



Standard Practice for Dealing With Outlying Observations¹

This standard is issued under the fixed designation E 178; the number immediately following the designation indicates the year of original adoption or, in the case of revision, the year of last revision. A number in parentheses indicates the year of last reapproval. A superscript epsilon (ϵ) indicates an editorial change since the last revision or reapproval.

1. Scope

1.1 This practice covers outlying observations in samples and how to test the statistical significance of them. An outlying observation, or “outlier,” is one that appears to deviate markedly from other members of the sample in which it occurs. In this connection, the following two alternatives are of interest:

1.1.1 An outlying observation may be merely an extreme manifestation of the random variability inherent in the data. If this is true, the value should be retained and processed in the same manner as the other observations in the sample.

1.1.2 On the other hand, an outlying observation may be the result of gross deviation from prescribed experimental procedure or an error in calculating or recording the numerical value. In such cases, it may be desirable to institute an investigation to ascertain the reason for the aberrant value. The observation may even actually be rejected as a result of the investigation, though not necessarily so. At any rate, in subsequent data analysis the outlier or outliers will be recognized as probably being from a different population than that of the other sample values.

1.2 It is our purpose here to provide statistical rules that will lead the experimenter almost unerringly to look for causes of outliers when they really exist, and hence to decide whether alternative 1.1.1 above, is not the more plausible hypothesis to accept, as compared to alternative 1.1.2, in order that the most appropriate action in further data analysis may be taken. The procedures covered herein apply primarily to the simplest kind of experimental data, that is, replicate measurements of some property of a given material, or observations in a supposedly single random sample. Nevertheless, the tests suggested do cover a wide enough range of cases in practice to have broad utility.

2. Referenced Documents

2.1 ASTM Standards:

E 456 Terminology Relating to Quality and Statistics²

¹ This practice is under the jurisdiction of ASTM Committee E11 on Quality and Statistics and is the direct responsibility of Subcommittee E11.10 on Sampling.

Current edition approved May 10, 2002. Published July 2002. Originally published as E 178 – 61 T. Last previous edition E 178 – 00.

² *Annual Book of ASTM Standards*, Vol 14.02.

3. Terminology

3.1 *Definitions*: The terminology defined in Terminology E 456 applies to this standard unless modified herein.

3.1.1 *outlier*—see **outlying observation**.

3.1.2 *outlying observation, n*—an observation that appears to deviate markedly in value from other members of the sample in which it appears.

4. Significance and Use

4.1 When the experimenter is clearly aware that a gross deviation from prescribed experimental procedure has taken place, the resultant observation should be discarded, whether or not it agrees with the rest of the data and without recourse to statistical tests for outliers. If a reliable correction procedure, for example, for temperature, is available, the observation may sometimes be corrected and retained.

4.2 In many cases evidence for deviation from prescribed procedure will consist primarily of the discordant value itself. In such cases it is advisable to adopt a cautious attitude. Use of one of the criteria discussed below will sometimes permit a clear-cut decision to be made. In doubtful cases the experimenter’s judgment will have considerable influence. When the experimenter cannot identify abnormal conditions, he should at least report the discordant values and indicate to what extent they have been used in the analysis of the data.

4.3 Thus, for purposes of orientation relative to the over-all problem of experimentation, our position on the matter of screening samples for outlying observations is precisely the following:

4.3.1 *Physical Reason Known or Discovered for Outlier(s)*:

4.3.1.1 Reject observation(s).

4.3.1.2 Correct observation(s) on physical grounds.

4.3.1.3 Reject it (them) and possibly take additional observation(s).

4.3.2 *Physical Reason Unknown—Use Statistical Test*:

4.3.2.1 Reject observation(s).

4.3.2.2 Correct observation(s) statistically.

4.3.2.3 Reject it (them) and possibly take additional observation(s).

4.3.2.4 Employ truncated-sample theory for censored observations.

4.4 The statistical test may always be used to support a judgment that a physical reason does actually exist for an outlier, or the statistical criterion may be used routinely as a basis to initiate action to find a physical cause.

