

Problem 1

(a) AAA since $P(O_n = A | S_n = T) = P_{AT}$

Prob. of one T

$P(\text{One T} | \text{observe} = \text{AAA})$

$$= \frac{P(O = \text{AAA}, S = \text{ATA}) + P(O = \text{AAA}, S = \text{TAA}) + P(O = \text{AAA}, S = \text{AAT})}{P(O = \text{AAA})}$$

$$= \frac{P(O = \text{AAA}, S = \text{AAA}) + P(O = \text{AAA}, S = \text{AAT}) + P(O = \text{AAA}, S = \text{ATA}) + P(O = \text{AAA}, S = \text{TAA}) + P(O = \text{AAA}, S = \text{ATT}) + P(O = \text{AAA}, S = \text{TAT}) + P(O = \text{AAA}, S = \text{TTT})}{P(O = \text{AAA})}$$

$$P(O = \text{AAA}, S = \text{ATA}) = P(O_1 = A | S_1 = A) P(O_2 = A | S_2 = T) P(O_3 = A | S_3 = A) P(S = \text{ATA})$$

Assume every sequence has the same probability in its unobserved state,

$P(S = \text{ATA})$ would be cancelled out

$$= P_A \cdot P_{AT} \cdot P_A = P_A^2 P_{AT}$$

$$\therefore P(\text{One T} | \text{observe} = \text{AAA})$$

$$= \frac{3 P_A^2 P_{AT}}{P_A^3 + P_A^2 P_{AT} + P_A^2 P_{AT} + P_A^2 P_{AT} + P_A \cdot P_{AT}^2 + P_A \cdot P_{AT}^2 + P_A \cdot P_{AT}^2 + P_{AT}^3}$$

$$= \frac{3 P_A^2 P_{AT}}{P_A^3 + 3 P_A^2 P_{AT} + 3 P_A P_{AT}^2 + P_{AT}^3}$$

$$(b) P(O = \text{AAA} | S = \text{TTT}) = \frac{P(O = \text{AAA}, S = \text{TTT})}{P(O = \text{AAA})}$$

$$= \frac{P_{AT}^3}{P_A^3 + 3 P_A^2 P_{AT} + 3 P_A P_{AT}^2 + P_{AT}^3}$$

$$= P_A P_{AT} P_{AT} + P_A P_{AT} P_{AT} + P_{AT} P_{AT} P_{AT} + P_{AT} P_{AT} P_{AT} + P_A \cdot P_{AT} \cdot P_{AT} + P_{AT} P_{AT} P_{AT} + P_{AT} P_{AT} P_{AT} + P_{AT} P_{AT} P_{AT}$$

$$(c) P(O = \text{AGC} | S = \text{ACC}) = \frac{P(O = \text{AGC}, S = \text{ACC})}{P(O = \text{AGC})}$$

$$= (P_A \cdot P_{AG} \cdot P_C) \text{ divided by}$$

$$P(O = \text{AGC}) = P(O = \text{AGC}, S = \text{ACC}) + P(O = \text{AGC}, S = \text{AGC}) + P(O = \text{AGC}, S = \text{TGC}) + P(O = \text{AGC}, S = \text{TCC}) + P(O = \text{AGC}, S = \text{ACG}) + P(O = \text{AGC}, S = \text{TGC}) + P(O = \text{AGC}, S = \text{TCC}) + P(O = \text{AGC}, S = \text{AGC})$$

(d) AAA exactly one error

would be same as (a)

$$\frac{3 P_A^2 P_{AT}}{P_A^3 + 3 P_A^2 P_{AT} + 3 P_A P_{AT}^2 + P_{AT}^3}$$

(e) AAA at least one error

$$P(O = AAA \mid \text{at least one error}) = 1 - P(O = AAA \mid \text{no error})$$

$$= 1 - \frac{P(O = AAA, S = AAA)}{P(O = AAA)}$$

$$= 1 - \frac{P_A^3}{P_A^3 + 3 P_A^2 P_{AT} + 3 P_A P_{AT}^2 + P_{AT}^3}$$

(f) observed sequence AAA.

The probability that it is correct = the probability of no error.

$$P(\text{no errors}) = P(S = AAA, E = 000 | O = AAA)$$

$$= P(S_1 = A) \cdot P(S_2 = A) \cdot P(S_3 = A) \cdot P(E_3 = 0 | E_2 = 0) \cdot P(E_2 = 0 | E_1 = 0) \cdot P(E_1 = 0) \cdot P(O_1 = A) \cdot P(O_2 = A) \cdot P(O_3 = A)$$

$$P(O = AAA)$$

$$= \frac{P(S_1 = A) \cdot P(S_2 = A) \cdot P(S_3 = A) \cdot (1 - PE)^3 \cdot P_A^3}{P_A^3 + 3P_A^2 P_{AT} + 3P_A P_{AT}^2 + P_{AT}^3}$$

