1. c

1.
$$\min_{w} \frac{1}{2} \int_{0}^{2} (e^{x} - \omega x)^{2} dx$$

$$= \min_{w} \frac{1}{2} e^{2x} + \frac{1}{3} w^{2} x^{3} - 2we^{x} (x-1) \Big|_{0}^{2}$$

$$= \min_{w} \frac{1}{3} w^{2} - \frac{1}{3} (e^{2} + 1) w + \frac{1}{2} e^{y} - \frac{1}{2}$$

$$= \min_{w} (w - \frac{3}{3} (e^{2} + 1))^{2} + \frac{1}{2} e^{y} - \frac{1}{2} - \frac{1}{64} e^{2} + 1)^{2}$$

when $w = \frac{3+3e^{2}}{8}$, there will be squared error with win value &

2. b

A(D) =
$$a(g_{\text{min}} \text{ Ein}(h))$$

$$A^{\dagger}(D) = avg_{\text{min}} \text{ Eintout}(h)$$

$$h_{\text{CH}}$$

$$E_{D}[E_{\text{in}}(AD))] \leq E_{D}[E_{\text{in}}(A^{\dagger}(D))]$$

$$E_{D}[E_{\text{out}}(A^{\dagger}(D))] \leq E_{D}[E_{\text{out}}(A^{\dagger}(D))]$$
Since $A^{\dagger}(D)$ is generated by considering both Pata in and out-samples
$$E_{D}[E_{\text{in}}(A^{\dagger}(D))] = E_{D}[E_{\text{out}}(A^{\dagger}(D))]$$

3. d

$$X_{h}^{T} = \begin{bmatrix} x_{1} \times x_{2} & x_{1} \in X_{2} + \xi & x_{1} + \xi \end{bmatrix}$$

$$X_{h}^{T} = \begin{bmatrix} x_{1} \times x_{2} & x_{1} + \xi & x_{1} + \xi \\ x_{1} + \xi & x_{1} \end{bmatrix}$$

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$$X_{h}^{T} = \begin{bmatrix} x_{1} \times x_{1} & x_{1} & x_{1} \\ x_{1} & x_{1} \end{bmatrix}$$

$$X_{h}^{T} = \begin{bmatrix} x_{$$

4. e

$$\begin{array}{l} \chi_{h}^{T} = \left[\chi_{1}\chi_{2} \cdot \chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{2} \cdot \chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{2} \cdot \chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{N} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \text{when} \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot \chi_{1} + 2 \cdot \chi_{1} + \epsilon\right]_{d+1} \times 2N \quad \chi_{h}^{T} = \left[\chi_{1}\chi_{1} + 2 \cdot$$

5. d

$$N = (z^{T}z + \lambda I)^{T}z^{T}y, \quad z = XQ, \quad x^{T}x = Q^{T}Q^{T}, \quad a^{T} = a^{T}$$

$$\lambda = 0, \quad (z^{T}z)^{-1} = ((xQ^{T}XQ)^{-1} = (Q^{T}X^{T}XQ)^{-1} = Q^{T}(x^{T}x^{-1})$$

$$= Q^{T}Q^{T}Q^{T}Q = P^{T} \quad ((x^{T}x)^{-1} = Q^{T}Q^{T})$$

$$\lambda > 0, \quad (z^{T}z + \lambda I)^{T} = (P + \lambda I)^{T}$$

$$P = \begin{bmatrix} F_{0} & F_{1} & F_{2} & F_{3} \\ F_{1} & F_{2} & F_{3} \end{bmatrix}, \quad (P + \lambda I)^{T} = \begin{bmatrix} F_{0} + \lambda I \\ F_{1} & F_{2} & F_{3} \end{bmatrix}$$

$$\frac{M^{2}}{V_{1}} = \frac{I_{1}}{I_{2}} = \frac{I_{2}}{I_{1}}$$

$$\frac{M^{2}}{I_{2}} = \frac{I_{2}}{I_{1}} = \frac{I_{2}}{I_{2}}$$

6. a

6.
$$\sqrt{|\Sigma(w \times n - y_n)^2 + \Sigma w^2} = |\Sigma \times n(w \times n - y_n)| + |\Sigma w| = 0.$$

$$w(\Sigma x_n^2 + \Sigma) = |\Sigma \times n y_n|, \quad w' = |\Sigma \times n y_n| = |\Sigma$$

7. d*

Following photo shows the regular deduction, while there is another straightforward view.

