#### Exercise 3

Probabilities of characters in normal text

$$Pr(Y = j) = 0.015$$
  
 $Pr(Y = k) = 0.045$   
 $\Rightarrow Pr((Y = j) \lor (Y = k)) = Pr(Y = j) + Pr(Y = k) = 0.06$ 

#### Part (a)

The most basic classifier we can choose is the classifier on based only on prior probabilities. Since Pr(Y = j)Pr(Y = k), let this classifier be  $f(x) = k \ \forall x \in X$ .

Calculating risk: 
$$\mathcal{R}(f) = \Pr(Y = j \mid ((Y = j) \land (Y = k))) = \frac{\Pr(Y = j)}{\Pr((Y = j) \lor (Y = k))} = \frac{1}{4} < 0.3.$$

Calculating risk:  $\mathcal{R}(f) = \Pr(Y = j \mid ((Y = j) \land (Y = k))) = \frac{\Pr(Y = j)}{\Pr((Y = j) \lor (Y = k))} = \frac{1}{4} < 0.3$ . Thus, the most basic classifier has a risk lower than the classifier being sold. So we would not buy the classifier being sold.

The lowest probability of error if we do not look at the input data is 0.25.

## Part (b)

Again let us choose the most basic classifier, which just predicts  $\sim j$  without regard to the input data.

$$\mathcal{R}(f) = \Pr(Y = j) = 0.015 < 0.2$$

By the same logic of Part (a), we would not buy the classifier being sold.

#### Exercise 5

Notations

S = message is actually spam

L =message is actually legitimate

 $\hat{S} = \text{message is classified as spam}$ 

 $\hat{L} = \text{message}$  is classified as legitimate

# Part (a)

From the given data, we fill up the confusion matrix.

	$\hat{S}$	Ĺ	
$\overline{S}$	$0.51 (= 0.85 \times 0.6)$	0.09 (= 0.6 - 0.51)	0.6
L	$0.02 (= 0.4 \times 0.05)$	0.38 (= 0.4 - 0.02)	0.4 (= 1 - 0.6)
	0.53 (= 0.51 + 0.02)	0.47 (= 1 - 0.53)	1

From the above table, we can conclude the following<sup>1</sup>:

- False positive =  $\Pr(\hat{S} \cap L) = 0.02$
- False negative =  $\Pr(\hat{L} \cap S) = 0.09$
- Average error =  $\Pr\left(\hat{S} \cap L\right)$  +  $\Pr\left(\hat{L} \cap S\right)$  = 0.02 + 0.09 = 0.11

## Part (b)

- A classifier which simply classifies everything as legitimate.
- A classifier which simply classifies everything as spam.

<sup>&</sup>lt;sup>1</sup>Definition from "Counting performance mea ure (2)" page of the lide

# Part (c)

