

6.1

Submitted

6.2

$$\alpha) P(a, b | c, d) = \frac{P(a, b, c, d)}{P(c, d)} \quad [P.R.]$$

$$= \frac{P(a) P(b | a) P(c | a, b) P(d | a, b, c)}{P(c, d)} \quad [P.R.]$$

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

rule-1
& marg

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

similar to numerator

b)

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

marg

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

marg

$$3) \quad \mathcal{L} = \sum_t^T \log P(C=c_t, D=d_t)$$

$$= \sum_t^T \log \sum_{\alpha, \beta} P(A=\alpha, B=\beta, C=c_t, D=d_t)$$

Marg

$$= \sum_t^T \log \sum_{\alpha, \beta} P(A=\alpha) P(B=\beta | A=\alpha) P(C=c_t | A=\alpha, B=\beta) \\ \cdot P(D=d_t | A=\alpha, B=\beta, C=c_t)$$

P.R.

$$= \sum_t^T \log \sum_{\alpha, \beta} P(A) P(B | A) P(C | A, B) P(D | B, C)$$

rule · 1

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

root node

$$② \quad P(B=b | A=\alpha) \leftarrow \frac{\sum_t^T P(\alpha, b | c_t, d_t)}{\sum_t^T P(\alpha | c_t, d_t)}$$

$$③ \quad P(C=c | A=\alpha, B=b) \leftarrow \frac{\sum_t^T P(\alpha, b, c | c_t, d_t)}{\sum_t^T P(\alpha, b | c_t, d_t)}$$

$$\leftarrow \frac{\sum_t^T P(\alpha, b | c_t, d_t) I(c, c_t)}{\sum_t^T P(\alpha, b | c_t, d_t)}$$

③

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

$$\leftarrow \frac{\sum_t I(c, c_t) I(d, d_t) P(b | c_t, d_t)}{\sum_t I(c, c_t) P(b | c_t, d_t)}$$

6.3

a)

$$\sum_{Z \in \{0,1\}^n} P(Y=1 | Z, X)$$

$$\Rightarrow \sum_{Z \in \{0,1\}^n} P(Y=1 | Z, X) P(Z | X) \quad [P.R.]$$

$$\Rightarrow \sum_{Z \in \{0,1\}^n} P(Y=1 | Z) P(Z | X) \quad [\text{Rule 1}]$$

$$\Rightarrow \sum_{Z \in \{0,1\}^n} (1 - I(Z, \{0\}^n)) P(Z | X)$$

$$\because P(Y=1 | Z) = \begin{cases} 1 & \text{if any } z_i \text{ is 1} \\ 0 & \text{if all } z_i \text{ are 0} \end{cases}$$

$$\Rightarrow P(Y=0 | Z) = \begin{cases} 0 & \text{if any } z_i \text{ is 1} \\ 1 & \text{if all } z_i \text{ are 0} \end{cases}$$

$$\Rightarrow \text{when } Z = \{0\}^n \Rightarrow P(Y=0 | Z = \{0\}^n) = 1 \\ \Rightarrow P(Y=0 | Z \neq \{0\}^n) = 0$$

$$\Rightarrow 1 - P(Y=0 | Z = \{0\}^n) = P(Y=1 | Z = \{0\}^n) = 0 \\ 1 - P(Y=0 | Z \neq \{0\}^n) = P(Y=1 | Z \neq \{0\}^n) = 1$$

$$\Rightarrow 1 - I(Z, \{0\}^n) = P(Y=1 | Z)$$

$$\Rightarrow \sum_{Z \in \{0,1\}^n} P(Z|X) - \sum_{Z \in \{0,1\}} P(Z|X) I(Z, \{0\}^n)$$

$$\Rightarrow 1 - \sum_{Z \in \{0,1\}} P(Z|X) I(Z, \{0\}^n) \quad [\text{moore}]$$

$$\Rightarrow 1 - P(Z = \{0\}^n | X) \quad \left[\begin{array}{l} \text{Indicator } f^n \text{ is} \\ \text{zero for rest} \end{array} \right]$$