(g) exactly 2 errors AAA

$$P(ATT) = P(S = ATT, E = 011 | O = AAA)$$

$$= P(S_1 = A) \cdot P(S_2 = A) \cdot P(S_3 = A) \cdot P(E_3 = 1 | E_2 = 1) \cdot P(E_2 = 1 | E_1 = 0) \cdot P(E_1 = 0) \cdot P(O_1 = A)^2 \cdot P(O_2 = A)^2$$

$$P(O = AAA)$$

$$= \frac{P(S_1 = A) \cdot P(S_2 = A) \cdot P(S_3 = A) \cdot 2PE \cdot PE \cdot (1 - PE) \cdot P_A^3}{P_A^3 + 3P_A^2 P_{AT} + 3P_A P_{AT}^2 + P_{AT}^3}$$

$$P(TTA) = \frac{P(S_1 = T) \cdot P(S_2 = T) \cdot P(S_3 = A) \cdot (1 - 2PE) \cdot (2PE) \cdot PE \cdot P_A^3}{P_A^3 + 3P_A^2 P_{AT} + 3P_A P_{AT}^2 + P_{AT}^3}$$

$$P(TAT) = \frac{P(S_1 = T) \cdot P(S_2 = A) \cdot P(S_3 = T) \cdot (PE) \cdot (1 - 2PE) \cdot (PE) \cdot P_A^3}{P_A^3 + 3P_A^2 P_{AT} + 3P_A P_{AT}^2 + P_{AT}^3}$$

$$P(\text{exactly 2 errors} | O = AAA) = P(ATT) + P(TTA) + P(TAT)$$

$$(h) P(E_n = 1) = P(E_n | E_{n-1}) \cdot P(E_{n-1} | E_{n-2}) \cdot P(E_{n-2} | E_{n-3}) \dots \cdot P(E_1)$$

let e denotes the probability of each conditional probability. eg $\sum_0^1 e_1 = P(E_1)$
 $\sum_0^1 e_2 = P(E_2 | E_1)$ etc

$$\therefore P(E_n = 1) = 1 \cdot \sum_0^1 e_{n-1} \cdot \sum_0^1 e_{n-2} \cdot \sum_0^1 e_{n-3} \dots \sum_0^1 e_1$$

$$(i) P(\text{at least 1 error}) = 1 - P(\text{no error}) \\ = 1 - (1 - PE)^n$$

Problem 2

- (a) since $P(S_i \in (1, \frac{1}{2}p_i)) \propto \frac{1}{p_i}$
and $P(S_i \in (\frac{1}{2}p_i + 1, p_i)) \propto \frac{3}{p_i}$

The probability of a fragment starting in the first half of the transcript is $\frac{p_i}{2} (c \cdot \frac{1}{p_i}) + \frac{p_i}{2} (c \cdot \frac{3}{p_i}) = 1$

$$\therefore P(S_i \in (1, \frac{1}{2}p_i)) = \frac{1}{2} \cdot \frac{1}{p_i} = \frac{1}{2p_i} \quad \left| \frac{p_i}{2} \cdot \frac{c}{p_i} + \frac{p_i}{2} \cdot \frac{3c}{p_i} = 1 \Rightarrow \frac{c}{2} + \frac{3c}{2} = 1 \Rightarrow \frac{4c}{2} = 1 \Rightarrow c = \frac{1}{2} \right.$$

- (b) The probability of a fragment starting in the second half of the transcript is $P(S_i \in (\frac{1}{2}p_i + 1, p_i)) = \frac{1}{2} \cdot \frac{3}{p_i} = \frac{3}{2p_i}$

(c) $E[\tilde{p}_i] = p_i - E[F] + 1$
 $E[F | p_i] = \sum_{f=1}^{p_i} f \cdot P(F=f)$

$0=0$ First half $E[F | p_i] = \sum_{f=1}^{p_i-s+1} f \cdot P(F=f)$
 $1 \sim \frac{1}{2}p_i \quad \frac{1}{2}p_i+1 \sim p_i$ $E[\tilde{p}_i] = p_i - \sum_{f=1}^{p_i-s+1} f \cdot P(F=f) + 1$

$0=1$ Second half $E[\tilde{p}_i] = p_i - \sum_{f=\frac{1}{2}p_i+1}^{p_i-s+1} f \cdot P(F=f) + 1$
 $\frac{1}{2}p_i+1 \sim p_i \quad 1 \sim \frac{1}{2}p_i$ first half $E[\tilde{p}_i] = p_i - \sum_{f=\frac{1}{2}p_i+1}^{p_i-s+1} f \cdot P(F=f) + 1$

second half $E[\tilde{p}_i] = p_i - \sum_{f=1}^{p_i-s+1} f \cdot P(F=f) + 1$

\therefore The effective length of any distribution F would be

$$E[\tilde{p}_i] = p_i - \sum_{i=1}^{p_i} \sum_{o=0}^{\frac{1}{2}p_i-s+1} \sum_{l=1}^{p_i-s+1} f \cdot P(F=f | S=s, O=o) \cdot P(S) \cdot P(O) + 1$$

```
#E[F|l] = sum of f = 1 to li of f * P(F = f)
import math
def effective_length(transcript_length, miu, sigma):
    expected_F_on_l = 0
    sum = 0
    for f in range(1, transcript_length + 1):
        sum += (math.exp((f - miu)**2 / (2 * sigma**2)) * (-1))

    normalizing_const = 1.0 / sum

    for f in range(1, transcript_length + 1):
        expected_F_on_l += f * normalizing_const * (math.exp((f - miu)**2 / (2 * sigma**2)) * (-1))

    expected_effective_length = transcript_length - expected_F_on_l + 1

    return expected_effective_length

Python

#(a) li = 1000,  $\mu = 200$ ,  $\sigma = 20$ ,
print(effective_length(1000, 200, 20))

Python

... 800.9999999999999

#(b) li = 1000,  $\mu = 200$ ,  $\sigma = 100$ .
print(effective_length(1000, 200, 100))

Python

... 795.418280719455
```

(e) The values decreased a little. It's a reasonable behavior b/c as value of σ gets bigger, the value of the probability mass function will get smaller.