

HW7:

(Q5) [Q] for problem 1, we chose the model with $K=6$ based on the code which yielded an out-of-sample classification error of 0.084. for problem 3, we chose the model with $K=6$ based on the code which yielded an out-of-sample classification error of 0.192. Hence the correct answer is [D] bcz it is the closest.

(87) with leave-one-out cross-validation, we leave

leaving out $\{-1, 0\}$ and find the using $\{1, 0\} \notin \{P, 1\}$

$$\mathcal{G}_0^{(0)}(\beta) = \frac{1}{\beta}$$

$$g_1^{(n)}(x) = \frac{x^n}{n+1}$$

bracing out: $(0, 1)$ and find $f(1)$ using $(1, 0) \in (-1, 0)$

$$g_0(r) = 0$$

$$g_1^{(b)}(x) \leq 0$$

swings out}: $(1, 0)$, and finds it using $(-1, 0) \in \{1, -1\}$

$$\mathcal{G}_0^{(e)}(r) \approx \frac{1}{r}$$

$$\bar{g}_1^{(n)}(x) = \frac{x-1}{n-1}$$

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$$((1-x)^2)((x-1)^2)(1-x^2)+1=0$$

For a more detailed error analysis

$$\left(\frac{-1}{p-1} - 0\right)^2 + (0-1)^2 + \left(\frac{1+1}{p+1} - 0\right)^2 = \frac{4}{(p-1)^2} + 1 + \frac{4}{(p+1)^2}$$

Now, see the two answers equal & solve.

$$\frac{3}{2} = \frac{4}{(P-1)^2} + \frac{4}{(P+1)^2} + 1 \Rightarrow \frac{1}{2} = \frac{4}{(P-1)^2} + \frac{4}{(P+1)^2}$$

$P = \pm \sqrt{9+4\sqrt{6}}$ hence the correct

Correct answer is [d]

(Q6) 57

```

import points
import numpy
import random
import math

def _in_data = points.get_sample_points()
def _out_data = points.get_test_points()

def _dot(x1, x2, c):
    return x1[0]*x2[0] + x1[1]*x2[1] + c*x1[2]*x2[2]

def _softmax(x):
    sum = 0.0
    for i in range(len(x)):
        sum += math.exp(x[i])
    if sum == 0.0:
        return [0.0] * len(x)
    else:
        return [math.exp(x[i]) / sum for i in range(len(x))]

def _get_weight(k):
    return np.array([random.uniform(-0.1, 0.1) for i in range(3)])
    #return np.array([0.0] * 3)

def _out(x):
    w = _get_weight(_out_data[k])
    return _softmax(_dot(x, w))

def _out_test(k):
    w = _get_weight(_out_data[k])
    return _softmax(_dot(_in_data, w))

def _validation_bias():
    e1 = random.uniform(-0.1, 0.1)
    e2 = random.uniform(-0.1, 0.1)
    e3 = random.uniform(-0.1, 0.1)
    e = min(e1, e2, e3)
    return (e1, e2, e3)

def _init_bias(b):
    e1_av = 0
    e2_av = 0
    e3_av = 0
    for i in range(N):
        curr = _validation_bias()
        e1_av += curr[0]
        e2_av += curr[1]
        e3_av += curr[2]
    return (e1_av / N, e2_av / N, e3_av / N)

def _gradient_bias():
    curr = _validation_bias()
    e1_av = curr[0]
    e2_av = curr[1]
    e3_av = curr[2]
    return (e1_av / N, e2_av / N, e3_av / N)

def _train():
    curr = _validation_bias()
    e1_av = curr[0]
    e2_av = curr[1]
    e3_av = curr[2]
    for i in range(1000000):
        curr = _validation_bias()
        e1_av += curr[0]
        e2_av += curr[1]
        e3_av += curr[2]
        if i % 10000 == 0:
            print("Iteration: " + str(i) + ", Loss: " + str(_loss(e1_av / 100000, e2_av / 100000, e3_av / 100000)))
    print("Training finished! Loss: " + str(_loss(e1_av / 100000, e2_av / 100000, e3_av / 100000)))

```

```

(4) [d]
1 import points
2 import numpy
3
4 in_data = points.get_sample_points()
5 out_data = points.get_test_points()
6
7 def transform(x1, x2, k):
8     transformation = [1, x1, x2, x1 + x2, x2 + x1, abs(x1 - x2), abs(x1 + x2)]
9     return transformation[::k]
10
11 def get_weight(k):
12     data = []
13     scores = []
14
15     for point in in_data:
16         data.append(transform(point[0], point[1], k))
17         scores.append(point[2])
18
19     data = np.array(data)
20     scores = np.array(scores)
21
22     w = np.dot(np.transpose(data), scores)
23     return w, scores, data
24
25 def f_out():
26
27 def f_out_test():
28     data = []
29     scores = []
30
31     for point in in_data:
32         data.append(transform(point[0], point[1], k))
33         scores.append(point[2])
34
35     data = np.array(data)
36     scores = np.array(scores)
37
38     sum_error = 0
39
40     for d, s in zip(data, scores):
41         if (numpy.sign(w[0]*d[0], w)) != s:
42             sum_error += 1
43
44     return sum_error/len(scores)
45
46
47 print("%3s", "Out_1")
48 print("%3s", "Out_2")
49 print("%3s", "Out_3")
50 print("%3s", "Out_4")
51 print("%3s", "Out_5")
52 print("%3s", "Out_6")
53 print("%3s", "Out_7")

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPITER
 thierry@thierry-MacBook-Pro CS 156a % /usr/bin/python "/Users/thierry/Desktop/CatchScore/fall 22/CS_156a/mw.py"
 k = 4.388
 ↪ According to the code the smallest is
 when k=6 hence the
 correct answer is [d]