Since we all know that the prior of the fair coin flip is half for two sides. To make prediction more close to prior probability, when y = 0.5, target function will receive the least penalty.

$$\frac{1}{dy} \Rightarrow \frac{2}{N} \sum_{n=1}^{N} \frac{1}{y-y_n} + \frac{2k}{N} \frac{d\Omega(y)}{dy} = 0 \qquad \frac{2}{N} \left(Ny - \sum_{n=1}^{N} \frac{1}{y_n} + k \frac{d\Omega(y)}{dy} \right) = 0.$$

$$\frac{1}{N} \sum_{n=1}^{N} \frac{1}{y_n} + k \frac{d\Omega(y)}{dy} = 0 \qquad \frac{2}{N} \sum_{n=1}^{N} \frac{1}{y_n} + k \frac{d\Omega(y)}{dy} = 0.$$

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$$\frac{1}{N} \sum_{n=1}^{N} \frac{1}{y_n} + k \frac{d\Omega(y)}{dy} = 0 \qquad \frac{2}{N} \sum_{n=1}^{N} \frac{1}{y_n} + k \frac{d\Omega(y)}{dy} = 0.$$

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8. b

9. b

9.
$$\lambda \stackrel{?}{\leq} B : \omega :^2 = \omega^T \lambda B \omega$$

$$\stackrel{!}{\leq} (\omega \times n - \hat{y}_n)^2 = \sum (\omega \times n)^2 - 2 \sum (\omega \times n)^2 + \sum (\omega \times n)^2$$

$$= \omega^T \hat{x}^T \hat{x} \omega + \hat{y} \hat{y}^T = 2 (\hat{x} \omega)^T \hat{y}$$

when $\hat{x} = \sqrt{n} \cdot \sqrt{$

10. e

When selecting each point in data, the class for validation is always different from the majority of sub training set. For example, choosing +1 as validation point, and the A_majority will return -1 since there are N-1 positive and N negative points in training set.

11. c

12. e

12. constant hypothesis

valid
$$h(x)$$
 error

 $(e,2)$ $h(x) = 0$, $e_1 = 2^2 = 4$
 $(3,0)$ $h(x) = 1$, $e_2 = 1^2 = 1$
 $(-3,0)$ $h(x) = 1$, $e_3 = 1$
 $(3,0)$ $h(x) = \frac{2}{e+3}(x+3)$ $e_3 = \frac{1}{(e+3)^2}$
 $(3,0)$ e_4 e_4

13. d

13. Var [Eval(h)] =
$$Var \left(\frac{1}{K} \sum err(h(x), y) \right) = \frac{1}{K^2} Var \left(\sum err(h(x), y) \right)$$

$$= \frac{1}{K^2} \sum Var \left(err(h(x), y) \right) = \frac{1}{K^2} \cdot K \quad Var \left(err(h(x), y) \right)$$

$$= \frac{1}{K^2} \left[Var \left(err(h(x), y) \right) - \frac{1}{K^2} \cdot K \quad Var \left(err(h(x), y) \right) \right]$$

$$= \frac{1}{K^2} \cdot Var \left(err(h(x), y) \right)$$

14. c

15. a

15.

$$P \cdot \xi_{+} + (-P) \cdot \xi_{-} = 1-P$$

 $P \cdot (\xi_{+} - \xi_{-} + 1) = 1 - \xi_{+}$
 $P = \frac{1 - \xi_{-}}{\xi_{+} - \xi_{-} + 1} *$

16. b

a. Shell script

```
./train -s 0 -c 0.00005 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt.model result16

./train -s 0 -c 0.005 -e 0.000001 -q hw4_test_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt.model result16

./train -s 0 -c 0.5 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt.model result16

./train -s 0 -c 50 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt.model result16

./train -s 0 -c 5000 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt.model result16
```

b. Termianl message

```
yiwenlai@YiWens-MBP ~/Desktop/liblinear-2.42 INSERT bash hw4_q16.sh

Accuracy = 51.6667% (155/300)

Accuracy = 51.6667% (155/300)

Accuracy = 80.6667% (242/300)

Accuracy = 87% (261/300)

Accuracy = 86.6667% (260/300)
```