$$\Rightarrow 1 - \prod_{i=1}^n P(Z_i = 0 | X) \quad [\text{rule-3}]$$

$$\Rightarrow 1 - \prod_{i=1}^n P(Z_i = 0 | X_i) \quad [\text{rule-3}]$$

$$\Rightarrow 1 - \prod_{i=1}^n (1 - P_i)^{x_i} (1)^{1-x_i}$$

$$\Rightarrow \boxed{1 - \prod_{i=1}^n (1 - P_i)^{x_i}}$$

$$b) P(Z_i=1, X_i=1 | X=x, Y=y)$$

$$\Rightarrow I(X_i, 1) P(Z_i=1 | X, Y)$$

X_i is observed node

$$\Rightarrow n_i \left[\frac{P(Z_i=1 | X) P(Y | X, Z_i=1)}{P(Y | X)} \right] \quad \text{Bayes}$$

$$\Rightarrow n_i \left[\frac{P(Z_i=1 | X_i) P(Y | Z_i=1)}{P(Y | X)} \right] \quad \begin{matrix} \text{rule-1} \\ \& \\ \text{rule-3} \end{matrix}$$

$\therefore Y$ is the observed node

$$\Rightarrow n_i \left[\frac{P_i I(Y_i, 1)}{P(Y | X)} \right]$$

$$\Rightarrow \frac{n_i P_i Y_i}{P(Y | X)}$$

$$\Rightarrow \frac{n_i P_i Y_i}{P(Y=1 | X)} \quad \because \text{if } Y_i = 0 \Rightarrow \text{numerator} = 0$$

$$\Rightarrow \boxed{\frac{x_i P_i y_i}{1 - \prod_j (1 - P_j)^{x_{ij}}}} \quad \text{from (a)}$$

c) From the M-step of EM algo:-

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

$$P(z_i | x_i) \leftarrow \frac{\sum_t^T P(z_i, x_i | X, Y)}{\sum_t^T I(x_i, x_i^{(+)})}$$

$$P(z_i | x_i) \leftarrow \frac{\sum_t^T P(z_i, x_i | X, Y)}{T_i}$$

\because only non-zero
 x_i will be added]

$$p_i = P(z_i = 1 | x_i = 1)$$

$$\Rightarrow P(z_i = 1 | x_i = 1) = p_i \leftarrow \frac{1}{T_i} \sum_t^T P(z_i = 1, x_i = 1 | X=x_i^{(+)}, Y=y)$$

d)

$$\mathcal{L} = \frac{1}{T} \sum_{t=1}^T \log P(Y=y^{(+)} | X=x^{(+)})$$

$$= \frac{1}{T} \sum_{t=1}^T \log \left[\left(1 - \prod_{i=1}^n (1 - p_i)^{x_i^{(+)}} \right) \left(\prod_{i=1}^n p_i^{x_i^{(+)}} \right)^{y^{(+)}} \right]$$

Q 6.3) d

In [1]:

```
import numpy as np
```

In [2]:

```
xfile = "noisyOrX.txt"
yfile = "noisyOrY.txt"
```

In [3]:

```
def parseFile(filename):
    data = []
    with open(filename) as f:
        for line in f:
            data.append([int(item) for item in line.strip('\n').strip().split()])
    return np.asarray(data)
```

In [48]:

```
x = parseFile(xfile)
y = parseFile(yfile)
```

In [49]:

```
p = np.ones(23) * 0.05
```

In [22]:

```
def P_Y_X(x, p):
    return 1 - np.prod(np.power(1 - p, x), axis=1)
```

In [7]:

```
def E_step(x, y, p):
    num = (x * p) * y
    den = P_Y_X(x, p)
    return num / np.expand_dims(den, axis=1)
```

In [24]:

```
def M_step(post, x):
    p = np.sum(post, axis=0)
    den = np.sum(x, axis=0)
    return np.squeeze(np.expand_dims(p, axis=1) / np.expand_dims(den, axis=1))
```

In [25]:

```
def log_likelihood(x, y, p):
    temp = np.zeros(y.shape)
    for i in range(y.shape[0]):
        temp[i] = P_Y_X(np.expand_dims(x[i], axis=0), p)
        if y[i] == 0:
            temp[i] = 1 - temp[i]
        temp[i] = np.log(temp[i])
    return np.sum(temp, axis=0)/y.shape[0]
```

In [41]:

```
def countMistakes(y, y_pred):
    count = 0
    for i in range(y.shape[0]):
        if y[i] != y_pred[i]:
            count += 1;
    return count;
```

