DISPERSION - A NEW APPROACH

E. J. Thomas

College of Agriculture, Vellayani-695 522, Kerala.

Dispersion of a set of observations x_1 , x_2 , x_3 , x_n is measured by variance, $s^2=(1/n-1)\sum\limits_{i=1}^n (x_i-\bar{x})^2$

In this paper it is shown that s^2 can be expressed in terms of the sum of squares of differences between all possible pairs of observations. The implications of this new definition are studied in a number of situations.

Results

The expression for variance $s^2 = (1/n-1) \underset{i=1}{\overset{n}{\leqslant}} (x_i - x_i)^2$ can be simplified to the form,

$$s^{2} = \left\{ \frac{1}{n(n-1)} \right\} \quad \stackrel{\leqslant}{>} \, \stackrel{\varsigma}{j} \quad (x_{i} - x_{j})^{2}$$

Proof

$$s^{2} = \frac{1}{(n-1)} = \{(x_{i} - x_{i})^{2}$$

$$= \frac{1}{(n-1)} \left\{ \begin{cases} \leq x_{i}^{2} & \leq x_{i}^{2} & - \leq \leq x_{i} \\ i & i \end{cases} \right\}$$

$$= \frac{1}{n(n-1)} = \begin{cases} \leq x_{i}^{2} & (x_{i} - x_{j})^{2} \end{cases}$$

Thus s^2 is $\frac{1}{n(n-1)}$ — times the sum of squares between pairs of observations x_i , x_j such that i>j. Further, this can be partitioned into (n-1) components d_1^2 , d_2^2 , d_{n-1}^2 , where $d^2i = \underbrace{\leq}_{i=1}^n (x_i - x_i)^2$, is the sum of squares of differences between pairs of observations which are j units apart. Thus d_1^2 is the sum of squares i=1 $(x_{i-1} - x_i)^2$. which on division by (n-1) gives the mean square successive difference D_1^2 . Thus $D_1^2 = \frac{d_1^2}{(n-1)}$ Similarly D_2^2 , D_2^2 D_2^2 can be defined as the mean square differences of orders

 $2, 3, \ldots (n-1)$, respectively, where

$$D_2^2 = \frac{d_2^2}{(n-2)}$$
, $v_3^2 = \frac{d_3^2}{(n-3)}$, $D_{n-1}^2 = d_{n-1}^2$

These are different from the mean square successive j-th difference $\frac{2}{\delta}$ where

$$\begin{array}{lll}
2 & \frac{1}{(n-1)} & \sum_{i=1}^{n-1} (x_{i+1} - x_i)^2 & D_1^2 \\
2 & \frac{1}{(n-2)} & \sum_{i=1}^{n-2} (x_{i+2} - 2x_{i+1} + x_i)^2 \\
2 & \frac{1}{(n-3)} & \sum_{i=1}^{n-3} (x_{i+3} - 3x_{i+2} + 3x_{i+1} - x_i)^3 \\
3 & \frac{2}{4} & \frac{1}{(n-4)} & \sum_{i=1}^{n-4} (x_{i+4} - 4x_{i+3} + 6x_{i+2} - 4x_{i+1} + x_i)^3
\end{array}$$

The measure of dispersion, s^2 can be represented in terms of \overrightarrow{D}_1^2 as shown below:

$$s^{2} = \frac{1}{n(n-1)} \left[(n-1) D_{1}^{2} + (n-2) D_{2}^{2} + \dots + D_{n-1}^{2} \right]$$

$$= \frac{1}{n(n-1)} \left[(n-1) D_{1}^{2} + (n-2) D_{2}^{2} + \dots + D_{n-1}^{2} \right]$$

$$= \frac{1}{n(n-1)} \sum_{i=1}^{n-1} (n-i) D_{i}^{2}$$

Hence s^a is (1/n) times the simple average of d_i^2 values or is half of the weighted average of the $D^{\frac{2}{n}}$ values.

These lead us to a number of useful results which are stated below: (a) The sum of squares of n observations x_1 , x_2 , x_3 , ... x_n namely (x_1, x_2, x_3) can be expressed as

(if)
$$\underset{i>j}{\leqslant \leqslant} (x_i - x_j)^2 = \underset{i=1}{\overset{n-1}{\leqslant}} (x_{i+1} - x_i)^3 + \underset{i+1}{\overset{n-2}{\leqslant}} (x_{i+2} - x_i)^2 + \cdots + (x_n - x_1)^2$$

$$= d_1^2 + d_2^2 + d_2^2 + d_2^2 + d_2^2 + d_2^2 + d_2^2 + \cdots$$

- (b) The sum of squares between v samplas of sizes r_1 , $r_{\overline{z}}$... r_v having sample totals T_1 , T_1 , ... T_v is given by
 - $\left(\frac{1}{N}\right)$ $\underset{i>j}{\lessapprox}$ $(r_i \ T_j \ -r_j \ T_i \)^2/r_i \ r_j$ where N is the sum $r_1+r_2+\ldots +r_v$

Obviously when sample sizes are equal [to r, this simplifies to

$$\left(\frac{1}{v_r}\right)$$
 $\lessapprox \leqslant (T_i - T_j)^2$ This shows that the SS between samples is derived as

the sum of squares of differences between all possible pairs of treatment totals adjusted by dividing by the total number of plots. The contribution of each pair can be easily seen and in fact is indicative of the distances between samples in the one-dimensional space.

