

(7) [C]

The hypothesis set $H(a_0, c_1, b_0)$ where $c=0$ would create a set of polynomials of order $\min(a_0, b_0)$ or less (as H_0 is order n or less)

HW 6

because $w_q = c = 0$ when $q \geq b_0$, as seen above, then we use an upper bound of $b_0 - 1$. However Q could also be a potential upper bound for when $c \neq 0$,

$$\cdot H(10, 0, 3) = \left\{ h | h(x) = w^T \sum_{q=0}^{10} w_q L_q(x), w_q = 0 \text{ for } q \geq 3 \right\} = \left\{ h | h(x) = \sum_{q=0}^2 w_q L_q(x) \right\} = H_2$$

$$\cdot H(10, 0, 4) = \left\{ h | h(x) = w^T \sum_{q=0}^{10} w_q L_q(x), w_q = 0 \text{ for } q \geq 4 \right\} = \left\{ h | h(x) = \sum_{q=0}^3 w_q L_q(x) \right\} = H_3$$

$$\cdot H(10, 1, 3) = \left\{ h | h(x) = w^T \sum_{q=0}^{10} w_q L_q(x), w_q = 1 \text{ for } q \geq 3 \right\} = \left\{ h | h(x) = \sum_{q=0}^3 w_q L_q(x) + \sum_{q=4}^{10} L_q(x) \right\} = H_{10}$$

$$\cdot H(10, 1, 4) = \left\{ h | h(x) = w^T \sum_{q=0}^{10} w_q L_q(x), w_q = 1 \text{ for } q \geq 4 \right\} = \left\{ h | h(x) = \sum_{q=0}^4 w_q L_q(x) + \sum_{q=5}^{10} L_q(x) \right\} = H_{10}$$

Now, [a] $H(10, 0, 3) \cup H(10, 0, 4) = H_4$

a is false because :

$$H(10, 0, 3) = H_2 \notin H_4 \text{ as shown above hence their union is } H_3 \text{ not } H_4$$

[b] $H(10, 1, 3) \cup H(10, 1, 4) = H_3$

b is false because :

$$H(10, 1, 3) = H_{10} \notin H_3 \text{ as shown above hence their union is greater than order 3 hence it's not } H_3 \text{ as it is for order 3 polynomials and less.}$$

[c] $H(10, 0, 3) \cap H(10, 0, 4) = H_2$

c is correct because :

$$H(10, 0, 3) = H_2 \notin H_4 \text{ as shown above hence their intersection is } H_2 \cap H_4 = H_2$$

[d] $H(10, 1, 3) \cap H(10, 1, 4) = H_4$

d is false because :

$$H(10, 1, 3) = H_{10} \notin H_4 \text{ as shown above hence their intersection is greater than } H_4,$$

(8) [d]

$1 \leq j \leq 2$ inputs; $0 \leq i \leq 5$ inputs; $1 \leq l \leq 3$ outputs

* Forward: Compute all $x_l^{(t)}$ for $l \geq 0$:

$$x_0^{(t)} = \theta\left(\sum_{i=0}^5 w_{0i}^{(t)} x_i^{(t)}\right) \quad \text{as sum to make 10}$$

$$x_1^{(t)} = \theta\left(\sum_{i=0}^5 w_{1i}^{(t)} x_i^{(t)}\right)$$

$$x_2^{(t)} = \theta\left(w_{21}^{(t)} x_1^{(t)} + w_{22}^{(t)} x_2^{(t)} + w_{23}^{(t)} x_3^{(t)} + w_{24}^{(t)} x_4^{(t)} + w_{25}^{(t)} x_5^{(t)}\right)$$

$$x_3^{(t)} = \theta\left(w_{31}^{(t)} x_1^{(t)} + w_{32}^{(t)} x_2^{(t)} + w_{33}^{(t)} x_3^{(t)} + w_{34}^{(t)} x_4^{(t)} + w_{35}^{(t)} x_5^{(t)}\right)$$

$$x_4^{(t)} = \theta\left(w_{41}^{(t)} x_1^{(t)} + w_{42}^{(t)} x_2^{(t)} + w_{43}^{(t)} x_3^{(t)} + w_{44}^{(t)} x_4^{(t)} + w_{45}^{(t)} x_5^{(t)}\right)$$

$$\text{Now, Count all } w_{ij}^{(t)} x_i^{(t)}. \text{ We get } (3 \times 6) + 4 = 18 + 4 = 22 \text{ operations}$$

