

# Understanding Z-Scores and P-Values: A Step-by-Step Guide to Manual Calculation

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## Introduction to Z-Scores and P-Values in Statistical Testing

The core of modern [inferential statistical procedures](#) relies heavily on the accurate calculation and interpretation of two fundamental metrics: the [Z-score](#) and the [P-value](#). While professional data analysts and researchers typically leverage specialized statistical software or digital calculators to find the P-value corresponding to a calculated Z-score, understanding the foundational manual process remains critical for academic rigor and conceptual mastery.

The [Z-score](#), often referred to as the standard score, serves as a measure of how many **standard deviations** a specific observation or data point is displaced from the mean of its distribution. By assuming a [Standard Normal Distribution](#), the Z-score enables us to standardize disparate results, facilitating rigorous and objective comparisons across various datasets and studies. This standardization is essential for drawing reliable conclusions in data analysis.

Conversely, the P-value represents the probability of observing data results that are as extreme as, or even more extreme than, the data observed in the study, provided that the [null hypothesis](#) is true. In the realm of [hypothesis testing](#), the P-value is the definitive measure used to determine **statistical significance**, guiding the decision whether to reject or fail to reject the null hypothesis.

Although automated tools provide immediate results, there are times--such as during educational assessments or when resources are constrained--when one must rely on manual calculation. Calculating a [P-value](#) from a [Z-score](#) by hand requires precise reference to a comprehensive standardized [Z-table](#). The following detailed examples illustrate this fundamental process for the three primary types of statistical tests.

## Deconstructing the Standard Normal Z-Table

The [Z-table](#) acts as the essential standardized reference tool for manual Z-score calculations. It systematically provides the **cumulative probability** associated with any given Z-score. This probability corresponds precisely to the area under the [Standard Normal Distribution](#) curve, spanning from the extreme left tail up to the specific Z-score value you are examining.

Mastering the structure of the Z-table is key to accurate manual computation. Typically, the far left column lists the Z-score value up to its first decimal place (e.g., 1.4 or -0.8). The top row then details the second decimal place (e.g., 0.03 or 0.04). By locating the intersection of the corresponding row and column, you can retrieve the exact probability value, which represents the cumulative area beneath the curve.

It is crucial to verify the convention used by the specific table you are referencing. Most commonly, Z-tables report the area to the left (the cumulative probability), which simplifies calculations for left-tailed tests. However, some tables report the area between the mean ( $Z=0$ ) and the specific Z-

score. For the purposes of the examples detailed below, we will assume the standard cumulative probability format, representing the area to the left of the score.

### Example 1: Determining the P-Value for a Left-Tailed Test

A left-tailed test is implemented when the alternative hypothesis posits that the true population parameter mean is significantly **less than** the hypothesized mean value. Since the standard [Z-table](#) conventionally measures the area to the left of a given score, the calculation for a left-tailed test is inherently the most straightforward approach among the three possibilities.

Let us assume we are conducting a left-tailed [hypothesis test](#) and have calculated a negative Z-score of **-1.22**. Our objective is to determine the exact [P-value](#) that corresponds to this observed test statistic. This P-value will represent the probability of observing a result this far into the negative tail.

To find the required probability, we must locate the value **-1.22** within the negative side of the Z-table. We look specifically for the row designated -1.2 and track across until we reach the column labeled 0.02. The intersection point provides the cumulative area:

z	0	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
-3.0	0.0013	0.0013	0.0013	0.0012	0.0012	0.0011	0.0011	0.0011	0.0010	0.0010
-2.9	0.0019	0.0018	0.0018	0.0017	0.0016	0.0016	0.0015	0.0015	0.0014	0.0014
-2.8	0.0026	0.0025	0.0024	0.0023	0.0023	0.0022	0.0021	0.0021	0.0020	0.0019
-2.7	0.0035	0.0034	0.0033	0.0032	0.0031	0.0030	0.0029	0.0028	0.0027	0.0026
-2.6	0.0047	0.0045	0.0044	0.0043	0.0041	0.0040	0.0039	0.0038	0.0037	0.0036
-2.5	0.0062	0.0060	0.0059	0.0057	0.0055	0.0054	0.0052	0.0051	0.0049	0.0048
-2.4	0.0082	0.0080	0.0078	0.0075	0.0073	0.0071	0.0069	0.0068	0.0066	0.0064
-2.3	0.0107	0.0104	0.0102	0.0099	0.0096	0.0094	0.0091	0.0089	0.0087	0.0084
-2.2	0.0139	0.0136	0.0132	0.0129	0.0125	0.0122	0.0119	0.0116	0.0113	0.0110
-2.1	0.0179	0.0174	0.0170	0.0166	0.0162	0.0158	0.0154	0.0150	0.0146	0.0143
-2.0	0.0228	0.0222	0.0217	0.0212	0.0207	0.0202	0.0197	0.0192	0.0188	0.0183
-1.9	0.0287	0.0281	0.0274	0.0268	0.0262	0.0256	0.0250	0.0244	0.0239	0.0233
-1.8	0.0359	0.0351	0.0344	0.0336	0.0329	0.0322	0.0314	0.0307	0.0301	0.0294
-1.7	0.0446	0.0436	0.0427	0.0418	0.0409	0.0401	0.0392	0.0384	0.0375	0.0367
-1.6	0.0548	0.0537	0.0526	0.0516	0.0505	0.0495	0.0485	0.0475	0.0465	0.0455
-1.5	0.0668	0.0655	0.0643	0.0630	0.0618	0.0606	0.0594	0.0582	0.0571	0.0559
-1.4	0.0808	0.0793	0.0778	0.0764	0.0749	0.0735	0.0721	0.0708	0.0694	0.0681
-1.3	0.0968	0.0951	0.0934	0.0918	0.0901	0.0885	0.0869	0.0853	0.0838	0.0823
-1.2	0.1151	0.1131	0.1112	0.1093	0.1075	0.1056	0.1038	0.1020	0.1003	0.0985
-1.1	0.1357	0.1335	0.1314	0.1292	0.1271	0.1251	0.1230	0.1210	0.1190	0.1170
-1.0	0.1587	0.1562	0.1539	0.1515	0.1492	0.1469	0.1446	0.1423	0.1401	0.1379
-0.9	0.1841	0.1814	0.1788	0.1762	0.1736	0.1711	0.1685	0.1660	0.1635	0.1611
-0.8	0.2119	0.2090	0.2061	0.2033	0.2005	0.1977	0.1949	0.1922	0.1894	0.1867
-0.7	0.2420	0.2389	0.2358	0.2327	0.2296	0.2266	0.2236	0.2206	0.2177	0.2148
-0.6	0.2743	0.2709	0.2676	0.2643	0.2611	0.2578	0.2546	0.2514	0.2483	0.2451