```

(3) [d]
1 import points
2 import numpy
3
4 in_data = points.get_sample_points()
5 out_data = points.get_test_points()
6
7 def transform(x1, x2, k):
8     transformation = [1, x1, x2, x1 + x2, x2 + x1, abs(x1 - x2), abs(x1 + x2)]
9     return transformation[::k]
10
11 def get_weight(k):
12     data = []
13     scores = []
14
15     for point in in_data:
16         data.append(transform(point[0], point[1], k))
17         scores.append(point[2])
18
19     data = np.array(data)
20     scores = np.array(scores)
21
22     w = np.dot(np.transpose(data), scores)
23     return w, scores, data
24
25 def f_out():
26
27 def f_out_test():
28     data = []
29     scores = []
30
31     for i in range(7):
32         data.append(transform(in_data[i][0], in_data[i][1], k))
33         scores.append(in_data[i][2])
34
35     data = np.array(data)
36     scores = np.array(scores)
37
38     sum_error = 0
39
40     w = get_weight(4)
41     for d, s in zip(data, scores):
42         if (numpy.sign(w[0]*d[0], w)) != s:
43             sum_error += 1
44
45     return sum_error/len(scores)
46
47 print("%3s", "Out_1")
48 print("%3s", "Out_2")
49 print("%3s", "Out_3")
50 print("%3s", "Out_4")
51 print("%3s", "Out_5")
52 print("%3s", "Out_6")
53 print("%3s", "Out_7")

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPITER
 thierry@thierry-MacBook-Pro CS 156a % /usr/bin/python "/Users/thierry/Desktop/CatchScore/fall 22/CS_156a/mw.py"
 k = 4.36
 ↪ According to the code the smallest is
 when k=6 hence the
 correct answer is [d]

```

(2) [e]
1 import points
2 import numpy
3
4 in_data = points.get_sample_points()
5 out_data = points.get_test_points()
6
7 def transform(x1, x2, k):
8     transformation = [1, x1, x2, x1 + x2, x2 + x1, abs(x1 - x2), abs(x1 + x2)]
9     return transformation[::k]
10
11 def get_weight(k):
12     data = []
13     scores = []
14
15     for point in in_data:
16         data.append(transform(point[0], point[1], k))
17         scores.append(point[2])
18
19     data = np.array(data)
20     scores = np.array(scores)
21
22     w = np.dot(np.transpose(data), scores)
23     return w, scores, data
24
25 def f_out():
26
27 def f_out_test():
28     data = []
29     scores = []
30
31     for point in in_data:
32         data.append(transform(point[0], point[1], k))
33         scores.append(point[2])
34
35     data = np.array(data)
36     scores = np.array(scores)
37
38     sum_error = 0
39
40     w = get_weight(4)
41     for d, s in zip(data, scores):
42         if (numpy.sign(w[0]*d[0], w)) != s:
43             sum_error += 1
44
45     return sum_error/len(scores)
46
47 print("%3s", "Out_1")
48 print("%3s", "Out_2")
49 print("%3s", "Out_3")
50 print("%3s", "Out_4")
51 print("%3s", "Out_5")
52 print("%3s", "Out_6")
53 print("%3s", "Out_7")

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPITER
 ValueError: shapes (4,) and (3,) don't align: inner dim must be same
 k = 3.000
 ↪ According to the code the smallest is
 when k=7 hence the
 correct answer is [e]

(10) [E]

