

Homework 5

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1

(a)

```
1 from itertools import combinations
2 import numpy as np
3 from sklearn.linear_model import LinearRegression
4 from sklearn.model_selection import train_test_split
5
6 X = np.random.rand(30, 5)
7 y = X[:, [3]] * 2
8 Xtr, Xts, ytr, yts = train_test_split(X, y)
9
10
11 scores = []
12 for i in range(Xtr.shape[1]):
13     model = LinearRegression() # Create a linear regression model object
14     model.fit(Xtr[:, [i]], ytr) # Fits the model
15     yhat = model.predict(Xts[:, [i]])
16     scores.append(np.mean((yhat - yts)**2))
17
18 print(f"Best model is order {np.argmin(scores)}, rss is {np.min(scores)}")
19
```

(b)

```
1 from itertools import combinations
2 import numpy as np
3 from sklearn.linear_model import LinearRegression
4 from sklearn.model_selection import train_test_split
5
6 X = np.random.rand(30, 5)
7 y = X[:, [4]] * 2 + X[:, [2]] * 6
8 Xtr, Xts, ytr, yts = train_test_split(X, y)
9
10
11 scores = []
12 columns = []
13 for i in combinations(range(Xtr.shape[1]), 2):
14     model = LinearRegression() # Create a linear regression model object
15     model.fit(Xtr[:, i], ytr) # Fits the model
16     yhat = model.predict(Xts[:, i])
```

```

17     scores.append(np.mean((yhat - yts)**2))
18     columns.append(i)
19
20     print(f"Best model is order {columns[np.argmin(scores)]}, rss is {np.min(scores)}")

```

(c)

As is shown in previous code, training times is $\binom{p}{k}$. For $k = 10$ and $p = 1000$, total training times is $\binom{1000}{10} \approx 2.634 \times 10^{23}$.

2

(a)

$$\phi(\mathbf{w}) = 0$$

(b)

$$\phi(\mathbf{w}) = \sum_{i=1}^N e^{-aw_i}$$

(c)

$$\phi(\mathbf{w}) = \sum_{i=2}^N |w_i - w_{i-1}|$$

(d)

$$\phi(\mathbf{w}) = \sum_{i=2}^N |\text{sgn}(w_i - w_{i-1})|$$

4

```

1  def normalize(x):
2      return (x - np.mean(x, axis=0))/np.std(x, axis=0)
3
4  Xtr = normalize(Xtr)
5  ytr = normalize(ytr)
6  Xts = normalize(Xts)
7  yts = normalize(yts)
8
9  model = SomeModel() # Creates a model
10 model.fit(Xtr,ytr) # Fits the model, expecting normalized features
11 yhat = model.predict(Xts) # Predicts targets given features
12 Rss = np.mean((yhat - yts)**2)

```

5

```
1  alphas = np.random.uniform(a, b, p)
2
3  Ztr = np.exp(-Xtr*alphas)
4  Zts = np.exp(-Xts*alphas)
5
6  model = Lasso(lam=lam)
7  beta = model.fit(Ztr, ytr)
8  yhat = model.predict(Zts)
9  rss = np.mean((yhat - yts)**2)
10
11 beta_tide = model.coef_
12 opt_alpha = alphas[np.argsort(-beta_tide) < p] # select biggest p in beta_tide
13 opt_beta = beta_tide[np.argsort(-beta_tide) < p]
```