

Bias

→ Bias - If your model has high bias it means that it has high training error rate and if it has low Bias it means it has low training rate.

→ Algorithms which have high Bias learn faster and are easier to understand but generally less flexible. In turn they have lower predictive performance on complex problems that fail to meet the simplifying assumption of the algorithm bias.

Variance

→ Variance - If your model has high variance it means that it has high testing error rate and if it has low Variance it means it has low testing error rate.

→ ML Algos that have a high variance are strongly influenced by the specifics of the training data. This means that the specifics of the training have influences the number and type of parameters used to characterize the mapping function.

TRADE OFF

→ Increasing the bias will decrease the variance and vice versa. There is a trade-off at play between these two concerns and the algorithm you choose and the way you choose to configure them are finding different balances in this trade-off.

	Class 1	Class 2
Class 1	50	30
Class 2	40	60

1) Precision:

Assuming Class 1 to be positive & Class 2 to be negative

$$P_1 \Rightarrow \frac{TP}{TP + FP} = \frac{50}{80}$$

P₂ → Assuming Class 2 to be positive and Class 1 to be negative

$$\Rightarrow \frac{TP}{TP + FN} = \frac{40}{100}$$

$$2) Recall = \frac{TP}{TP + FN}$$

a) When Class 1 is positive

$$\Rightarrow \frac{50}{90} = \frac{5}{9}$$

b) When Class 2 is positive

$$\therefore \frac{40}{90}$$

- To reduce bias → you can increase the model complexity and have more parameters to tune & allow for more learning.
- Reducing variance → add more data & increase regularization to prevent overfitting
- from overfitting → Reduce dimensionality by removing redundant data.

→

3)

$$3) F1_{\text{swr}} = 2 \cdot \frac{\text{Recall} * \text{Precision}}{\text{Recall} + \text{Precision}}$$

$$= 2 \cdot \frac{50}{90} * \frac{50}{80}$$

$$\frac{50}{90} + \frac{50}{80}$$

$$\Rightarrow 2 \cdot \frac{5}{9} * \frac{5}{8}$$

$$\frac{85}{72}$$

$$\Rightarrow \frac{25}{36} \div \frac{85}{72}$$

$$\Rightarrow \frac{25}{36} \times \frac{72}{85} = \frac{5}{9}$$

$$\text{Grim} (\text{Win } \pi) = 0.81$$

W Y N

W 3 0

S 0 1

$$\text{highest gain} = \text{win}^+$$

Taking sunny sit

O	T	H	W	P
S	H	H	W	N
S	U	H	S	N
S	M	U	W	N
S	C	N	W	Y

$$E_{\text{sum}} = -\frac{1}{2} \log \frac{1}{2} - \frac{3}{2} \log \frac{3}{4} = 0.81$$

$$E_{\text{weak}} = 0.918$$

Wind	Y	N
W	1	2
S	0	1

$$E_{\text{strong}} = 0$$

$$\text{Grim} = 0.811 - \left(\frac{3}{2} \times 0.918 \right) = 0.1225$$

$$\text{Grim temp} = 0.811$$

+	Y	N	B
H	6	2	0
M	0	1	0
C	1	0	0

$$\text{Grim humid} = 0.811$$

H	Y	N
high	0	3
normal	1	0

Out look

sun

overcast

rain

Temperature

Yes

Wind

hot

cold

mild

No

Yes

9

10

11

12

13

strong

weak

No

Yes

$$E_n = -\frac{1}{3} \log \frac{1}{3} - \frac{2}{3} \log \frac{2}{3} = 0.91$$

$$E_m = -\frac{2}{3} \log \frac{2}{3} - \frac{1}{3} \log \frac{1}{3} = 0.91$$

$$E_c = -\frac{3}{4} \log \frac{3}{4} - \frac{1}{4} \log \frac{1}{4} = 0.81$$

$$\text{Grain (temp)} = 0.971 - (0.3 \times 0.91) - (0.3 \times 0.91) - (0.3 \times 0.81)$$

$$= 0.096$$

$$E_h = -\frac{2}{5} \log \frac{2}{5} - \frac{3}{5} \log \frac{3}{5} = 0.971$$

$$E_n = -\frac{4}{5} \log \frac{4}{5} - \frac{1}{5} \log \frac{1}{5} = 0.72$$

$$\text{Grain (humidity)} = 0.971 - (0.5 \times 0.97) - (0.5 \times 0.72)$$

$$= 0.125$$

$$E_s = -\frac{1}{3} \log \frac{1}{3} - \frac{2}{3} \log \frac{2}{3} = 0.91$$

$$E_w = -\frac{5}{7} \log \frac{5}{7} - \frac{2}{7} \log \frac{2}{7} = 0.86$$

$$\text{Grain (wind)} = 0.971 - (0.3 \times 0.91) - (0.7 \times 0.86)$$

$$= 0.091$$

$$3) \text{ Entropy} = -P_{\text{in}} \log_2 \frac{P_{\text{in}}}{P_{\text{in}}} - P_{\text{out}} \log_2 \frac{P_{\text{out}}}{P_{\text{in}}} \rightarrow$$

$$\rightarrow = \frac{6}{10} \log_2 \frac{6}{10} - \frac{4}{10} \log_2 \frac{4}{10} = 0.971$$

$$\rightarrow \bar{t}_{\text{sum}} = \frac{1}{3} \log \frac{1}{3} - \frac{3}{3} \log \frac{3}{3} = 0.811$$

Summer	1	3
Forecast	2	0
Rain	3	1

$$\bar{t}_{\text{min}} = -\frac{3}{2} \log \frac{2}{2} = 0$$

$$\text{Crisis(outlook)} = 0.971 - \left(\frac{6}{10} \cdot 0.811 \right) - \left(0.2 \cdot 0 \right) - \left(0.1 \cdot 0.81 \right)$$

$$= 0.32$$

→ Highest Info Gain = Outlook

→ Taking Rain side

T	H	W	P
M	H	W	Y
C	N	W	Y
C	N	S	N
M	N	W	Y

$$\rightarrow E_{\text{Gain}} = H - \frac{1}{3} \cdot \log \frac{18}{3} + \frac{3}{3} \cdot \log \frac{3}{3} = 0.81$$

$$E_M = 0$$

$$E_C = 1$$

T	2	Y	N
M	2		0
C	1		1

$$\text{Gain - temp} : 0.81 - \left(\frac{2}{3} \times 1 \right) = 0.31$$

$$E_H = 0$$

$$E_N = -\frac{2}{3} \log \frac{2}{3} - \frac{1}{3} \log \frac{1}{3}$$

$$= 0.918$$

H	2	Y	N
H	1		0
N	2		1

$$\text{Gain} = 0.81 \left(1 - \left(\frac{3}{2} \times 0.918 \right) \right)$$

$$= 0.1225$$