In [46]:

```
peek_iterations = [0, 1, 2, 4, 8, 16, 32, 64, 128, 256]
```

In [50]:

```
for i in range(257):
    y_pred = P_Y_X(x, p)
    y_pred[y_pred >= 0.5] = 1
    y_pred[y_pred < 0.5] = 0
    if i in peek_iterations:
        print(i, countMistakes(y, y_pred), log_likelihood(x, y, p))
p = M_step(E_step(x, y, p), x)
```

```
0 175 [-0.95808541]
1 56 [-0.49591639]
2 43 [-0.40822082]
4 42 [-0.36461498]
8 44 [-0.34750062]
16 40 [-0.33461705]
32 37 [-0.3225814]
64 37 [-0.3148267]
128 36 [-0.31115585]
256 36 [-0.31016135]
```

6.4

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

$$\cosh(n) = \frac{e^n + e^{-n}}{2}$$

$$\Rightarrow f'(n) = \frac{1}{\cosh(n)} \times \frac{d}{dn} \left(\frac{e^n + e^{-n}}{2} \right)$$

$$= \frac{\sinh(n)}{\cosh(n)} = \tanh(n)$$

\Rightarrow To find the min. :-

$$f'(n) = 0 \Rightarrow \tanh(n) = 0$$

$$\Rightarrow \frac{e^n - e^{-n}}{e^n + e^{-n}} = 0$$

$$\Rightarrow e^{2n} = 1$$

$$\Rightarrow n = 0$$

$$f''(n) = \frac{(e^n + e^{-n})(e^n + e^{-n}) - (e^n - e^{-n})(e^n - e^{-n})}{(e^n + e^{-n})^2}$$

$$= \frac{4}{(e^n + e^{-n})^2}$$

$$= \frac{1}{\cosh^2(n)} = \operatorname{sech}^2(n)$$

$$f''(0) > 0$$

$\Rightarrow f'(0)$ gives minima at $x=0$

b) $f'(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} = \tanh(x)$

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

$$f'''(x) = 4 \left(\frac{-2(e^x + e^{-x})(e^x - e^{-x})}{(e^x + e^{-x})^3} \right)$$

$$= -\frac{8(e^x - e^{-x})}{(e^x + e^{-x})^3}$$

To find maxima of $f''(x)$

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

$$-\frac{8(e^x - e^{-x})}{(e^x + e^{-x})^3} = 0$$

$$\Rightarrow x = 0$$

$$f'''(x) = \frac{d}{dx} \left(\frac{-2 \sinh(x)}{\cosh^3(x)} \right)$$

$$= -2 \left(\frac{\cosh^3(x) \cosh(x) - \sinh(x) \times 3 \cosh^2(x) \sinh(x)}{\cosh^6(x)} \right)$$