5. Basis of Statistical Criteria for Outliers

5.1 There are a number of criteria for testing outliers. In all of these, the doubtful observation is included in the calculation of the numerical value of a sample criterion (or statistic), which is then compared with a critical value based on the theory of random sampling to determine whether the doubtful observation is to be retained or rejected. The critical value is that value of the sample criterion which would be exceeded by chance with some specified (small) probability on the assumption that all the observations did indeed constitute a random sample from a common system of causes, a single parent population, distribution or universe. The specified small probability is called the “significance level” or “percentage point” and can be thought of as the risk of erroneously rejecting a good observation. It becomes clear, therefore, that if there exists a real shift or change in the value of an observation that arises from nonrandom causes (human error, loss of calibration of instrument, change of measuring instrument, or even change of time of measurements, etc.), then the observed value of the sample criterion used would exceed the “critical value” based on random-sampling theory. Tables of critical values are usually given for several different significance levels, for example, 5 %, 1 %. For statistical tests of outlying observations, it is generally recommended that a low significance level, such as 1 %, be used and that significance levels greater than 5 % should not be common practice.

NOTE 1—In this practice, we will usually illustrate the use of the 5 % significance level. Proper choice of level in probability depends on the particular problem and just what may be involved, along with the risk that one is willing to take in rejecting a good observation, that is, if the null-hypothesis stating “all observations in the sample come from the same normal population” may be assumed correct.

5.2 It should be pointed out that almost all criteria for outliers are based on an assumed underlying normal (Gaussian) population or distribution. When the data are not normally or approximately normally distributed, the probabilities associated with these tests will be different. Until such time as criteria not sensitive to the normality assumption are developed, the experimenter is cautioned against interpreting the probabilities too literally.

5.3 Although our primary interest here is that of detecting outlying observations, we remark that some of the statistical criteria presented may also be used to test the hypothesis of normality or that the random sample taken did come from a normal or Gaussian population. The end result is for all practical purposes the same, that is, we really wish to know whether we ought to proceed as if we have in hand a sample of homogeneous normal observations.

6. Recommended Criteria for Single Samples

6.1 Let the sample of n observations be denoted in order of increasing magnitude by $x_1 \leq x_2 \leq x_3 \leq \dots \leq x_n$. Let x_n be the doubtful value, that is the largest value. The test criterion, T_n , recommended here for a single outlier is as follows:

$$T_n = (x_n - \bar{x})/s \quad (1)$$

where:

\bar{x} = arithmetic average of all n values, and

s = estimate of the population standard deviation based on the sample data, calculated as follows:

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

If x_1 rather than x_n is the doubtful value, the criterion is as follows:

$$T_1 = (\bar{x} - x_1)/s \quad (2)$$

The critical values for either case, for the 1 and 5 % levels of significance, are given in **Table 1**. **Table 1** and the following tables give the “one-sided” significance levels. In the previous tentative recommended practice (1961), the tables listed values of significance levels double those in the present practice, since it was considered that the experimenter would test either the lowest or the highest observation (or both) for statistical significance. However, to be consistent with actual practice and in an attempt to avoid further misunderstanding, single-sided significance levels are tabulated here so that both viewpoints can be represented.

6.2 The hypothesis that we are testing in every case is that all observations in the sample come from the same normal population. Let us adopt, for example, a significance level of 0.05. If we are interested *only* in outliers that occur on the *high side*, we should always use the statistic $T_n = (x_n - \bar{x})/s$ and take as critical value the 0.05 point of **Table 1**. On the other hand, if we are interested *only* in outliers occurring on the *low side*, we would always use the statistic $T_1 = (\bar{x} - x_1)/s$ and again take as a critical value the 0.05 point of **Table 1**. Suppose, however, that we are interested in outliers occurring on *either side*, but do not believe that outliers can occur on both sides simultaneously. We might, for example, believe that at some time during the experiment something possibly happened to cause an extraneous variation on the high side or on the low side, but that it was very unlikely that two or more such events could have occurred, one being an extraneous variation on the high side *and* the other an extraneous variation on the low side. With this point of view we should use the statistic $T_n = (x_n - \bar{x})/s$ or the statistic $T_1 = (\bar{x} - x_1)/s$ whichever is larger. If in this instance we use the 0.05 point of **Table 1** as our critical value, the true significance level would be twice 0.05 or 0.10. If we wish a significance level of 0.05 and not 0.10, we must in this case use as a critical value the 0.025 point of **Table 1**. Similar considerations apply to the other tests given below.

