

## 1. Using the number of vowels to detect ciphertext rectangles

(In English approximately 40% of plaintext consists of vowels).

總共有 77 個字母，可以拆成 7\*11 和 11\*7 的 rectangle

然後分別計算兩個 rectangle 的 variance (每一橫行理應有 40% 為母音)

如下圖，可以發現 11\*7 的 average difference 比較小

可以推出 rectangle 應該為 11\*7

```
rectangle 7*11
E R H N E R C R N E C   difference: 1.4
O N H G T A S W Y Y E   difference: 1.4
E P E N O E E K A I I   difference: 3.6
Y S T D C M U D R C E   difference: 2.4
E E Y D R H S R A N T   difference: 1.4
G C H D A T I I N T U   difference: 0.4
T E S D E E A R U T S   difference: 0.6
average difference: 1.6

rectangle 11*7
E E G A E R C   difference: 1.2
O C N E U N N   difference: 0.2
E E D R S Y T   difference: 0.8
Y H D A I A T   difference: 0.2
E H D E A R C   difference: 0.2
G E D M R A E   difference: 0.2
T T E H W N I   difference: 0.8
R Y T T K U E   difference: 0.8
N H O E D E T   difference: 0.2
P S C C R Y U   difference: 1.8
S N R S I I S   difference: 0.8
average difference: 0.6545454545454545
```

## 2. Using plaintext bigrams and trigrams to calculate conditional probabilities for Markov decision processing (MDP).

我將作業提供的 plaintext 用來訓練

分別以 bigrams 和 trigrams 的形式記錄各個次數(也就是兩個一數或三個一數)

將這些次數用來下一步計算條件機率的參考資料(conditional probabilities for MDP )

```
11 ▼ def train(msg, trigram, bigram):
12 ▼     for i in range(len(msg)-2):
13         if (msg[i:i+3] not in trigram):
14             trigram.update({msg[i:i+3]:1})
15 ▼     else:
16         tmp = trigram[msg[i:i+3]]
17         trigram.update({msg[i:i+3]:tmp+1})
18         if (msg[i:i+2] not in bigram):
19             bigram.update({msg[i:i+2]:1})
20 ▼     else:
21         tmp = bigram[msg[i:i+2]]
22         bigram.update({msg[i:i+2]:tmp+1})
```

### 3. Using MDP to recover columnar transposition ciphers.

接著上述得到了 11\*7 的 rectangle

根據提示前面 3 個字為 GRE

所以可以先將開頭為 G、R 兩直欄換到第一、二欄

```
107 # training
108 ▼ for i in range(col-2):
109     #test column
110     change=i+2
111     Pr=0
112 ▼     for k in range(i+2, col):
113         prob=0
114 ▼         for j in range(row):
115             word2=ansRect[i][j]+ansRect[i+1][j]
116             word3=word2+ansRect[k][j]
117             if ((word2 not in bigram) or (word3 not in trigram)):
118                 prob += 0
119             else:
120                 prob += math.log(26*(trigram[word3]/bigram[word2]))
121 ▼         if(prob>Pr):
122             Pr = prob
123             change = k
124     #swap column
125     for j in range(row):
126         ansRect[i+2][j], ansRect[change][j] = ansRect[change][j], ansRect[i+2][j]
```

如上，接著利用計算條件機率的公式： $\log 26 * (\#trigram / \#bigram)$

也就是看到 bigram 的兩個字母時，三個連續字母是 trigram 的機率是多少

接著把各欄條件機率加總起來

看哪一直欄的加總條件機率最高，就能得出下一直欄應為哪一欄

往後做到最後一欄可以得出答案，如下

G	R	E	E	C	E	A
N	N	O	U	N	C	E
D	Y	E	S	T	E	R
D	A	Y	I	T	H	A
D	R	E	A	C	H	E
D	A	G	R	E	E	M
E	N	T	W	I	T	H
T	U	R	K	E	Y	T
O	E	N	D	T	H	E
C	Y	P	R	U	S	C
R	I	S	I	S	N	S