$$= -2 \left(\frac{\cosh^2(x)(\cosh^2(x) - 3 \sinh^2(x))}{\cosh^6(x)} \right)$$

$$= -2 \left(\frac{\cosh^2(x) - 3 \sinh^2(x)}{\cosh^6(x)} \right)$$

$$= -2 \left(\frac{1 - 2 \sinh^2(x)}{\cosh^6(x)} \right)$$

$$f'''(x) = -2 \left(\frac{2.5 - \cosh^2(x)}{\cosh^2(x)} \right)$$

$$2.5 - \cosh^2(0) > 0$$

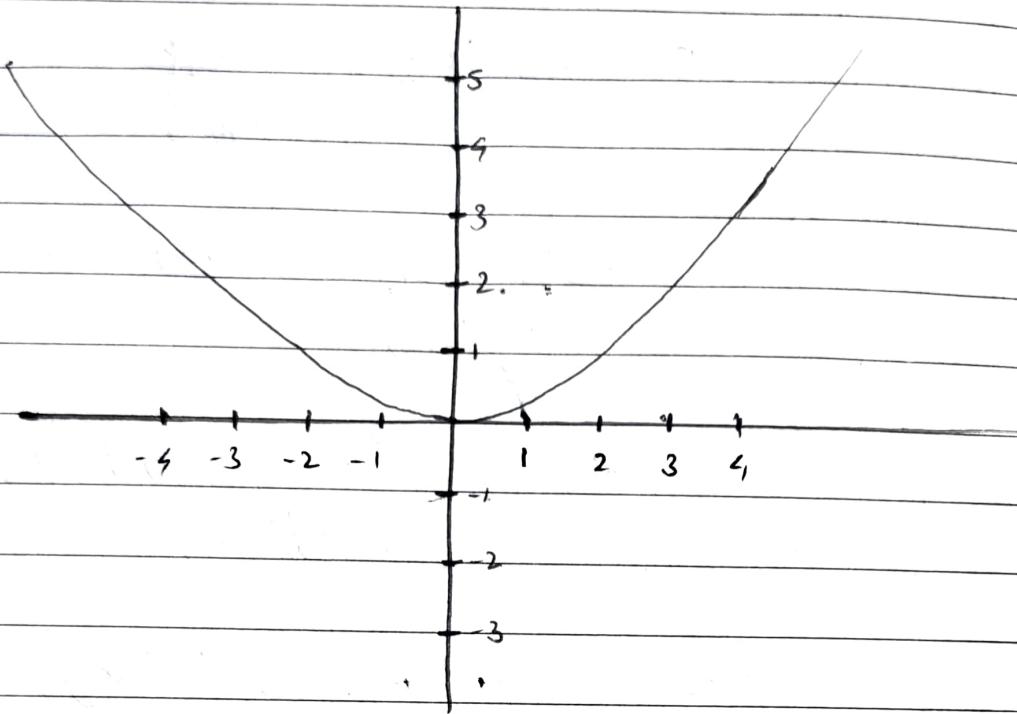
$$\Rightarrow f'''(0) < 0$$

$\Rightarrow f''(0)$ is the local maxima of $f''(x)$

$$\Rightarrow f''(x) \leq f''(0)$$

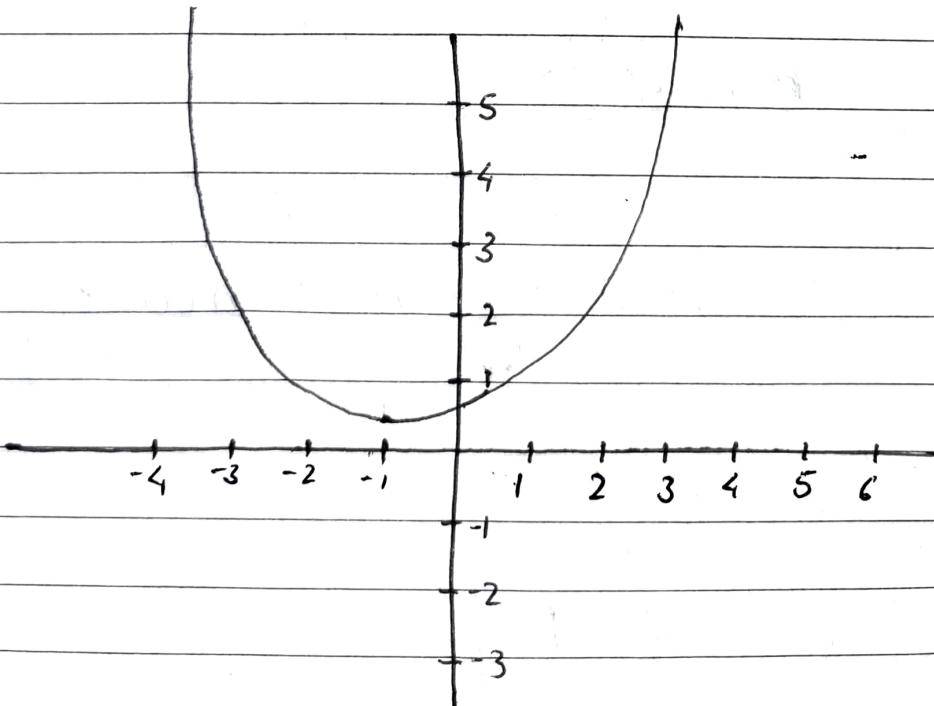
$$\Rightarrow |f''(x)| \leq 1$$

c) $f(n)$

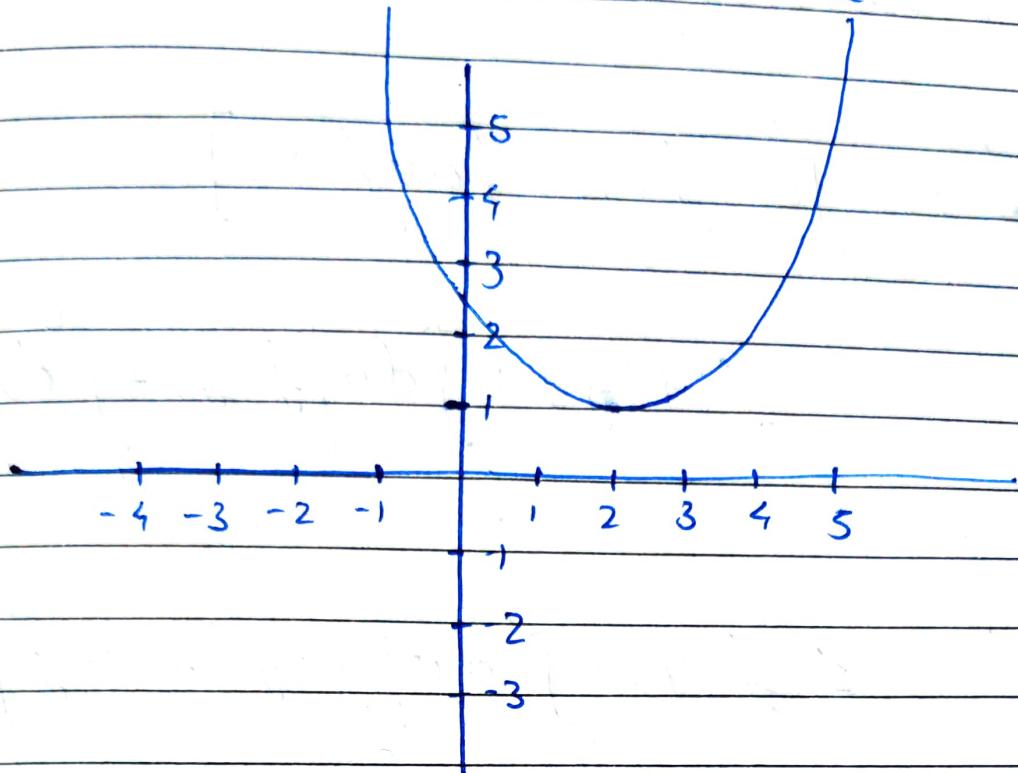


$$g(x, y) = f(y) + f'(y)(x-y) + \frac{1}{2}(x-y)^2$$

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



$$f(x, 3) = f(3) + f'(3)(x-3) + \frac{f''(3)}{2}(x-3)^2$$



a)

$$\text{i) } g(x, x) = f(x) + f'(x)(x-x) + \frac{f''(x)}{2}(x-x)^2$$