17. a

a. Shell script

```
./train -s 0 -c 0.00005 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_train_convert.txt hw4_train_convert.txt.model result17
./train -s 0 -c 0.005 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_train_convert.txt hw4_train_convert.txt.model result17
./train -s 0 -c 0.5 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_train_convert.txt hw4_train_convert.txt.model result17
./train -s 0 -c 50 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_train_convert.txt hw4_train_convert.txt.model result17
./train -s 0 -c 5000 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_train_convert.txt hw4_train_convert.txt.model result17
```

b. Terminal messge:

```
yiwenlai@YiWens-MBP ~/Desktop/liblinear-2.42 INSERT bash hw4_q17.sh Accuracy = 46.5% (93/200)
Accuracy = 80.5% (161/200)
Accuracy = 87% (174/200)
Accuracy = 90% (180/200)
Accuracy = 91% (182/200)
```

18. e.

a. Run the shell script first

```
python3 hw4_q18.py
./train -s 0 -c 0.00005 -e 0.000001 -q hw4_sub_train.txt
./predict -b 1 hw4_val.txt hw4_sub_train.txt.model result18
./train -s 0 -c 0.005 -e 0.000001 -q hw4_sub_train.txt
./predict -b 1 hw4_val.txt hw4_sub_train.txt.model result18
./train -s 0 -c 0.5 -e 0.000001 -q hw4_sub_train.txt
./predict -b 1 hw4_val.txt hw4_sub_train.txt.model result18
./train -s 0 -c 50 -e 0.000001 -q hw4_sub_train.txt
./predict -b 1 hw4_val.txt hw4_sub_train.txt.model result18
./train -s 0 -c 5000 -e 0.000001 -q hw4_sub_train.txt
./predict -b 1 hw4_val.txt hw4_sub_train.txt.model result18
./train -s 0 -c 5000 -e 0.000001 -q hw4_sub_train.txt
```

b. Code(hw4_q18.py) to process data in q18:

```
import numpy as np
sub_train = []
val = []
with open("hw4_train_convert.txt") as f:
    lines = f.readlines()
    for i in range(len(lines)):
        if i >= 120:
            val.append(lines[i])
        else:
```

```
sub_train.append(lines[i])
with open("hw4_sub_train.txt",'w') as f1:
    for row in sub_train:
        f1.write(row)
with open("hw4_val.txt",'w') as f2:
    for row in val:
        f2.write(row)
```

c. After running the shell script, we choose the \lambda with highest accuracy, and train with it again and test on the whole test set. 1 minus the accuracy is the answer (1 - 85.667% := 0.143).

19. d

a. Shell Script

```
./train -s 0 -c 50 -e 0.000001 -q hw4_train_convert.txt
./predict -b 1 hw4_test_convert.txt hw4_train_convert.txt.model result19
```

b. Terminal message: 1 minus the highest accuracy is the answer (1 - 87% "= 0.13).

```
yiwenlai@YiWens-MBP ~/Desktop/liblinear-2.42 INSERT bash hw4_q19.sh
Accuracy = 87% (261/300)
```

20. c, 0.12

a. Table corresponding to each log10(\lamda), and last row is for the E cv

4	2	0	-2	-4
17	31	32	34	35
26	37	36	32	31
19	34	36	38	38
16	30	32	34	31
18	32	33	38	36
19.2	32.8	33.8	35.2	34.2
0.52	0.18	0.155	0.12	0.145

