

$$(a) P(a, b | c, d)$$

$$= \frac{P(a, b, c, d)}{P(c, d)}$$

$$= \frac{P(a) P(b|a) P(c|a,b) P(d|b,c)}{\sum_a \sum_b P(a, b, c, d)}$$

$$= \frac{P(a) P(b|a) P(c|a,b) P(d|b,c)}{\sum_a \sum_b P(a) P(b|a) P(c|a,b) P(d|b,c)}$$

$$(b) P(a | c, d)$$

$$= \sum_b P(a, b | c, d)$$

$$P(b | c, d)$$

$$= \sum_a P(a, b | c, d)$$

$$(c) \quad L = \sum_t^T \log P(C=c_t, D=d_t)$$

$$= \sum_t^T \log \sum_A \sum_B P(A=a, B=b, c=c_t, d=d_t)$$

$$= \sum_t^T \log \sum_A \sum_B P(A=a) P(B=b | A=a) P(C=c_t | A=a, B=b) P(D=d_t | B=b, C=c_t)$$

$$= \sum_t^T \log \sum_a \sum_b P(a) P(b | a) P(c_t | a, b) P(d_t | b, c_t)$$

$$P(A=a) \leftarrow \frac{1}{T} \sum_t^T P(a | c_t, d_t)$$

$$P(B=b | A=a) \leftarrow \frac{\sum_t^T P(a, b | c_t, d_t)}{\sum_t^T P(a | c_t, d_t)}$$

$$P(C=c | A=a, B=b) \leftarrow \frac{\sum_t^T P(a, b | c_t, d_t) I(c, c_t)}{P(a, b | c_t, d_t)}$$

$$P(D=d | B=b, C=c) \leftarrow \frac{\sum_t^T P(b | c_t, d_t) I(c, c_t) (d, d_t)}{P(b | c_t, d_t) I(c, c_t)}$$

(a)

$$\begin{aligned} P(Y=1 | X) &= \sum_{Z \in \{0,1\}^n} P(Y=1, Z | X) \\ &= \sum_{Z \in \{0,1\}^n} P(Y=1, Z) \cdot P(Z | X) \\ &= \sum_{Z \in \{0,1\}^n} (1 - I(Z, \{0\}^n)) P(Z | X) \\ &= \sum_{Z \in \{0,1\}^n} P(Z | X) - P(Z \in \{0\}^n | X) \\ &= 1 - \prod_i^n P(Z_i = 0 | X_i) \\ &= 1 - \prod_i^n (1 - p_i)^{x_i} \end{aligned}$$

(b)

$$P(z_i=1, x_i=1 | x, y)$$

$$= I(x_i=1) P(z_i=1 | x, y)$$

$$= I(x_i=1) \left( \frac{P(y|x, z_i=1) P(z_i=1 | x)}{P(y|x)} \right)$$

$$= I(x_i=1) \left( \frac{I(y=1) P(z_i=1 | x_i)}{P(y|x)} \right)$$

$$= I(x_i=1) \left( \frac{I(y=1) I(x_i=1) p_i}{P(y|x)} \right)$$

$$= I(x_i=1) \left( \frac{I(y=1) p_i}{P(y|x)} \right)$$

∴  $P(y|x)$  in denominator, so only do  $P(y=1|x)$

$$\Rightarrow \frac{x_i y_i p_i}{P(y=1|x)}$$

$$= \frac{x_i y_i p_i}{1 - \prod_j ((1-p_j)^{x_j})}$$

$$(C) P(z_i | X_i) \leftarrow \frac{\sum_t P(z_i, x_i | X, Y)}{\sum_t P(x_i | X, Y)}$$

Only non-zero  $X$ , therefore non-zero  $Z$  is going to be contributing to  $P(z_i | X_i)$  [denominator non-zero]

$$\Rightarrow P(z_i=1 | X_i=1) \rightarrow \frac{\sum_t P(z_i=1, x_i=1 | X, Y)}{\sum_t I(X_i=1, x_i)}$$

$$P(z_i=1 | X_i=1) \rightarrow \frac{\sum_t P(z_i=1, x_i=1 | X=x^t, Y=y^t)}{T_i}$$

$$p_i \rightarrow \frac{\sum_t P(z_i=1, x_i=1 | X=x^t, Y=y^t)}{T_i}$$

$$(d) L = \frac{1}{T} \sum_t^T \log P(Y = \vec{y}_t | X = \vec{x}_t)$$

$$= \frac{1}{T} \sum_t^T (1-y_t) \log P(y=0 | \vec{x}_t) + y_t \log P(y=1 | \vec{x}_t)$$

