Report.md

Homework: 2024/9/11

1.

```
1. Assume Y_1 = X, Y_2 = X^2, X \sim N(0,1)

Cov(Y_1, Y_2) = E[Y_1Y_2] - E[Y_1]E[Y_2]
= E[X^3] - E[X]E[X^3] = 0
Cov(Y_1, Y_2) = 0 \quad \text{means} \quad Z \quad \text{is} \quad a \quad \text{diagonal} \quad \text{matrix}, \quad \text{but} \quad Y_1 \quad \text{and} \quad Y_2 \quad \text{are} \quad \text{not} \quad \text{independent}
Assume \quad Y_1 = X + Z, \quad Y_2 = X - Z, \quad X, Z \sim N(0,1)
Cov(Y_1, Y_2) = E[X^2 - Z^2] = 0
\therefore \text{ In this case, } Z \quad \text{is a diagonal} \quad \text{matrix, but} \quad Y_1 \quad \text{and} \quad Y_2 \quad \text{are also not} \quad \text{independent}
```

2.

The distribution of Y can be written as

$$q(y_1, 0) = \frac{1}{(2\pi)^{\frac{1}{2}} det(\Sigma)^{\frac{1}{2}}} \exp\left(-\frac{1}{2}(y_1 - M)^2 \sum_{i=1}^{-1} (y_1 - M)^2\right)$$
and the distribution of Yi respectively is

$$f(y_i) = \frac{1}{(2\pi)^{\frac{1}{2}} 6} \exp\left(-\frac{1}{2}(\frac{y_1 - M}{6})^2\right)$$

$$\therefore q(y_1, 0) = \prod_{i=1}^{M} f(y_i), \text{ which means } Y_i, Y_j \text{ are independent for all } i \neq j, i.e., Y_i \text{ is a vector of independent}$$
variables, $0 \in V$.

3.

3. Since
$$Y'Y = \sum_{i=1}^{M} Y_i^{\pm}$$
 and $Y_i^{\pm} \sim e^2 X^{\pm}$

$$\frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = M e^{2i}$$
For $(Y'Y)^{\frac{1}{2}} = \left(\sum_{i=1}^{M} Y_i^{\pm} \right)^{\frac{1}{2}} = \sum_{i=1}^{M} Y_i^{i+} + 2 \sum_{i=1}^{M} \sum_{j=1}^{i} Y_i^{i} Y_j^{\pm}$

$$\frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = 3 e^{4i}$$

$$\frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = 2 \cdot {\binom{M}{2}} \cdot \frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = 3 m e^{4i}$$

$$\frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = 2 \cdot {\binom{M}{2}} \cdot \frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = 3 m e^{4i}$$

$$\frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = 2 \cdot {\binom{M}{2}} \cdot \frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = 3 m e^{4i}$$

$$\frac{1}{2} \left[\sum_{i=1}^{M} Y_i^{\pm} \right] = m e^{2i}$$

$$\frac{1}{2} \left[\sum_{i=1}^{M}$$

Report.md 2024-09-17

4.

Apply the law of total probability, we can count each probability that satisfies the condition:

```
[lin1214@archlinux hw2]$ Rscript homework2.r
[1] "alpha_1 = 0.091"
[1] "alpha_2 = 0.909"
[1] "beta_11 = 0.056"
[1] "beta_12 = 0.036"
[1] "beta_21 = 0.478"
[1] "beta_22 = 0.431"
[1] "gamma_1 = 0.135"
[1] "gamma_2 = 0.865"
[1] "beta = -0.00102822569589241"
```

5.

```
4. Arg min (E[(m(X) - X'b)^2]) to find B, we can calculate \frac{d}{db}(E[(m(X) - X'b)^2]) = 0

\frac{d}{db}(E[m(X)^2 - 2m(X)X'b + (X'b)^2]) = 2E[XX'b] - 2E[m(X)X'] = 0

Apply the law of iterated expectations, we have E[m(X)X'] = E[E[Y|X]X'] = E[XY]

E = E[XX']^{-1} E[XY]
E = E[XX']^{-1} E[XY]
```

Report.md 2024-09-17

6.

```
The linear projection coefficient \mathcal{B} can be written as

\mathcal{B} = \frac{E[XY]}{E[X']}, \quad X \sim t(3), \quad Y = \frac{1}{1 + X^{4}}

To have \mathcal{B} be caculated properly, we must compute the second momentum of X, which can be written as E[X^{2}] = \frac{D}{D-2}, \quad X \sim t(D)

The degree of freedoms must be set to D > 2, in this case, D is set to D > 2.
```

We can then calculate beta by constructing the linear model with X and Y:

7. Source Code

Source Code