(c) The formula for computation of variance in frequency tables can be expressed

as
$$(x_i - x_i)^2$$

Variance is defined as
$$\bar{s}^2 = \frac{1}{N-1} \sum_{i=1}^{N} f_i (x_i - \bar{x})^2$$

This can be brought to the form
$$s^2 = \underbrace{\text{M}(M-1)}_{i>j} \leqslant f_i .f_j (x_i - x_j)^2$$

When the classes are of uniform width c, variance can be computed by using the formula,

$$s^2 = \frac{C^2}{N(N-1)} \quad \lessapprox f_i .f_j \ (i-j)^2$$

(d) Similar to the simplified form of s^2 , the covariance between x and y or the product moment can be put in the form,

$$P_{xy} = \left\lceil \frac{1}{n-1} \right\rceil \stackrel{\text{\tiny{\leqslant}}}{\underset{i=j}{\leqslant}} (x_i - x_j) (y_i - y_j)$$

(e) In the analysis of variance of data relating to v samples of sizes r each, the

SS for total is
$$(1)^{1}$$
 $\leq \leq (y_{ij} - y_{i'j'})^{2}$

and SS between samples is $\frac{1}{rv} \stackrel{<}{<} \stackrel{<}{<} (y.j')^2.j > j$.

(f) In randomised block experiments, block SS will be,

$$\frac{1}{rv}$$
 $_{i>j}$ $(y_{i.} - y_{i'.})^2$

(g) The new method of defining the **SS** makes the method of estimation of missing plot yields simple.

Consider a randomised block experiment with v treatments and r replications. Where y_{ij} is missing the method of estimation of y_{ij} is by minimising

error SS
$$\frac{1}{rv} \left[\begin{cases} \begin{cases} \begin{cases} x_i \\ y_i \end{cases} \\ y_{ij} - y_{i'j} \end{cases}^{\frac{1}{3}} - \begin{cases} \begin{cases} (y_i - y_{i'})^{\frac{1}{2}} - \begin{cases} \\ y_{i'} \end{cases} \\ \end{cases}^{\frac{1}{2}} (y_j - y_{j'})^{\frac{1}{2}} \right]$$

This leads to the equation,

$$\underset{i'j'}{\overset{\textstyle \leq}{\underset{i''j'}{\overset{}}{\overset{}}}} (y_{ii} - y_{i'j'}) \cdot \underset{i'}{\overset{\textstyle \leq}{\underset{i''}{\overset{}}{\overset{}}}} (y_{i} - y_{i'j}) + \underset{j'}{\overset{\textstyle \leq}{\underset{i''}{\overset{}}{\overset{}}}} (y_{j} - y_{i'j}) \\
(rv-1)y_{ij} - G' = (r-1)y_{ij} - (G'-y_{ij}) + (v-1)y_{ij} \cdot - (G'-y_{ij}) \\
= r.y_{ij} - G' + v.y_{ij} - G'$$

Where G' = total of y's except from the missing plot.

Thus the estimate of
$$y_{ij}$$
 is $\frac{r.y: + v.y: -G'}{(r-1)(v-1)}$

Summary

The variance s^2 can be expressed in terms of the sum of squares of differences between all possible pairs of observations. This new approach to the definition of s^2 is useful in understanding many concepts in statistics and also provides simpler computational formulae. Implications of this new definition are studied in a number of situations.

സം ഗഹം

ഒരു നിരീക്ഷണ സമുച്ചയത്തിനെറെ വൃതിയാനം നിർദ്ദിഷ്ട സമുച്ചയത്തിലെ അംഗങ്ങളായ നിരീക്ഷണങ്ങളുടെ സാധ്യമായ എല്ലാ ജോഡികളും അന്യോന്യം പ്രകടിപ്പി കുന്ന വൃത്യാസങ്ങളുടെ വർഗ്ഗയോഗത്തിൻെറ ഒരു ഫലനമായി വ്യംജിപ്പിക്കാവുന്നതാണ്. ഈ പുതിയ ആശയം സാംഖികത്തിലെ പല സംഹിതകളും മനസ്സിലാക്കുന്നതിനും കൂടുതൽ ലഘുവായ പരികമന സൂത്രങ്ങടം ആവിഷ്കരിക്കുന്നതിനും സഹായിക്കുന്നു. ഈ പുതിയ നിർവചനത്തിൻെറ പ്രതികരണങ്ങടെ പല സാഹചര്യങ്ങളിലും പാനവിധേയമാ ക്കുകയുണ്ടായി.