$$\boxed{g(x, x) = f(x)}$$

$$\text{ii) } f(x) = f(y) + \int_y^x dv \left[f'(y) + \int_y^v du f''(v) \right] \dots \text{j)$$

$$\therefore f''(x) \leq 1$$

$$\Rightarrow \int_y^v du f''(v) \leq \int_y^v du \quad \dots \text{j)}$$

\Rightarrow Substituting (ii) in (i)

$$f(y) + \int_y^x [f'(y) + \int_y^v f''(v) dv]$$

$$\leq f(y) + \int_y^n [f'(y) + \int_y^v dv]$$

$$\Rightarrow f(n) \leq f(y) + \int_y^n [f'(y) + v-y] dv$$

$$\Rightarrow f(n) \leq f(y) + \int_y^n du [f'(y) + \int_y^u (u-y) du]$$

$$\Rightarrow f(n) \leq f(y) + f'(y)(n-y) + \frac{(n-y)^2}{2}$$

$$\Rightarrow \boxed{f(n) \leq Q(n, y)}$$

Thus, $Q(n, y)$ is an auxiliary function

for $f(n)$

$$e) x_{n+1} = \underset{x}{\operatorname{argmin}} \Phi(x, x_n)$$

$$= \underset{x}{\operatorname{argmin}} f(x_n) + f'(x_n)(x - x_n) + \frac{(x - x_n)^2}{2}$$