6.2.1 *Example 1*—As an illustration of the use of T_n and **Table 1**, consider the following ten observations on breaking strength (in pounds) of 0.104-in. hard-drawn copper wire: 568, 570, 570, 570, 572, 572, 572, 578, 584, 596. The doubtful observation is the high value, $x_{10} = 596$. Is the value of 596

significantly high? The mean is $\bar{x} = 575.2$ and the estimated standard deviation is $s = 8.70$. We compute

$$T_{10} = (596 - 575.2)/8.70 = 2.39 \quad (3)$$

From **Table 1**, for $n = 10$, note that a T_{10} as large as 2.39 would occur by chance with probability less than 0.05. In fact, so large a value would occur by chance not much more often

than 1 % of the time. Thus, the weight of the evidence is against the doubtful value having come from the same population as the others (assuming the population is normally distributed). Investigation of the doubtful value is therefore indicated.

TABLE 1 Critical Values for T (One-Sided Test) When Standard Deviation is Calculated from the Same Sample^A

Number of Observations, n	Upper 0.1 % Significance Level	Upper 0.5 % Significance Level	Upper 1 % Significance Level	Upper 2.5 % Significance Level	Upper 5 % Significance Level	Upper 10 % Significance Level
3	1.155	1.155	1.155	1.155	1.153	1.148
4	1.499	1.496	1.492	1.481	1.463	1.425
5	1.780	1.764	1.749	1.715	1.672	1.602
6	2.011	1.973	1.944	1.887	1.822	1.729
7	2.201	2.139	2.097	2.020	1.938	1.828
8	2.358	2.274	2.221	2.126	2.032	1.909
9	2.492	2.387	2.323	2.215	2.110	1.977
10	2.606	2.482	2.410	2.290	2.176	2.036
11	2.705	2.564	2.485	2.355	2.234	2.088
12	2.791	2.636	2.550	2.412	2.285	2.134
13	2.867	2.699	2.607	2.462	2.331	2.175
14	2.935	2.755	2.659	2.507	2.371	2.213
15	2.997	2.806	2.705	2.549	2.409	2.247
16	3.052	2.852	2.747	2.585	2.443	2.279
17	3.103	2.894	2.785	2.620	2.475	2.309
18	3.149	2.932	2.821	2.651	2.504	2.335
19	3.191	2.968	2.854	2.681	2.532	2.361
20	3.230	3.001	2.884	2.709	2.557	2.385
21	3.266	3.031	2.912	2.733	2.580	2.408
22	3.300	3.060	2.939	2.758	2.603	2.429
23	3.332	3.087	2.963	2.781	2.624	2.448
24	3.362	3.112	2.987	2.802	2.644	2.467
25	3.389	3.135	3.009	2.822	2.663	2.486
26	3.415	3.157	3.029	2.841	2.681	2.502
27	3.440	3.178	3.049	2.859	2.698	2.519
28	3.464	3.199	3.068	2.876	2.714	2.534
29	3.486	3.218	3.085	2.893	2.730	2.549
30	3.507	3.236	3.103	2.908	2.745	2.563
31	3.528	3.253	3.119	2.924	2.759	2.577
32	3.546	3.270	3.135	2.938	2.773	2.591
33	3.565	3.286	3.150	2.952	2.786	2.604
34	3.582	3.301	3.164	2.965	2.799	2.616
35	3.599	3.316	3.178	2.979	2.811	2.628
36	3.616	3.330	3.191	2.991	2.823	2.639
37	3.631	3.343	3.204	3.003	2.835	2.650
38	3.646	3.356	3.216	3.014	2.846	2.661
39	3.660	3.369	3.228	3.025	2.857	2.671
40	3.673	3.381	3.240	3.036	2.866	2.682
41	3.687	3.393	3.251	3.046	2.877	2.692
42	3.700	3.404	3.261	3.057	2.887	2.700
43	3.712	3.415	3.271	3.067	2.896	2.710
44	3.724	3.425	3.282	3.075	2.905	2.719
45	3.736	3.435	3.292	3.085	2.914	2.727
46	3.747	3.445	3.302	3.094	2.923	2.736
47	3.757	3.455	3.310	3.103	2.931	2.744
48	3.768	3.464	3.319	3.111	2.940	2.753
49	3.779	3.474	3.329	3.120	2.948	2.760
50	3.789	3.483	3.336	3.128	2.956	2.768
51	3.798	3.491	3.345	3.136	2.964	2.775
52	3.808	3.500	3.353	3.143	2.971	2.783
53	3.816	3.507	3.361	3.151	2.978	2.790
54	3.825	3.516	3.368	3.158	2.986	2.798

