Covariance Sensitivity Proofs

March 21, 2020

1 Preliminaries

Lemma 1. $\forall i$,

$$\sum_{j=1}^{n} (x_{ji} - \bar{X}_i) = 0.$$

Proof.

$$\sum_{i=1}^{n} (x_{ji} - \bar{X}_i) = \sum_{i=1}^{n} x_{ji} - n\bar{X}_i$$

$$= \sum_{i=1}^{n} x_{ji} - n \left(\frac{1}{n} \sum_{i=1}^{n} x_{ji}\right)$$

$$= 0.$$

Lemma 2. Let

$$f_{ij} = \sum_{k=1}^{n} (x_{ki} - \bar{X}_i)(x_{kj} - \bar{X}_j),$$

and let $X'_i = X_i \cup \{y_i\}$, and say X_i has size n. Let \bar{X}'_i and \bar{X}'_j be the sample means of X'_i and X'_j respectively. Then,

$$f_{ij}(X') = f_{ij}(X) + n(\bar{X}_i - \bar{X}_i')(\bar{X}_j - \bar{X}_i') + (y_i - \bar{X}_i)(y_j - \bar{X}_j).$$

Proof. Note that

$$f_{ij}(X') = \sum_{k=1}^{n+1} (x'_{ki} - \bar{X}'_i)(x'_{kj} - \bar{X}'_j),$$

$$= \sum_{k=1}^{n} (x_{ki} - \bar{X}'_i)(x_{kj} - \bar{X}'_j) + (y_i - \bar{X}'_i)(y_j - \bar{X}'_j),$$

$$= \sum_{k=1}^{n} ((x_{ki} - \bar{X}_i) + (\bar{X}_i - \bar{X}'_i)) ((x_{kj} - \bar{X}_j) + (\bar{X}_j - \bar{X}'_j)) + (y_i - \bar{X}'_i)(y_j - \bar{X}'_j),$$

$$= \sum_{k=1}^{n} (x_{ki} - \bar{X}_i)(x_{kj} - \bar{X}_j) + (\bar{X}_j - \bar{X}'_j) \sum_{k=1}^{n} (x_{ki} - \bar{X}_i) + (\bar{X}_i - \bar{X}'_i) \sum_{k=1}^{n} (x_{kj} - \bar{X}_j),$$

$$+ \sum_{k=1}^{n} (\bar{X}_i - \bar{X}'_i)(\bar{X}_j - \bar{X}'_j) + (y_i - \bar{X}'_i)(y_j - \bar{X}'_j),$$

$$= f_{ij}(X) + n(\bar{X}_i - \bar{X}'_i)(\bar{X}_j - \bar{X}'_j) + (y_i - \bar{X}'_i)(y_j - \bar{X}'_j),$$

where the cancellation of the second and third terms in the second-to-last line is due to Lemma 1.

Lemma 3. Let X_i have size n and say $X_i' = X_i \cup \{y_i\}$ where $y_i \in \mathcal{X}_i$. Say that the space of datapoints \mathcal{X}_i is bounded above by M_i and bounded below by m_i , and let $y_i \in \mathcal{X}_i$. Let \bar{X}_i , \bar{X}_j , \bar{X}_i' , and \bar{X}_j' be the sample means of X_i, X_j, X_i' and X_j' respectively. Then,

$$n \left| (\bar{X}_i - \bar{X}'_i)(\bar{X}_j - \bar{X}'_j) \right| \le \frac{n}{(n+1)^2} (M_i - m_i)(M_j - m_j).$$

Proof. Note that

$$n \left| (\bar{X}_i - \bar{X}_i')(\bar{X}_j - \bar{X}_j') \right| = n \left| \left(\frac{1}{n} \sum_{k=1}^n x_{ki} - \frac{1}{n+1} \sum_{k=1}^{n+1} x_{ki}' \right) \left(\frac{1}{n} \sum_{k=1}^n x_{kj} - \frac{1}{n+1} \sum_{k=1}^{n+1} x_{kj}' \right) \right|,$$

$$= n \left| \left(\left(\frac{1}{n} - \frac{1}{n+1} \right) \sum_{k=1}^n x_{ki} - \frac{y_i}{n+1} \right) \left(\left(\frac{1}{n} - \frac{1}{n+1} \right) \sum_{k=1}^n x_{kj} - \frac{y_j}{n+1} \right) \right|,$$

$$= n \left| \left(\frac{1}{n(n+1)} \sum_{k=1}^n x_{ki} - \frac{y_i}{n+1} \right) \left(\frac{1}{n(n+1)} \sum_{k=1}^n x_{kj} - \frac{y_j}{n+1} \right) \right|,$$

$$= \frac{n}{(n+1)^2} \left| \left(\frac{1}{n} \sum_{k=1}^n x_{ki} - \frac{y_i}{n+1} \right) \left(\frac{1}{n} \sum_{k=1}^n x_{kj} - \frac{y_j}{n+1} \right) \right|,$$

$$\leq \frac{n}{(n+1)^2} (M_i - m_i) (M_j - m_j).$$

Lemma 4. Let X_i have size n and say $X_i' = X_i \cup \{y_i\}$ where $y_i \in \mathcal{X}_i$. Say that the space of datapoints \mathcal{X}_i is bounded above by M_i and bounded below by m_i , and let $y_i \in \mathcal{X}_i$. Let \bar{X}_i , \bar{X}_j , \bar{X}_i' , and \bar{X}_j' be the sample means of X_i, X_j, X_i' and X_j' respectively. Then,

$$\left| (y_i - \bar{X}_i')(y_j - \bar{X}_j') \right| \le \frac{n^2}{(n+1)^2} (M_i - m_i)(M_j - m_j).$$

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Proof. Note that

$$\begin{aligned} |(y_i - \bar{X}_i')(y_j - \bar{X}_j')| &= \left| \left(y_i - \frac{y_i + n\bar{X}_i}{n+1} \right) \left(y_j - \frac{y_j + n\bar{X}_j}{n+1} \right) \right|, \\ &= \frac{1}{(n+1)^2} \left| \left((n+1)y_i - y_i - n\bar{X}_i \right) \left((n+1)y_j - y_j - n\bar{X}_j \right) \right|, \\ &= \frac{n^2}{(n+1)^2} \left| (y_i - \bar{X}_i)(y_j - \bar{X}_j) \right|, \\ &\leq \frac{n^2}{(n+1)^2} (M_i - m_i)(M_j - m_j). \end{aligned}$$

2 NEIGHBORING DEFINITION: CHANGE ONE

- 2.1 ℓ_1 -sensitivity
- 2.2 ℓ_2 -sensitivity
 - 3 NEIGHBORING DEFINITION: ADD/DROP ONE
- 3.1 ℓ_1 -sensitivity
- 3.2 ℓ_2 -sensitivity