... (i)

$$\text{let } g(x) = f(x_n) + f'(x_n)(x - x_n) + \frac{(x - x_n)^2}{2}$$

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

To find minima -

$$g'(x) = 0$$

$$\Rightarrow f'(x_n) + (x - x_n) = 0$$

$$\Rightarrow \boxed{x = x_n - f'(x_n)}$$

$$\Rightarrow \boxed{x = x_n - \tanh(x_n)}$$

$$\Rightarrow \boxed{x_{n+1} = x_n - \tanh(x_n)}$$

substituting
in (i)

In []:

```
import numpy as np
import matplotlib.pyplot as plt
import math
```

In [53]:

```
def fx(x):
    return np.log(np.cosh(x))
    """try:

        except OverflowError as oe:
            return float(math.inf)"""


```

In [54]:

```
def d_fx(x):
    return np.tanh(x)
    """try:

        except OverflowError as oe:
            return float(math.inf)"""


```

In [55]:

```
def d_d_fx(x):
    return 4/pow(2 * np.cosh(x), 2)
    """try:

        except OverflowError as oe:
            return float(math.inf)"""


```

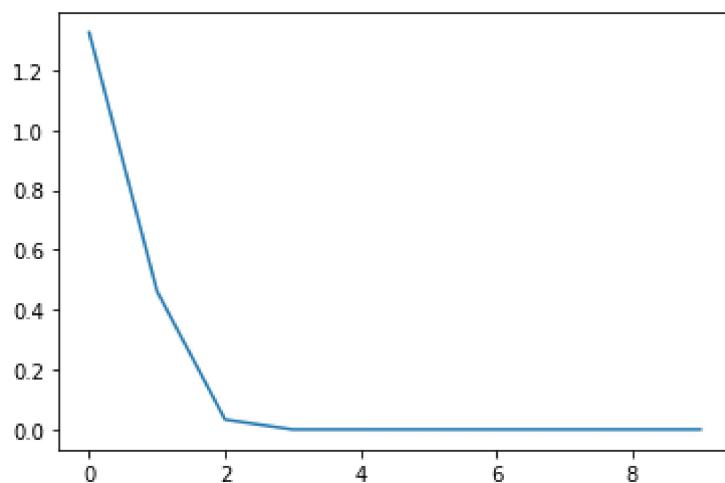
In [56]:

```
def update1(x_0, n):
    f_vals = []
    x = x_0
    for i in range(n):
        f_vals.append(fx(x))
        x = x - d_fx(x)
    return f_vals
```

6.4) f

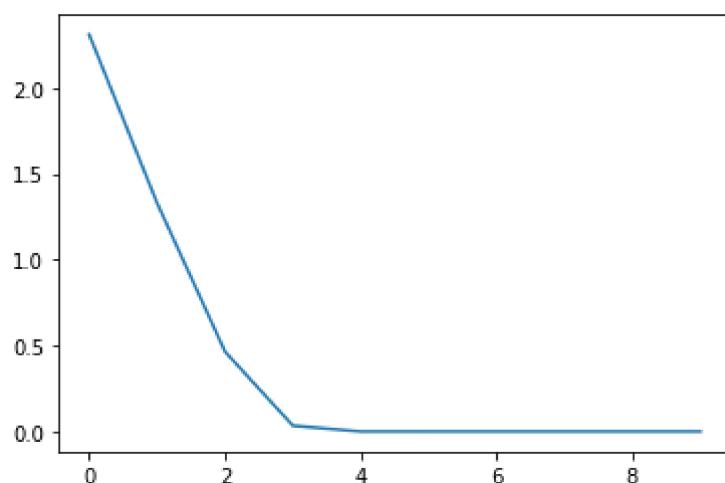
In [57]:

```
plt.plot(update1(-2, 10))
plt.show()
```