TABLE 1 *Continued*

Number of Observations, <i>n</i>	Upper 0.1 % Significance Level	Upper 0.5 % Significance Level	Upper 1 % Significance Level	Upper 2.5 % Significance Level	Upper 5 % Significance Level	Upper 10 % Significance Level
55	3.834	3.524	3.376	3.166	2.992	2.804
56	3.842	3.531	3.383	3.172	3.000	2.811
57	3.851	3.539	3.391	3.180	3.006	2.818
58	3.858	3.546	3.397	3.186	3.013	2.824
59	3.867	3.553	3.405	3.193	3.019	2.831
60	3.874	3.560	3.411	3.199	3.025	2.837
61	3.882	3.566	3.418	3.205	3.032	2.842
62	3.889	3.573	3.424	3.212	3.037	2.849
63	3.896	3.579	3.430	3.218	3.044	2.854
64	3.903	3.586	3.437	3.224	3.049	2.860
65	3.910	3.592	3.442	3.230	3.055	2.866
66	3.917	3.598	3.449	3.235	3.061	2.871
67	3.923	3.605	3.454	3.241	3.066	2.877
68	3.930	3.610	3.460	3.246	3.071	2.883
69	3.936	3.617	3.466	3.252	3.076	2.888
70	3.942	3.622	3.471	3.257	3.082	2.893
71	3.948	3.627	3.476	3.262	3.087	2.897
72	3.954	3.633	3.482	3.267	3.092	2.903
73	3.960	3.638	3.487	3.272	3.098	2.908
74	3.965	3.643	3.492	3.278	3.102	2.912
75	3.971	3.648	3.496	3.282	3.107	2.917
76	3.977	3.654	3.502	3.287	3.111	2.922
77	3.982	3.658	3.507	3.291	3.117	2.927
78	3.987	3.663	3.511	3.297	3.121	2.931
79	3.992	3.669	3.516	3.301	3.125	2.935
80	3.998	3.673	3.521	3.305	3.130	2.940
81	4.002	3.677	3.525	3.309	3.134	2.945
82	4.007	3.682	3.529	3.315	3.139	2.949
83	4.012	3.687	3.534	3.319	3.143	2.953
84	4.017	3.691	3.539	3.323	3.147	2.957
85	4.021	3.695	3.543	3.327	3.151	2.961
86	4.026	3.699	3.547	3.331	3.155	2.966
87	4.031	3.704	3.551	3.335	3.160	2.970
88	4.035	3.708	3.555	3.339	3.163	2.973
89	4.039	3.712	3.559	3.343	3.167	2.977
90	4.044	3.716	3.563	3.347	3.171	2.981
91	4.049	3.720	3.567	3.350	3.174	2.984
92	4.053	3.725	3.570	3.355	3.179	2.989
93	4.057	3.728	3.575	3.358	3.182	2.993
94	4.060	3.732	3.579	3.362	3.186	2.996
95	4.064	3.736	3.582	3.365	3.189	3.000
96	4.069	3.739	3.586	3.369	3.193	3.003
97	4.073	3.744	3.589	3.372	3.196	3.006
98	4.076	3.747	3.593	3.377	3.201	3.011
99	4.080	3.750	3.597	3.380	3.204	3.014
100	4.084	3.754	3.600	3.383	3.207	3.017
101	4.088	3.757	3.603	3.386	3.210	3.021
102	4.092	3.760	3.607	3.390	3.214	3.024
103	4.095	3.765	3.610	3.393	3.217	3.027
104	4.098	3.768	3.614	3.397	3.220	3.030
105	4.102	3.771	3.617	3.400	3.224	3.033
106	4.105	3.774	3.620	3.403	3.227	3.037
107	4.109	3.777	3.623	3.406	3.230	3.040
108	4.112	3.780	3.626	3.409	3.233	3.043
109	4.116	3.784	3.629	3.412	3.236	3.046
110	4.119	3.787	3.632	3.415	3.239	3.049
111	4.122	3.790	3.636	3.418	3.242	3.052
112	4.125	3.793	3.639	3.422	3.245	3.055
113	4.129	3.796	3.642	3.424	3.248	3.058
114	4.132	3.799	3.645	3.427	3.251	3.061