```

1 import numpy
2 import random
3 from sklearn import linear_model
4 from sklearn.linear_model import Perceptron
5 def get_slope(x1, x2, y1, y2):
6     slope = (y2 - y1) / (x2 - x1)
7     y_intercept = y1 - (slope * x1)
8     return slope, y_intercept
9 def classifyIf(slope, b, x1, x2):
10    if (slope * x1 + b) > x2:
11        return -1
12    else:
13        return 1
14 def Euclidean_d(score_1):
15    for s1, s2 in zip(score_1, score_2):
16        if np.sign(s1) != np.sign(s2):
17            av += 1
18    return av/len(score_1)
19 def make_points(x1, b, slope):
20    if (b == None):
21        points = [(random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))]
22    else:
23        slope, b = get_slope(points[0], points[1], points[2])
24    data, score, b, slope = [], [], None
25    for i in range(100):
26        point = (i, random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))
27        data.append(point)
28        score.append(classifyIf(slope, b, point[0], point[1]))
29    plt.title("Support Vector Machine")
30    plt.xlabel("x1")
31    plt.ylabel("x2")
32    plt.show()
33    sv_better = 0
34    support_x = []
35    for i in range(av):
36        data, score, b, slope = make_points(b, None, None)
37        points = [(random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))]
38        slope, b = get_slope(points[0], points[1], points[2])
39        data, score, b, slope = [], [], None
40        for i in range(100):
41            point = (i, random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))
42            data.append(point)
43            score.append(classifyIf(slope, b, point[0], point[1]))
44        plt.title("Support Vector Machine")
45        plt.xlabel("x1")
46        plt.ylabel("x2")
47        plt.show()
48    return sv_better / runs, support_x
49
50 result = 0
51 print(result)

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPITER + - x
 Number of support vectors: 3
 SVM was better this many times: 0.0000000000000000
 SVM was better this many times: 0.0000000000000000

3 is the closest answer for N=100.

(9) [E]

```

1 import numpy
2 import random
3 from sklearn import linear_model
4 from sklearn.linear_model import Perceptron
5 def get_slope(x1, x2, y1, y2):
6     slope = (y2 - y1) / (x2 - x1)
7     y_intercept = y1 - (slope * x1)
8     return slope, y_intercept
9 def classifyIf(slope, b, x1, x2):
10    if (slope * x1 + b) > x2:
11        return -1
12    else:
13        return 1
14 def Euclidean_d(score_1):
15    av = 0
16    for s1, s2 in zip(score_1, score_2):
17        if np.sign(s1) != np.sign(s2):
18            av += 1
19    return av/len(score_1)
20 def make_points(x1, b, slope):
21    if (b == None):
22        points = [(random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))]
23    else:
24        slope, b = get_slope(points[0], points[1], points[2])
25    data, score, b, slope = [], [], None
26    for i in range(100):
27        point = (i, random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))
28        data.append(point)
29        score.append(classifyIf(slope, b, point[0], point[1]))
30    plt.title("Support Vector Machine")
31    plt.xlabel("x1")
32    plt.ylabel("x2")
33    plt.show()
34    sv_better = 0
35    support_x = []
36    for i in range(av):
37        data, score, b, slope = make_points(b, None, None)
38        points = [(random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))]
39        slope, b = get_slope(points[0], points[1], points[2])
40        data, score, b, slope = [], [], None
41        for i in range(100):
42            point = (i, random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))
43            data.append(point)
44            score.append(classifyIf(slope, b, point[0], point[1]))
45        plt.title("Support Vector Machine")
46        plt.xlabel("x1")
47        plt.ylabel("x2")
48        plt.show()
49    return sv_better / runs, support_x
50 result = 0
51 print("SVM was better this many times: ", result)

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPITER + - x
 SVM was better this many times: 0.792
 Hence by the above code, the correct answer is [d] as 65% is closest.

(8) [E]

```

1 import numpy
2 import random
3 from sklearn import linear_model
4 from sklearn.linear_model import Perceptron
5 def get_slope(x1, x2, y1, y2):
6     slope = (y2 - y1) / (x2 - x1)
7     y_intercept = y1 - (slope * x1)
8     return slope, y_intercept
9 def classifyIf(slope, b, x1, x2):
10    if ((x2 - x1) * slope + b) > x2:
11        return -1
12    else:
13        return 1
14 def Euclidean_d(score_1):
15    for s1, s2 in zip(score_1, score_2):
16        if np.sign(s1) != np.sign(s2):
17            av += 1
18    return av/len(score_1)
19 def make_points(x1, b, slope):
20    if (b == None):
21        points = [(random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))]
22    else:
23        slope, b = get_slope(points[0], points[1], points[2])
24    data, score, b, slope = [], [], None
25    for i in range(100):
26        point = (i, random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))
27        data.append(point)
28        score.append(classifyIf(slope, b, point[0], point[1]))
29    plt.title("Support Vector Machine")
30    plt.xlabel("x1")
31    plt.ylabel("x2")
32    plt.show()
33    sv_better = 0
34    support_x = []
35    for i in range(av):
36        data, score, b, slope = make_points(b, None, None)
37        points = [(random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))]
38        slope, b = get_slope(points[0], points[1], points[2])
39        data, score, b, slope = [], [], None
40        for i in range(100):
41            point = (i, random.uniform(-1, 1), random.uniform(-1, 1), random.uniform(-1, 1))
42            data.append(point)
43            score.append(classifyIf(slope, b, point[0], point[1]))
44        plt.title("Support Vector Machine")
45        plt.xlabel("x1")
46        plt.ylabel("x2")
47        plt.show()
48    return sv_better / runs, support_x
49
50 result = 0
51 print(result)

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPITER + - x
 SVM was better this many times: 0.636
 Hence by the above code, the correct answer is [c] as 63.6% is the closest.