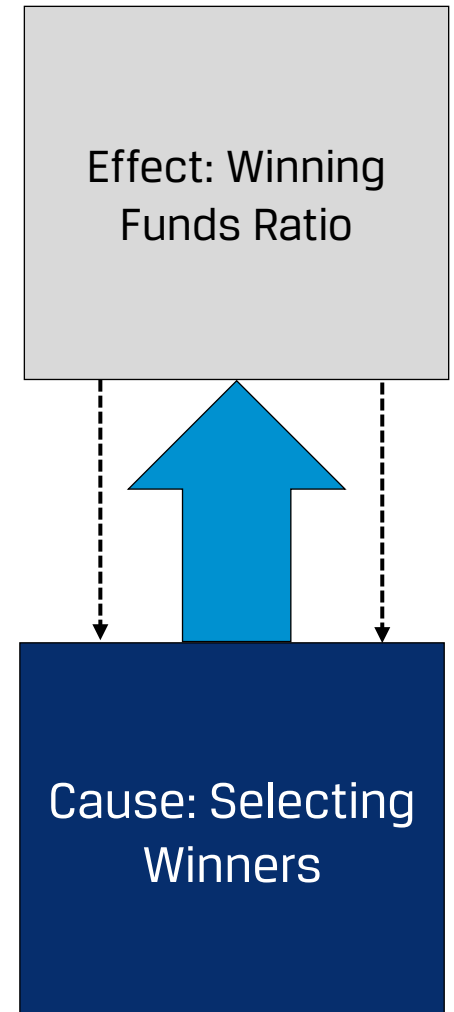
Using Bayes:

$$P(W_A|W_{AI}) = \frac{P(W_{AI}|W_A)}{P(W_{AI})} * P(W_A)$$

Where:

$$P(W_{AI}) = P(W_{AI}|W_A) * P(W_A) + P(W_{AI}|\sim W_A) * P(\sim W_A)$$

Is your updated probability answer 91%? If yes, you have made the correct substitutions.





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The substitutions are:

$$P(W_A) = 65\%$$

$$P(W_{AI}|W_A) = \text{Sensitivity of Test} = 85\%$$

$$P(W_{AI}|\sim W_A) = \text{False Positives} = 15\%$$

$$P(W_{AI}) = P(W_{AI}|W_A) * P(W_A) + P(W_{AI}|\sim W_A) * P(\sim W_A)$$

$$\text{That is, } P(W_{AI}) = 0.85 * 0.65 + 0.15 * (1 - 0.65) = 0.605$$

$$\text{Likelihood ratio} = \frac{P(W_{AI}|W_A)}{P(W_{AI})} = 0.85/0.605 = 1.40$$

$$\text{Modified probability, } P(W_A|W_{AI}) = \text{Likelihood ratio} * P(W_A) = 1.4 * 0.65 = 91\%$$