TABLE 1 *Continued*

Number of Observations, n	Upper 0.1 % Significance Level	Upper 0.5 % Significance Level	Upper 1 % Significance Level	Upper 2.5 % Significance Level	Upper 5 % Significance Level	Upper 10 % Significance Level
115	4.135	3.802	3.647	3.430	3.254	3.064
116	4.138	3.805	3.650	3.433	3.257	3.067
117	4.141	3.808	3.653	3.435	3.259	3.070
118	4.144	3.811	3.656	3.438	3.262	3.073
119	4.146	3.814	3.659	3.441	3.265	3.075
120	4.150	3.817	3.662	3.444	3.267	3.078
121	4.153	3.819	3.665	3.447	3.270	3.081
122	4.156	3.822	3.667	3.450	3.274	3.083
123	4.159	3.824	3.670	3.452	3.276	3.086
124	4.161	3.827	3.672	3.455	3.279	3.089
125	4.164	3.831	3.675	3.457	3.281	3.092
126	4.166	3.833	3.677	3.460	3.284	3.095
127	4.169	3.836	3.680	3.462	3.286	3.097
128	4.173	3.838	3.683	3.465	3.289	3.100
129	4.175	3.840	3.686	3.467	3.291	3.102
130	4.178	3.843	3.688	3.470	3.294	3.104
131	4.180	3.845	3.690	3.473	3.296	3.107
132	4.183	3.848	3.693	3.475	3.298	3.109
133	4.185	3.850	3.695	3.478	3.302	3.112
134	4.188	3.853	3.697	3.480	3.304	3.114
135	4.190	3.856	3.700	3.482	3.306	3.116
136	4.193	3.858	3.702	3.484	3.309	3.119
137	4.196	3.860	3.704	3.487	3.311	3.122
138	4.198	3.863	3.707	3.489	3.313	3.124
139	4.200	3.865	3.710	3.491	3.315	3.126
140	4.203	3.867	3.712	3.493	3.318	3.129
141	4.205	3.869	3.714	3.497	3.320	3.131
142	4.207	3.871	3.716	3.499	3.322	3.133
143	4.209	3.874	3.719	3.501	3.324	3.135
144	4.212	3.876	3.721	3.503	3.326	3.138
145	4.214	3.879	3.723	3.505	3.328	3.140
146	4.216	3.881	3.725	3.507	3.331	3.142
147	4.219	3.883	3.727	3.509	3.334	3.144

$$T_n = \frac{(x_n - \bar{x})/s}{\sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}}} = \sqrt{\frac{\sum_{i=1}^n x_i^2 - n \cdot \bar{x}^2}{n-1}} = \sqrt{\frac{\sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2/n}{n-1}} = T_1 = \frac{(x_n - \bar{x}_1)/s}{\sqrt{\frac{\sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2/n}{n-1}}}$$

^AValues of T are taken from Ref (2). All values have been adjusted for division by $n - 1$ instead of n in calculating s .

6.3 An alternative system, the Dixon criteria, based entirely on ratios of differences between the observations is described in the literature (1)³ and may be used in cases where it is desirable to avoid calculation of s or where quick judgment is called for. For the Dixon test, the sample criterion or statistic changes with sample size. Table 2 gives the appropriate statistic to calculate and also gives the critical values of the statistic for the 1, 5, and 10 % levels of significance.

6.3.1 Example 2—As an illustration of the use of Dixon's test, consider again the observations on breaking strength given in Example 1, and suppose that a large number of such samples had to be screened quickly for outliers and it was judged too time-consuming to compute s . Table 2 indicates use of

$$r_{11} = (x_n - x_{n-1})/(x_n - x_2) \quad (4)$$

Thus, for $n = 10$,

$$r_{11} = (x_{10} - x_9)/(x_{10} - x_2) \quad (5)$$

For the measurements of breaking strength above,

$$r_{11} = (596 - 584)/(596 - 570) = 0.462 \quad (6)$$

which is a little less than 0.477, the 5 % critical value for $n = 10$. Under the Dixon criterion, we should therefore *not* consider this observation as an outlier at the 5 % level of significance. These results illustrate how borderline cases may be accepted under one test but rejected under another. It should be remembered, however, that the T -statistic discussed above is the best one to use for the single-outlier case, and final statistical judgment should be based on it. See Ferguson (3,4).

6.3.2 Further examination of the sample observations on breaking strength of hand-drawn copper wire indicates that none of the other values need testing.

NOTE 2—With experience we may usually just look at the sample

³ The boldface numbers in parentheses refer to the list of references at the end of this practice.