Backwards & compute all:

$$\delta_t^{(l)} = \left(1 - (x_t^{(l-1)})^2\right) \sum_{j=1}^m w_{lj}^{(t)} \delta_j^{(t)} \quad \text{from lecture slide 10.}$$

$$\text{for } l=1: \quad \delta_1^{(1)} = \left(1 - (x_1^{(0)})^2\right) \sum_{j=1}^m w_{j1}^{(1)} \delta_j^{(1)} \quad \text{since each layer has 6 units}$$

$$\delta_1^{(1)} = K \sum_{j=1}^3 w_{j1}^{(1)} \delta_j^{(1)} \quad \text{where } K \text{ is the number of operations for the layer.}$$

$$\text{Number of operations for the super layer is } 3.$$

Zero as there is no signal to 0th node.

for $l=2:$

$$\delta_2^{(2)} = K \left(w_{02}^{(2)} \delta_0^{(2)} + w_{12}^{(2)} \delta_1^{(2)} + w_{22}^{(2)} \delta_2^{(2)} + w_{32}^{(2)} \delta_3^{(2)} + w_{42}^{(2)} \delta_4^{(2)} + w_{52}^{(2)} \delta_5^{(2)} \right) \quad \text{we will have more of these operations for } i < 2, \text{ hence } 3 \times 3 = 9 \text{ total operations}$$

* Now for updating the weights:

$$w_{ij}^{(t)} = w_{ij}^{(t)} + \eta x_i^{(t)} \delta_j^{(t)} \quad \text{from lecture slide 20.}$$

we only care about this part.

$$\text{for } l=1: 0 \leq i \leq 5; 1 \leq j \leq 3$$

$$w_{01}^{(1)} = w_{01}^{(1)} + \eta x_0^{(0)} \delta_1^{(1)} \quad \text{we will have 9 total of } 3 \times 6 = 18 \text{ operations for } j=1 \text{ if } i=0, 1, 2, 3, 4, 5 \text{ as for each } j \text{ we must iterate through 1's.}$$

$$\text{for } l=2: 0 \leq i \leq 3; 1 \leq j \leq 1$$

$$w_{02}^{(2)} = w_{02}^{(2)} + \eta x_0^{(1)} \delta_2^{(2)} \quad \text{again we will have 9 total of } 4 \times 1 = 4 \text{ of these operations for } j=1 \text{ if } i=0, 1, 2, 3 \text{ as for each } j \text{ we must iterate through 1's.}$$

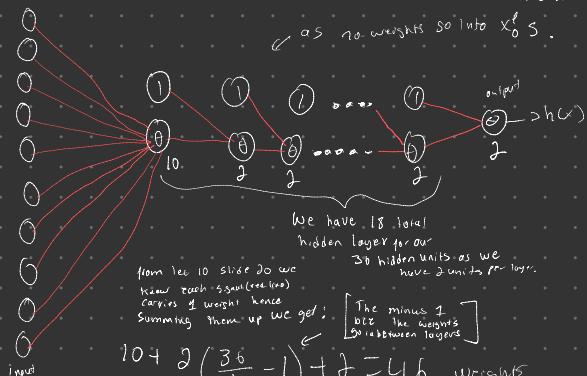
Hence we have a total number of operations equal to $4 \times 18 = 72$ for weight updates.

Hence our total number of all 3 operations is:

$22 + 3 + 32 = 47$ hence the correct answer is [d] as it is the closest.

(9) [a]

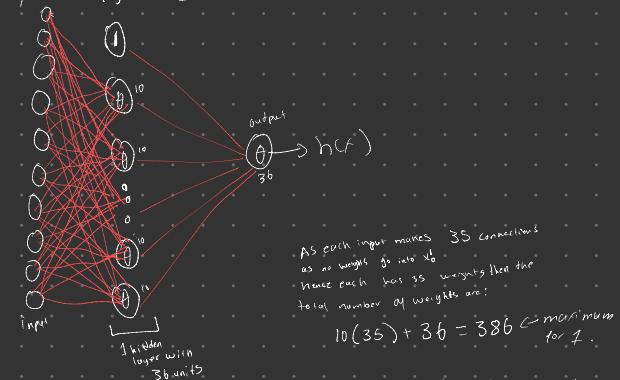
The minimum possible number of weights is achieved when we put 2 units in each hidden layers as seen below and as similarly drawn in lecture slide 13.



hence the correct answer is [a].