$$= \frac{1}{T} \sum_t^T (1-y_t) \log x_{it} (1-p_i) + y_t \log \left( 1 - \prod_{i=1}^h (1-p_i)^{x_{it}} \right)$$

```
In [267]: import numpy as np
import math

In [268]: files =['X.txt', 'Y.txt']
def parse_data(files):
    res = []
    for file in files:
        data = []
        f = open(file, 'r')
        for line in f:
            data.append([int(num) for num in line.strip('\n').split()])
        f.close()
    res.append(data)
    return res

X, Y = parse_data(files)

X = np.array(X)
Y = np.array([y[0] for y in Y])
p = np.ones((23,1)) * 0.05

In [269]: def log_likelihood(p,X,Y):
    return 1 / X.shape[0] * np.sum(Y[:,np.newaxis] * np.log(1-np.exp(X.dot(np.log(1-p)))) + (1-Y[:,np.newaxis]) * X.dot(np.log(1-p)))

In [270]: def M_update(p,X,Y):
    before = (X.T * p).T * Y[:,np.newaxis] / (1 - np.exp(X.dot(np.log(1-p))))
    new = ((np.sum(before, axis = 0) / np.sum(X, axis = 0))[:,np.newaxis]
    return new

In [271]: def predict(p,X,Y):
    y_prob = 1 - np.exp(X.dot(np.log(1-p)))
    y_pred = np.where(y_prob >= 0.5, 1, 0)
    return np.sum(np.abs(y_pred - Y[:,np.newaxis]))

In [272]: likelihood_list = [log_likelihood(p,X,Y)]
error_list = [predict(p,X,Y)]
print(f"i=0 mistakies={error_list[0]} likelihood={likelihood_list[0]}")
max_i = 256
for i in range(1,max_i+1):
    p = M_update(p,X,Y)
    error = predict(p,X,Y)
    error_list.append(error)
    likelihood = log_likelihood(p,X,Y)
    likelihood_list.append(likelihood)
    if math.log(i, 2).is_integer():
        print(f"i={i} mistakies={error} likelihood={likelihood}")

i=0 mistakies=175 likelihood=-0.958085408215791
i=1 mistakies=56 likelihood=-0.49591639407753624
i=2 mistakies=43 likelihood=-0.4082208170583913
i=4 mistakies=42 likelihood=-0.3646149825001877
i=8 mistakies=44 likelihood=-0.3475006162087825
i=16 mistakies=40 likelihood=-0.33461704895854844
i=32 mistakies=37 likelihood=-0.3225814031674977
i=64 mistakies=37 likelihood=-0.31482669836285615
i=128 mistakies=36 likelihood=-0.3111558472151894
i=256 mistakies=36 likelihood=-0.31016135347407603
```

6.4

$$(a) f(x) = \log \cosh(x)$$

$$f'(x) = \frac{\sinh(x)}{\cosh(x)} = \tanh(x)$$

$$f'(x) = 0 \rightarrow x=0$$

$$f''(x) = \frac{d}{dx} \tanh(x) = \frac{1}{\cosh^2 x} = \operatorname{sech}^2 x$$

$$f''(x=0) > 0$$

$\therefore x=0$  is the minimum

$$(b) f''(x) = \operatorname{sech}^2 x$$

$$= \left( \frac{2}{e^x + e^{-x}} \right)^2$$

$$\text{By definition, } e^x = 1+x+\frac{x^2}{2!} + \frac{x^3}{3!} + \dots$$

$$[\text{Taylor series}], \text{ so } e^x \geq 1+x, e^{-x} \geq 1-x$$

$$f''(x) = \left( \frac{2}{e^x + e^{-x}} \right)^2$$

$$\leq \left( \frac{2}{1+x+1-x} \right)^2$$

$$\leq 1$$

C.

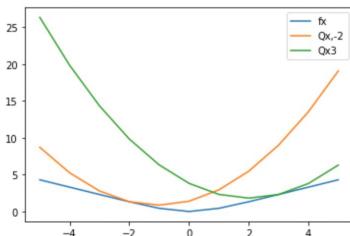
```
In [24]: import matplotlib.pyplot as plt
import numpy as np
import math

x1 = [i for i in range(-5, 6)]
y1 = [math.log(math.cosh(i)) for i in x1]

y2 = [math.log(math.cosh(-2))+math.tanh(-2)*(i+2)+ 0.5*(i+2)**2 for i in x1]
y3 = [math.log(math.cosh(3))+math.tanh(3)*(i-3)+ 0.5*(i-3)**2 for i in x1]

plt.plot(x1, y1, label="fx")
plt.plot(x1, y2, label="Qx,-2")
plt.plot(x1, y3, label="Qx3")
plt.legend()

# show the plot
plt.show()
```



$$(d) Q(x,y) = f(y) + f'(y)(x-y) + \frac{1}{2}(x-y)^2$$

$$\begin{aligned} Q(x,x) &= f(x) + f'(x)(x-x) + \frac{1}{2}(x-x)^2 \\ &= f(x) \end{aligned}$$

$$f(x) = f(y) + \int_y^x f'(u) du$$

$$= f(y) + \left( \int_y^x du \left( f'(y) + \int_y^u f''(v) dv \right) \right)$$

$$\text{from (b)} \quad f''_x \leq 1$$

$$\Rightarrow f(x) \leq f(y) + f'(y)(x-y) + \int_y^x du \int_y^u f''_x dv$$

$$= f(y) + f'(y)(x-y) + \frac{1}{2}(x-y)^2$$

$$= Q(x,y)$$

$$(e) \quad \frac{d}{dx} Q(x, x_n) = \frac{d}{dx} (f(x_n) + f'(x_n)(x - x_n) + \frac{1}{2} (x - x_n)^2)$$

$$= \frac{d}{dx} (f(x_n) + f'(x_n)x - \cancel{f'(x_n) \cdot x_n} + \frac{1}{2} x^2 - x x_n + \cancel{\frac{x^2}{2}})$$

$$= f'(x_n) + x - x_n$$

$$\frac{d}{dx} Q = 0 \Rightarrow f'(x_n) + x - x_n = 0$$

$$\Rightarrow x = x_n - f'(x_n)$$

$$\frac{d^2}{dx^2} Q(x, x_n) = \frac{d}{dx} (f'(x_n) + x - x_n)$$

$$= 1$$

$$\text{When } x = x_n - f'(x_n), \frac{d^2}{dx^2} Q > 0$$

$\therefore \operatorname{argmin}_x Q(x, x_n) \text{ when } x = x_n - f'(x_n)$

$$x_{n+1} = x_n - f'(x_n) = x_n - \tanh(x_n)$$