In [58]:

```
plt.plot(update1(3, 10))
plt.show()
```



6.4) g

$$(g) \quad x_{n+1} = x_n - \frac{f'(x_n)}{f''(x_n)}$$

$$\Rightarrow x_{n+1} = x_n - \frac{\tanh(n)}{\operatorname{sech}^2(n)}$$

$$x_{n+1} = x_n - \sinh(n) \cosh(n)$$

$$\Rightarrow x_1 = x_0 - \sinh(n) \cosh(n)$$

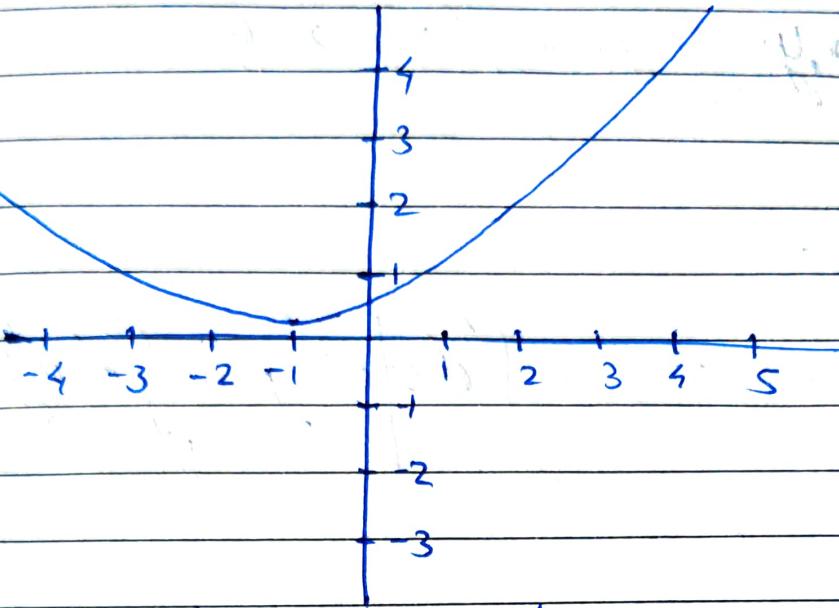
\Rightarrow for convergence

$$|x_1| < |x_0|$$

$$\Rightarrow |x_0 - \sinh(n) \cosh(n)| < |x_0|$$

$$\Rightarrow -1.08866 < n < 1.08866$$

$$h) g(x) = \frac{1}{10} \sum_{k=1}^{10} \log \cosh\left(x + \frac{2}{\sqrt{k}}\right)$$



No, can't find minima through analytical methods.
Thus, will use numerical methods.

$$i) R(x, y) = g(y) + g'(y)(x-y) + \frac{1}{2}(x-y)^2$$

$$1) R(x, x) = g(x) + g'(x)(x-x) + \frac{1}{2}(x-x)^2$$

$$R(x, x) = g(x)$$

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

$$\therefore f''(x) \leq 1 \Rightarrow f''\left(x + \frac{2}{\sqrt{k}}\right) \leq 1$$

$$\Rightarrow \frac{1}{10} \sum_{k=1}^{10} f''\left(\frac{x+2}{\sqrt{k}}\right) \leq 1$$

$$\Rightarrow g''(x) \leq 1$$

\Rightarrow similar to part(c)

$$R(x,y) = g(y) + g'(y)(x-y) + \frac{1}{2}(x-y)^2 \quad (i)$$

Also,

$$g(x) = g(y) + \int_y^x du \left[g'(y) + \int_y^u dv g''(v) \right] \quad (ii)$$

$$\therefore g''(x) \leq 1$$

$$\Rightarrow \int_y^u dv g''(v) \leq \int_y^u dv$$

$$\Rightarrow g(x) \leq g(y) + \int_y^x du \left[g'(y) + \int_y^u dv \right]$$

$$\Rightarrow g(x) \leq g(y) + \int_y^x du (g'(y)) + \int_y^x du (u-y)$$

$$\Rightarrow g(x) \leq g(y) + g'(y)(x-y) + \frac{(x-y)^2}{2}$$

$$\Rightarrow g(x) \leq R(x,y)$$

$\Rightarrow R(x, y)$ is auxiliary function to
 $g(x)$

(i) $x_{n+1} = \underset{x}{\operatorname{argmin}} R(x, x_n)$

$$= \underset{x}{\operatorname{argmin}} g(x_n) + g'(x_n)(x - x_n) + \frac{(x - x_n)^2}{2}$$

let,

$$l(x) = g(x_n) + g'(x_n)(x - x_n) + \frac{(x - x_n)^2}{2}$$

$$l'(x) = g'(x_n) + x - x_n$$

To find minima

$$l'(x) = 0$$

$$\Rightarrow 0 = g'(x_n) + x - x_n$$

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

$$\Rightarrow \boxed{x_{n+1} = x_n - g'(x)}$$

In [64]:

```
def update2(x_0, n):
    f_vals = []
    x = x_0
    for i in range(n):
        print(x, fx(x))
        f_vals.append(fx(x))
        x = x - (d_fx(x)/d_d_fx(x))
    return f_vals
```