b. Shell Script

```
thon3 hw4_q20.py @
/train -s 0 -c 0.00005 -e 0.000001 -q hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
oython3 hw4_q20.py 1
./train -s 0 -c 0.00005 -e 0.000001 -q hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
ython3 hw4_q20.py 2
/train -s 0 -c 0.00005 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
oython3 hw4_g20.py 3
/train -s 0 -c 0.00005 -e 0.000001 -q hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
ython3 hw4_q20.py 4
/train -s 0 -c 0.00005 -e 0.000001 -q hw4_sub_train_20.txt
predict -b 1 hw4 val 20.txt hw4 sub train 20.txt.model result20.
ovthon3 hw4 a20.pv 0
/train -s 0 -c 0.005 -e 0.000001 -q hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
ython3 hw4_q20.py 1
/train -s 0 -c 0.005 -e 0.000001 -q hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
ython3 hw4_g20.py 2
/train -s 0 -c 0.005 -e 0.000001 -q hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
ython3 hw4_g20.py 3
/train -s 0 -c 0.005 -e 0.000001 -q hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
vthon3 hw4 g20.pv 4
/train -s 0 -c 0.005 -e 0.000001 -g hw4_sub_train_20.txt
/predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
ython3 hw4_q20.py 0
./train -s 0 -c 0.5 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_g20.py 1
./train -s 0 -c 0.5 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
ython3 hw4_q20.py 2
./train -s 0 -c 0.5 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 3
./train -s 0 -c 0.5 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 4
./train -s 0 -c 0.5 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 0
./train -s 0 -c 50 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 1
./train -s 0 -c 50 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 2
./train -s 0 -c 50 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4 g20.py 3
./train -s 0 -c 50 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_g20.py 4
/train -s 0 -c 50 -e 0.000001 -q hw4_sub_train_20.txt
 /predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
```

```
python3 hw4_q20.py 0
./train -s 0 -c 5000 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 1
./train -s 0 -c 5000 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 2
./train -s 0 -c 5000 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 3
./train -s 0 -c 5000 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
python3 hw4_q20.py 4
./train -s 0 -c 5000 -e 0.000001 -q hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt
./predict -b 1 hw4_val_20.txt hw4_sub_train_20.txt.model result20
```

c. Terminal Message:

```
yiwenlai@YiWens-MBP ~/Desktop/liblinear-2.42 INSERT bash hw4_q20.sh
Accuracy = 40\% (16/40)
Accuracy = 45\% (18/40)
start
Accuracy = 85\% (34/40)
Accuracy = 75\% (30/40)
Accuracy = 80\% (32/40)
start
Accuracy = 80\% (32/40)
Accuracy = 80\% (32/40)
Accuracy = 82.5\% (33/40)
start
Accuracy = 85\% (34/40)
Accuracy = 80\% (32/40)
Accuracy = 95\% (38/40)
Accuracy = 85\% (34/40)
Accuracy = 95\% (38/40)
start
Accuracy = 87.5\% (35/40)
Accuracy = 77.5\% (31/40)
Accuracy = 95\% (38/40)
Accuracy = 77.5\% (31/40)
Accuracy = 90\% (36/40)
```

d. Code to process data in q20:

```
import sys
start_index = 0
end_index = 0
sub_train = []
```

```
val = []
if len(sys.argv) < 2:
   print('no argument')
   sys.exit()
else:
   if sys.arqv[1] == '0':
       print('start')
   start index = 40 * int(sys.argv[1])
   end index = start index + 39
with open("hw4 train convert.txt") as f:
   lines = f.readlines()
   for i in range(len(lines)):
       if start index <= i <= end index:</pre>
           val.append(lines[i])
           sub train.append(lines[i])
with open("hw4 sub train 20.txt",'w') as f1:
   for row in sub train:
       f1.write(row)
with open("hw4 val 20.txt",'w') as f2:
   for row in val:
       f2.write(row)
```

Code to process all training and testing data in right format:

```
nums = line.strip().split()
        data = [float(num) for num in nums]
        transform.append(data[len(data)-1])
        transform.append(1)
        for i in range(6):
            transform.append(data[i])
        for i in range(6):
            for j in range (6-i):
                transform.append(data[i]*data[i+j])
        text.append(transform)
    f.close()
with open(outputfile[x],'w')as f1:
    for t in text:
        row text = ""
        for i in range(len(t)):
            if i == 0:
                if t[i] == 1:
                    row text += ("+1 ")
                    row text += ("-1 ")
            else:
                row text += (str(i)+":")
                row text += str(t[i])
        f1.write(row text+"\n")
    f1.close()
```