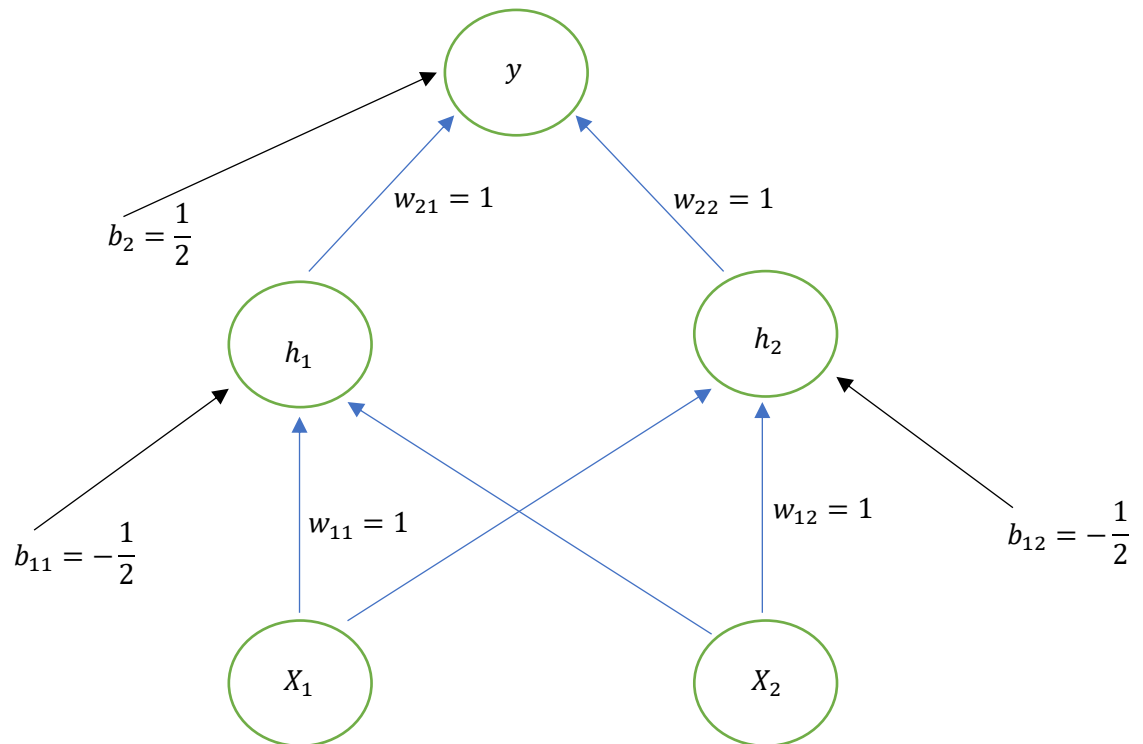


Solutions

Ans1:



$$w_1 = [1 \quad 1]$$

$$b_1 = \left[-\frac{1}{2} \quad -\frac{1}{2}\right]$$

$$w_2 = [1 \quad 1]$$

$$b_2 = \frac{1}{2}$$

Activation Function: $f(h) = \begin{cases} 1, & \text{if } w_x + b > 0 \\ 0, & \text{else} \end{cases}$

X_1	X_2	h_1	h_2	y
1	1	1	1	1
1	0	1	1	1
0	1	1	1	1
0	0	0	0	0

Ans2:

(1)

$Pr(C) \rightarrow$ Prior Probability: The prior probability of an event is the probability of the event computed before the collection of new data. One begins with a prior probability of an event and revises it in the light of new data.

$Pr(X|C) \rightarrow$ Class Conditional Probability: Conditional probability is a measure of probability to an event that is occurring, given other event has already occurred.

$Pr(C|X) \rightarrow$ Discriminative Model: Also referred as conditional models. This is a class of logistical models used for classification or regression. It should distinguish decision boundary through observed data.

$Pr(C), Pr(X|C) \rightarrow$ Generative Model: A generative model describes how a dataset is generated, in terms of a probabilistic model. By sampling this model, we generate data.

(2)

Beam Size $k = 1$: $\langle \text{BOS} \rangle$ Montreal a great playground.

Beam Size $k = 2$: $\langle \text{BOS} \rangle$ Montreal, giant playground.

Pros: More words would be traversed and chosen. Therefore, a higher probability to get better results.

Cons: More computational resources and more memory are required.

Ans3:

(1) **Trigram model of language modelling:** A trigram model restricts the conditional information to the previous two words. Using this method, the conditional distribution can calculate a certain word combination frequency based on the previous two words.

(2) **Procedure of 5-fold cross-validation:** Dataset is split into 5 sets. Each one set is taken as test set by turn while other four are training sets. Eventually, we'll have 5 accuracies and the average is the accuracy of 5-fold cross-validation.

Pros: Avoiding the randomness and bias by training and testing all the data.

(3) **Bagging:** Bootstraps the training set, estimates many copies of a model on the resulting samples and then averages their predictions.

Boosting: Sequentially reweights the training samples forcing the model to attend the training examples with higher loss.

Stacking: Used a separate validation set to train a meta-model that combines predictions of multiple models.