In [65]:

```
update2(-2, 10)
```

```
-2 1.3250027473578645
11.644958598563875 10.951811418080721
-3255536207.1877036 inf
inf inf
nan nan
C:\Users\vaibh\AppData\Local\Temp\ipykernel_6836/3182926395.py:2: RuntimeWarning: overflow encountered in cosh
    return np.log(np.cosh(x))
C:\Users\vaibh\AppData\Local\Temp\ipykernel_6836/2058034218.py:7: RuntimeWarning: divide by zero encountered in double_scalars
    x = x - (d_fx(x)/d_d_fx(x))
C:\Users\vaibh\AppData\Local\Temp\ipykernel_6836/2058034218.py:7: RuntimeWarning: invalid value encountered in double_scalars
    x = x - (d_fx(x)/d_d_fx(x))
```

Out[65]:

```
[1.3250027473578645,
 10.951811418080721,
 inf,
 inf,
 nan,
 nan,
 nan,
 nan,
 nan,
 nan,
 nan]
```

In [66]:

```
update2(3, 10)
```

```
3 2.309328504577785
-97.85657868513961 97.16343150457966
2.4836150932578143e+84 inf
-inf inf
nan nan
C:\Users\vaibh\AppData\Local\Temp\ipykernel_6836/3182926395.py:2: RuntimeWarning: overflow encountered in cosh
    return np.log(np.cosh(x))
C:\Users\vaibh\AppData\Local\Temp\ipykernel_6836/2058034218.py:7: RuntimeWarning: divide by zero encountered in double_scalars
    x = x - (d_fx(x)/d_d_fx(x))
C:\Users\vaibh\AppData\Local\Temp\ipykernel_6836/2058034218.py:7: RuntimeWarning: invalid value encountered in double_scalars
    x = x - (d_fx(x)/d_d_fx(x))
```

Out[66]:

```
[2.309328504577785, 97.16343150457966, inf, inf, nan, nan, nan, nan, nan, na
n]
```

6.4) k

In [67]:

```
def gx(x):
    result = 0
    for k in range(1, 11):
        result += fx(x + (2 / np.sqrt(k)))

    return result/10
```

In [68]:

```
def d_gx(x):
    result = 0
    for k in range(1, 11):
        result += np.tanh(x + (2 / np.sqrt(k)))

    return result/10
```

In [69]:

```
def update3(x_0, n):
    g_vals = []
    x = x_0
    for i in range(n):
        g_vals.append(gx(x))
        x = x - d_gx(x)
    return g_vals
```

In [73]:

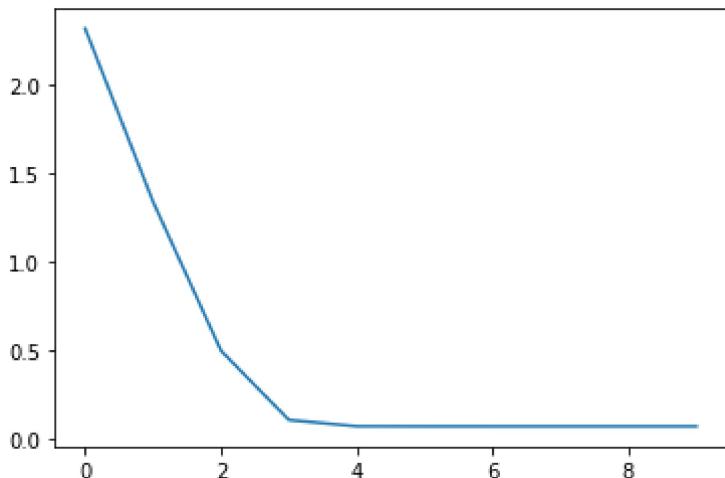
```
update3(2, 10)
```

Out[73]:

```
[2.3141309089320035,
 1.339399484892293,
 0.5005916001821507,
 0.11050257838877894,
 0.07474292148721459,
 0.07421617940304373,
 0.07420940457263744,
 0.07420931571860692,
 0.0742093145501224,
 0.07420931453475131]
```

In [74]:

```
plt.plot(update3(2, 10))
plt.show()
```



In []: