Lecture 18 - Probabability and Combinatorics Examples



DSC 40A, Winter 2024

Announcements

- Homework 6 is posted and due next Wednesday.
- HDSI undergrad & faculty mixer will be this afternoon 3-5pm at HDSI patio
 - Light refreshment will be provided

Agenda

- Invited Algorithm Presentation
- Review of combinatorics.
- Lots of examples.

Invited Algorithm Presentation: Owen Shi

HW4 Algorithm

Owen Shi





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In [8]: waveforms.plot(kind='scatter', x='tDrift90', y='Energy');

tDrift90

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The Framework

In [6]: def observation_vector():
 return waveforms['Energy']

```
In [31]: np.random.seed(10)
         lambdas = np.logspace(-10, 10, 100)
         mses = []
         for 1 in lambdas:
             mse = 0
             W = W_{star}(1)
             for i in range(waveforms.shape[0]):
                 row = waveforms.iloc[i]
                 feat1 = 1 / row[4] + 1 / row[5]
                 feat2 = 1 / row[5] * 1 / feat1
                 X = pd.Series([1, row[0], feat1, feat2])
                 pred = X 🙆 w
                 mse += (pred - row['Energy']) ** 2
             mse /= waveforms.shape[0]
             mses.append(mse)
         lambdas[np.argmin(mses)]
Out[31]: 4.132012400115335e-09
 In [ ]: best_lambda = 0
```

```
[7]: def w_star(lam):
    X = design_matrix(waveforms)
    y = observation_vector()
    return np.linalg.inv(X.T @ X + lam * np.eye(X.shape[1])) @ X.T @ y
```

```
def design_matrix(d):
    df = d.copy()
    df['Intercept'] = 1
    feat1 = 1 / df['blnoise'] + 1 / df['tslope']
    feat2 = 1 / df['tslope'] * 1 / (1 / df['blnoise'] + 1 / df['tslope'])
    feat3 = df['tslope'] / df['Max_Amp']
    df['feat1'] = feat1
    df['feat2'] = feat2
    df['feat3'] = feat3
    return df.get(['Intercept', 'Max_Amp', 'feat1', 'feat2', 'feat3']).to_numpy()
```

```
def design_matrix(d):
    df = d.copy()
    df['Intercept'] = 1
    feat1 = (1 / df['blnoise'] + 1 / df['tslope']) / df['Max_Amp']
    feat2 = df['Max_Amp'] / df['blnoise']
    feat3 = df['Max_Amp'] / df['tslope']
    df['feat1'] = feat1
    df['feat2'] = feat2
    df['feat3'] = feat3
    return df.get(['Intercept', 'Max_Amp', 'feat1', 'feat2', 'feat3']).to_numpy()
```

Submission History

Submitted On (PST) Submitters Score Active



without_f1 = 793.7921119778865
without_f2 = 766.2772193373274
without_f3 = 727.4278430354087

```
In [5]: def design matrix(d):
            df = d.copy()
            df['Intercept'] = 1
            feat1 = (1 / df['blnoise']**2 + 1 / df['tslope']) / df['Max_Amp']
            feat2 = (1 / df['blnoise'])**1.5 / (df['Max Amp'] + df['blnoise']**3)
            feat3 = (df['Max Amp'] * df['blnoise'] * df['tslope']) / \
                    (df['Max Amp']**2 + 1 / df['blnoise'] + 1 / df['tslope'])
            df['feat1'] = feat1
            df['feat2'] = feat2
            df['feat3'] = feat3
             df.plot(kind='scatter', y='Energy', x='feat3')
        #
            return df.get(['Intercept', 'Max Amp', 'feat1', 'feat2', 'feat3']).to numpy()
        design matrix(waveforms)
```

One More Extra Credit Opportunity

- Building a Naive Bayes classifier to separate neutrino signals from unwanted noises!
 - This one will be **Optional:** chances to earn extra credit, but does not count as part of homework problem.
 - Will be released and due together wit HW7
 - More details in the following weeks.
- The full HPGe dataset is released at https://zenodo.org/records/8257027
 - In raw waveform format, no extracted parameters.

Extra Credit Rules

- The classifier competition will earn you up to 10% extra credit on Midterm 2, depending on your leaderboard ranking
 - Same as the energy regression challenge
- However, the maximum extra credit you can earn from both challenges is capped at 10%
- Example: Owen ranked No. 2 on regression challenge, he will get 9% EC on Midterm 1, so the maximum amount of EC he can get on Midterm 2 is 1%
- This is to encourage students who did not get EC from the regression challenge to participate.

Review of combinatorics

Combinatorics as a tool for probability

- ► If S is a sample space consisting of equally-likely outcomes, and A is an event, then $P(A) = \frac{|A|}{|S|}$.
- In many examples, this will boil down to using permutations and/or combinations to count |A| and |S|.
- Tip: Before starting a probability problem, always think about what the sample space S is!

Sequences

- A sequence of length k is obtained by selecting k elements from a group of n possible elements with replacement, such that order matters.
- **Example:** You roll a die 10 times. How many different sequences of results are possible?



In general, the number of ways to select *k* elements from a group of *n* possible elements such that **repetition is allowed** and **order matters** is

n^k.

Permutations

- A permutation is obtained by selecting k elements from a group of n possible elements without replacement, such that order matters.
- **Example:** How many ways are there to select a president, vice president, and secretary from a group of 8 people?

Permutations

In general, the number of ways to select k elements from a group of n possible elements such that repetition is not allowed and order matters is

$$P(n,k) = (n)(n-1)...(n-k+1)$$
$$= \frac{n!}{(n-k)!}$$

Combinations

- A combination is a set of k items selected from a group of n possible elements without replacement, such that order does not matter.
- Example: How many ways are there to select a committee of 3 people from a group of 8 people?

Combinations

In general, the number of ways to select *k* elements from a group of *n* elements such that **repetition is not allowed** and **order does not matter** is

$$C(n,k) = \binom{n}{k}$$
$$= \frac{P(n,k)}{k!}$$
$$= \frac{n!}{(n-k)!k!}$$

The symbol $\binom{n}{k}$ is pronounced "*n* choose *k*", and is also known as the **binomial coefficient**.

Lots of examples

Discussion Question

A domino consists of two faces, each with anywhere between 0 and 6 dots. A set of dominoes consists of every possible combination of dots on each face. How many dominoes are in the set of dominoes?

a)
$$\binom{7}{2}$$

b) $\binom{7}{1} + \binom{7}{2}$
c) $P(7, 2)$
d) $\frac{P(7, 2)}{P(7, 1)}7!$

Selecting students — overview

We're going answer the same question using several different techniques.

Selecting students (Method 1: using permutations)

Selecting students (Method 2: using permutations and the complement)

Question 1, Part 1 (Denominator): If you draw a sample of size 5 at random without replacement from a population of size 20, how many different **sets** of individuals could you draw?

Question 1, Part 2 (Numerator): If you draw a sample of size 5 at random without replacement from a population of size 20, how many different **sets** of individuals include Avi?

Selecting students (Method 4: "the easy way")

With vs. without replacement

Discussion Question

We've determined that a probability that a random sample of 5 students from a class of 20 **without replacement** contains Avi (one student in particular) is $\frac{1}{4}$. Suppose we instead sampled **with replacement**. Would the resulting probability be equal to, greater than, or less than $\frac{1}{4}$?

- a) Equal to
- b) Greater than
- c) Less than

Summary

Summary

- A sequence is obtained by selecting k elements from a group of n possible elements with replacement, such that order matters.
 - Number of sequences: n^k .
- A permutation is obtained by selecting k elements from a group of n possible elements without replacement, such that order matters.

Number of permutations: $P(n, k) = \frac{n!}{(n-k)!}$.

A combination is obtained by selecting k elements from a group of n possible elements without replacement, such that order does not matter.

Number of combinations:
$$\binom{n}{k} = \frac{n!}{(n-k)!k!}$$
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