

Variance Sensitivity Proofs

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Definition 1. *Let variance be defined as*

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2.$$

1 NEIGHBORING DEFINITION: CHANGE ONE

1.1 ℓ_1 -sensitivity

1.2 ℓ_2 -sensitivity

2 NEIGHBORING DEFINITION: ADD/DROP ONE

2.1 ℓ_1 -sensitivity

Lemma 1. *For arbitrary a ,*

$$\sum_{i=1}^n (x_i - a)^2 = \sum_{i=1}^n (x_i - \bar{x})^2 + n(a - \bar{x})^2.$$

Proof.

$$\begin{aligned}
\sum_{i=1}^n (x_i - a)^2 &= \sum_{i=1}^n ((x_i - \bar{x}) - (a - \bar{x}))^2 \\
&= \sum_{i=1}^n ((x_i - \bar{x})^2 - 2(x_i - \bar{x})(a - \bar{x}) + (a - \bar{x})^2) \\
&= \sum_{i=1}^n (x_i - \bar{x})^2 - 2 \sum_{i=1}^n (x_i a - x_i \bar{x} - \bar{x} a + \bar{x}^2) + \sum_{i=1}^n (a^2 - 2a\bar{x} + \bar{x}^2) \\
&= \sum_{i=1}^n (x_i - \bar{x})^2 - 2a \sum_{i=1}^n x_i + 2\bar{x} \sum_{i=1}^n x_i + 2\bar{x}an - 2\bar{x}^2n + a^2n - 2a\bar{x}n + \bar{x}^2n \\
&= \sum_{i=1}^n (x_i - \bar{x})^2 + a^2n - 2a\bar{x}n + \bar{x}^2n \\
&= \sum_{i=1}^n (x_i - \bar{x})^2 + n(a - \bar{x})^2
\end{aligned}$$

□

Theorem 1. *Let*

$$f(\mathbf{x}) = \sum_{i=1}^n (x_i - \bar{x})^2.$$

Then for \mathbf{x} bounded between m and M , f has sensitivity bounded above by

$$\frac{n-1}{n}(M-m)^2.$$

Proof. Consider databases \mathbf{x}' and \mathbf{x}'' which differ in a single point. For notational ease, call \mathbf{x} the part of \mathbf{x}' and \mathbf{x}'' that is the same, and say that \mathbf{x} contains n points. WLOG say that the last data point in the database is the one that differs. I.e., $\mathbf{x}' = \mathbf{x} \cup \{x_{n+1}\}$, and $\mathbf{x}'' = \mathbf{x} \cup \{x'_{n+1}\}$. This proof assumes that a “neighboring database” is one that differs in a single data-point, so we will ultimately be comparing $f(\mathbf{x}')$ and $f(\mathbf{x}'')$. However, it is useful to first write $f(\mathbf{x}')$ in terms of $f(\mathbf{x})$. Note that

$$\begin{aligned}
\bar{x}' &= \frac{1}{n+1} \sum_{i=1}^{n+1} x_i \\
&= \frac{n\bar{x} + x_{n+1}}{n+1}.
\end{aligned} \tag{2.1}$$

Then,

$$\begin{aligned}
f(\mathbf{x}') &= \sum_{i=1}^n (x_i - \bar{x}')^2 + (x_{n+1} - \bar{x}')^2 \\
&= \sum_{i=1}^n (x_i - \bar{x})^2 + n(\bar{x}' - \bar{x})^2 + (x_{n+1} - \bar{x}')^2 && \text{(By Lemma 1)} \\
&= f(\mathbf{x}) + n \left(\frac{n\bar{x} + x_{n+1}}{n+1} - \bar{x} \right)^2 + \left(x_{n+1} - \frac{n\bar{x} + x_{n+1}}{n+1} \right)^2 && \text{(By Equation 2.1)} \\
&= f(\mathbf{x}) + n \left(\frac{x_{n+1} - \bar{x}}{n+1} \right)^2 + \left(\frac{nx_{n+1} - n\bar{x}}{n+1} \right)^2 \\
&= f(\mathbf{x}) + (x_{n+1} - \bar{x})^2 \frac{n + n^2}{(n+1)^2} \\
&= f(\mathbf{x}) + (x_{n+1} - \bar{x})^2 \frac{n}{n+1}
\end{aligned}$$

Now, to bound the sensitivity of f , note that

$$\begin{aligned}
|f(\mathbf{x}') - f(\mathbf{x}'')| &= \left| (x_{n+1} - \bar{x})^2 \frac{n}{n+1} - (x'_{n+1} - \bar{x})^2 \frac{n}{n+1} \right| \\
&\leq (M - m)^2 \frac{n}{n+1}.
\end{aligned}$$

Now, usually we're interested in sensitivities in terms of the total number of values in the database, which here is $n + 1$. So, redefining n as $n + 1$ in the above equation gives

$$(M - m)^2 \frac{n - 1}{n}.$$

□

Corollary 1. *Sample variance has sensitivity bounded above by*

$$\frac{(M - m)^2}{n}.$$

2.2 ℓ_2 -sensitivity