

1. We consider the training examples shown in the following table for a binary classification problem.

Instance	a_1	a_2	a_3	Target Class
1	T	T	1	+
2	T	T	6	+
3	T	F	5	-
4	F	F	4	+
5	F	T	7	-
6	F	T	3	-
7	F	F	8	-
8	T	F	7	+
9	F	T	5	-

- a) What is the original entropy of this set of training instances?

The original entropy is $-\frac{4}{9}\log_2 \frac{4}{9} - \frac{5}{9}\log_2 \frac{5}{9} = 0.991$ bit.

- b) What are the information gains when a_1 and a_2 are used for partitioning the training set respectively?

After splitting on a_1 , the entropy becomes

$$\frac{4}{9}\left(-\frac{3}{4}\log_2 \frac{3}{4} - \frac{1}{4}\log_2 \frac{1}{4}\right) + \frac{5}{9}\left(-\frac{1}{5}\log_2 \frac{1}{5} - \frac{4}{5}\log_2 \frac{4}{5}\right) = 0.762 \text{ bit.}$$

As a result,

$$\text{gain}(a_1) = 0.991 - 0.762 = 0.229 \text{ bit.}$$

After splitting on a_2 , the entropy becomes

$$\frac{5}{9}(-\frac{2}{5}\log_2 \frac{2}{5} - \frac{3}{5}\log_2 \frac{3}{5}) + \frac{4}{9}(-\frac{2}{4}\log_2 \frac{2}{4} - \frac{2}{4}\log_2 \frac{2}{4}) = 0.984 \text{ bit.}$$

As a result,

$$\text{gain}(a_2) = 0.991 - 0.984 = 0.007 \text{ bit.}$$