(10) [e]

We will achieve the maximum number of possible weights when we minimize our hidden layers. Let's look at when the number of hidden layers is 1 hence similar to the situation in q we will have 6



Now for the case we have 2 hidden layers with 18 units in each then as seen by the previous networks, all 10 inputs connect to 17 units in the first hidden layer as $x_0^{(t)}$ don't receive any weight. Then each of the 18 units in the first hidden layer will connect to 17 units in hidden layer 2. Then each of these 18 units will connect to our one output hence giving us:

$$(10 \times 17) + (18 \times 17) + 18 = 494 \text{ connections hence 494 weights. This also gives us a general formula for total number of weights for } x \text{ layers: } 10(x-1) + x(36-x-1) + (36-x)$$

$$x \text{ being the number of units for the first layer as the units are connected.}$$

$$w = 10 \times 17 + 36 \times 17 - x - 1 = 3x + 494 \text{ for critical points}$$

$$0 = 494 - 2x; x = 22; \frac{x}{22} = 22 \text{ is maximum}$$

so we have a max no when $x = 22$, hence 22 units in layer 1.

$$\text{for } x = 22 \rightarrow w(22) = 10(21) + 22(36-22-1) + (36-22) = 510 \text{ hence is the maximum weight.}$$

Hence the correct answer is [e].

(1) [b]

Deterministic noise is our bias as seen in lecture 11 slide 20. As we know from lecture 8 slide 15, our Bias increases with less complex hypothesis sets hence increasing deterministic noise for H' as it is a subset of H hence it is less complex.

↗

By this reasoning the correct answer is [b].

(2) [a]

```
def get_training_points():
    data = []
    for i in range(1000):
        x = np.random.uniform(-1, 1)
        y = np.sin(x) + np.random.normal(0, 0.05)
        data.append([x, y])
    return np.array(data)

def get_test_points():
    data = []
    for i in range(100):
        x = np.random.uniform(-1, 1)
        y = np.sin(x) + np.random.normal(0, 0.05)
        data.append([x, y])
    return np.array(data)
```

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[1]:
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```
[1]:
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```
def get_training_points():
    data = []
    for i in range(1000):
        x = np.random.uniform(-1, 1)
        y = np.sin(x) + np.random.normal(0, 0.05)
        data.append([x, y])
    return np.array(data)

def get_test_points():
    data = []
    for i in range(100):
        x = np.random.uniform(-1, 1)
        y = np.sin(x) + np.random.normal(0, 0.05)
        data.append([x, y])
    return np.array(data)
```

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[2]:
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```
[2]:
```

hence according to
inc code the
correct answe is
[a]

(3) [d]

```
def get_weight(data, identity):
    data = np.array(data)
    identity = np.array(identity)
    data_t = np.transpose(data)
    identity_t = np.transpose(identity)
    dot = np.dot(data_t, identity)
    inv = np.linalg.inv(np.dot(data_t, data))
    identity_dot = np.dot(inv, dot)
    return identity_dot

def get_weight_reg(data, identity, reg):
    data = np.array(data)
    identity = np.array(identity)
    data_t = np.transpose(data)
    identity_t = np.transpose(identity)
    dot = np.dot(data_t, identity)
    identity_dot = np.dot(data_t, identity)
    identity_dot = np.dot(identity_dot, data)
    identity_dot = identity_dot - reg * identity
    inv = np.linalg.inv(np.dot(data_t, data))
    identity_dot = np.dot(inv, identity_dot)
    return identity_dot

def sum_error(data, identity):
    data = np.array(data)
    identity = np.array(identity)
    error = 0
    for d in data:
        error += np.sum((d - identity)**2)
    return error
```

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[1]:
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```
[1]:
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hence according to the
code, the correct answer is [d]

(4) [c]

```
def get_weight(data, identity):
    data = np.array(data)
    identity = np.array(identity)
    data_t = np.transpose(data)
    identity_t = np.transpose(identity)
    dot = np.dot(data_t, identity)
    inv = np.linalg.inv(np.dot(data_t, data))
    identity_dot = np.dot(inv, dot)
    return identity_dot

def get_weight_reg(data, identity, reg):
    data = np.array(data)
    identity = np.array(identity)
    data_t = np.transpose(data)
    identity_t = np.transpose(identity)
    dot = np.dot(data_t, identity)
    identity_dot = np.dot(data_t, identity)
    identity_dot = np.dot(identity_dot, data)
    identity_dot = identity_dot - reg * identity
    inv = np.linalg.inv(np.dot(data_t, data))
    identity_dot = np.dot(inv, identity_dot)
    return identity_dot

def sum_error(data, identity):
    data = np.array(data)
    identity = np.array(identity)
    error = 0
    for d in data:
        error += np.sum((d - identity)**2)
    return error
```

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[1]:
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[1]:
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hence according to the
following code, the correct
answer is [c] as it is
closest.

(a)

6

CC

6

✓ 65 items

Hence according to the
above code correct
answer is [d].