

P1

- (a) Centering: No. Because centering only changes the position of the distribution
Standardization: Yes. Because the variance of the distribution has changed.
- (b) Centering: No. Because distance between the points have not changed.
Standardization: Yes. Because

$$D = \sqrt{\sum_{i=1}^d (u_i - v_i)^2}$$

Upon Standardization,

$$D = \frac{\sqrt{\sum_{i=1}^d (u_i - v_i)^2}}{\hat{\sigma}_i}$$

- (c) Centering: No, because the relative position of the datas and the median of the datas have not changed.
Standardization: No, because the median of the points have not changed.
- (d) Centering: No. Because as the data moves to the center, the classifier doesnt need to change
Standardization: No, the data will still be linearly seperable after standardization

P2

- (a) The forth data representation I choose is bigram representation without any unigram features.
- (b) PLA-Unigram: 11.3575% / 11.4475% / 10.7775% / 11.8275% / 11.3025% / avg = 11.3425%
PLA-Bigram : 9.8225% / 9.885% / 9.935% / 10.2375% / 9.345% / avg = 9.845%
PLA-Chosen : 12.325% / 12.68% / 11.9725% / 12.045% / 12.17% / avg = 12.2385%
PLA-Tfidf : 11.4325% / 11.2575% / 11.5125% / 11.26% / 11.4975% / avg = 11.392%
Bayes-Unigram: 14.005% / 14.315% / 14.205% / 14.065% / 14.0475% / avg = 14.1275%
- (c) Bigram representation
- (d) Training ER = 8.42% , Test ER = 11.23%

P3

(a)

$$\mathcal{L} = -\frac{n}{2} \log(2\pi\sigma^2) - \sum_{i=1}^n \frac{(x_i - \mu)^2}{2\sigma^2}$$

Get the derivative respect to σ .

$$0 = -\frac{n}{2} \cdot \frac{1}{\sigma^2} + \frac{\sum_{i=1}^n (x_i - \mu)^2}{2\sigma^4}$$

$$0 = -n\sigma^2 + \sum_{i=1}^n (x_i - \mu)^2$$

$$\sigma^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2$$