TABLE 2 Dixon Criteria for Testing of Extreme Observation (Single Sample)^A

n	Criterion	Significance Level (One-Sided Test)		
		10 percent	5 percent	1 percent
3	$r_{10} = (x_2 - x_1)/(x_n - x_1)$ if smallest value is suspected; $= (x_n - x_{n-1})/(x_n - x_1)$ if largest value is suspected	0.886	0.941	0.988
4		0.679	0.765	0.889
5		0.557	0.642	0.780
6		0.482	0.560	0.698
7		0.434	0.507	0.637
8	$r_{11} = (x_2 - x_1)/(x_{n-1} - x_1)$ if smallest value is suspected; $= (x_n - x_{n-1})/(x_n - x_2)$ if largest value is suspected.	0.479	0.554	0.683
9		0.441	0.512	0.635
10		0.409	0.477	0.597
11	$r_{21} = (x_3 - x_1)/(x_{n-1} - x_1)$ if smallest value is suspected; $= (x_n - x_{n-2})/(x_n - x_2)$ if largest value is suspected.	0.517	0.576	0.679
12		0.490	0.546	0.642
13		0.467	0.521	0.615
14	$r_{22} = (x_3 - x_1)/(x_{n-2} - x_1)$ if smallest value is suspected; $= (x_n - x_{n-2})/(x_n - x_3)$ if largest value is suspected.	0.492	0.546	0.641
15		0.472	0.525	0.616
16		0.454	0.507	0.595
17		0.438	0.490	0.577
18		0.424	0.475	0.561
19		0.412	0.462	0.547
20		0.401	0.450	0.535
21		0.391	0.440	0.524
22		0.382	0.430	0.514
23		0.374	0.421	0.505
24		0.367	0.413	0.497
25		0.360	0.406	0.489
26		0.354	0.399	0.486
27		0.348	0.393	0.475
28		0.342	0.387	0.469
29		0.337	0.381	0.463
30		0.332	0.376	0.457

^A $x_1 \leq x_2 \leq \dots \leq x_n$. (See Ref (1), Appendix.)

values to observe if an outlier is present. However, strictly speaking the statistical test should be applied to all samples to guarantee the significance levels used. Concerning “multiple” tests on a single sample, we comment on this below.

6.4 A test equivalent to T_n (or T_1) based on the sample sum of squared deviations from the mean for all the observations and the sum of squared deviations omitting the “outlier” is given by Grubbs (5).

6.5 The next type of problem to consider is the case where we have the possibility of two outlying observations, the least and the greatest observation in a sample. (The problem of testing the two highest or the two lowest observations is considered below.) In testing the least and the greatest observations simultaneously as probable outliers in a sample, we use the ratio of sample range to sample standard deviation test of David, Hartley, and Pearson (6). The significance levels for this sample criterion are given in Table 3. Alternatively, the largest residuals test of Tietjen and Moore (7) could be used. An example in astronomy follows.

6.5.1 *Example 3*—There is one rather famous set of observations that a number of writers on the subject of outlying observations have referred to in applying their various tests for “outliers.” This classic set consists of a sample of 15 observations of the vertical semidiameters of Venus made by Lieutenant Herndon in 1846 (8). In the reduction of the observations, Prof. Pierce assumed two unknown quantities and found the following residuals which have been arranged in ascending order of magnitude:

-1.40 in.	-0.24	-0.05	0.18	0.48
-0.44	-0.22	0.06	0.20	0.63
-0.30	-0.13	0.10	0.39	1.01

The deviations – 1.40 and 1.01 appear to be outliers. Here the suspected observations lie at each end of the sample. Much less work has been accomplished for the case of outliers at both ends of the sample than for the case of one or more outliers at only one end of the sample. This is not necessarily because the “one-sided” case occurs more frequently in practice but because “two-sided” tests are much more difficult to deal with. For a high and a low outlier in a single sample, we give two procedures below, the first being a combination of tests, and the second a single test of Tietjen and Moore (7) which may have nearly optimum properties. For optimum procedures when there is an independent estimate at hand, s^2 or σ^2 , see (9).

6.6 For the observations on the semi-diameter of Venus given above, all the information on the measurement error is contained in the sample of 15 residuals. In cases like this, where no independent estimate of variance is available (that is, we still have the single sample case), a useful statistic is the ratio of the range of the observations to the sample standard deviation:

$$w/s = (x_n - x_1)/s \quad (7)$$

where:

$$s = \sqrt{\sum[(x_i - \bar{x})^2/(n - 1)]} \quad (8)$$

If x_n is about as far above the mean, \bar{x} , as x_1 is below \bar{x} , and if w/s exceeds some chosen critical value, then one would conclude that *both* the doubtful values are outliers. If, however, x_1 and x_n are displaced from the mean by different amounts, some further test would have to be made to decide whether to reject as outlying only the lowest value or only the highest value or both the lowest and highest values.