The numerical value retrieved directly from the table represents the probability mass that falls into the critical region defined by the area in the far left tail. For a Z-score of -1.22, the table yields **0.1112**. Because this is a left-tailed test, this value is immediately our final P-value, quantifying the probability of observing a result equally or more extreme than our finding.

## Example 2: Calculating the P-Value for a Right-Tailed Test

A right-tailed test is applied when the alternative hypothesis suggests that the true population parameter is significantly **greater than** the hypothesized value. This scenario requires an extra step in the manual calculation. Since standard Z-tables report the cumulative area to the left of the score, we must subtract this cumulative probability from 1 (representing the total area under the curve) to isolate the area located in the desired right tail.

Consider a right-tailed [hypothesis test](#) where we calculate a positive Z-score of **1.43**. We must meticulously determine the corresponding [P-value](#) that reflects the probability of results occurring at or beyond this positive Z-score.

The initial step is to locate the value **1.43** within the positive side of the Z-table. We find the intersection of the row labeled 1.4 and the column labeled 0.03:

z	0	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857

The value obtained directly from the table, 0.9236, signifies the total area **to the left** of  $Z = 1.43$ . Given that we are conducting a right-tailed test, our critical region lies on the opposite side. To find the area of the right tail, we apply the rule of complements, subtracting the cumulative probability from 1.

The essential calculation for determining our final P-value is:  $1 - 0.9236 = \mathbf{0.0764}$ . This resulting P-value represents the probability of observing a Z-score of 1.43 or greater, assuming the hypothesized conditions (the null hypothesis) are accurate.

### Example 3: Finding the P-Value for a Two-Tailed Test

The two-tailed test is necessary when the alternative hypothesis specifies that the true parameter is simply **different from** the hypothesized value, without predicting a specific direction (i.e., the parameter is "not equal" to the mean). This framework requires us to account for both extremely low and extremely high outcomes symmetrically across the distribution.

In a two-tailed test, the critical region is divided equally between both tails of the [Standard Normal Distribution](#). Consequently, the calculation requires finding the probability associated with the calculated Z-score in one tail and then **doubling that result** to account for the symmetrical probability in the opposite tail.

Suppose we perform a two-tailed [hypothesis test](#) and achieve a Z-score of **-0.84**. We must calculate the corresponding [P-value](#) that accurately reflects the probability of observing a result this extreme in either direction.

We begin by locating the absolute value of the Z-score, **-0.84**, within the negative side of the Z-table. We find the intersection of the row -0.8 and the column 0.04:

z	0	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
-3.0	0.0013	0.0013	0.0013	0.0012	0.0012	0.0011	0.0011	0.0011	0.0010	0.0010
-2.9	0.0019	0.0018	0.0018	0.0017	0.0016	0.0016	0.0015	0.0015	0.0014	0.0014
-2.8	0.0026	0.0025	0.0024	0.0023	0.0023	0.0022	0.0021	0.0021	0.0020	0.0019
-2.7	0.0035	0.0034	0.0033	0.0032	0.0031	0.0030	0.0029	0.0028	0.0027	0.0026
-2.6	0.0047	0.0045	0.0044	0.0043	0.0041	0.0040	0.0039	0.0038	0.0037	0.0036
-2.5	0.0062	0.0060	0.0059	0.0057	0.0055	0.0054	0.0052	0.0051	0.0049	0.0048
-2.4	0.0082	0.0080	0.0078	0.0075	0.0073	0.0071	0.0069	0.0068	0.0066	0.0064
-2.3	0.0107	0.0104	0.0102	0.0099	0.0096	0.0094	0.0091	0.0089	0.0087	0.0084
-2.2	0.0139	0.0136	0.0132	0.0129	0.0125	0.0122	0.0119	0.0116	0.0113	0.0110
-2.1	0.0179	0.0174	0.0170	0.0166	0.0162	0.0158	0.0154	0.0150	0.0146	0.0143
-2.0	0.0228	0.0222	0.0217	0.0212	0.0207	0.0202	0.0197	0.0192	0.0188	0.0183
-1.9	0.0287	0.0281	0.0274	0.0268	0.0262	0.0256	0.0250	0.0244	0.0239	0.0233
-1.8	0.0359	0.0351	0.0344	0.0336	0.0329	0.0322	0.0314	0.0307	0.0301	0.0294
-1.7	0.0446	0.0436	0.0427	0.0418	0.0409	0.0401	0.0392	0.0384	0.0375	0.0367
-1.6	0.0548	0.0537	0.0526	0.0516	0.0505	0.0495	0.0485	0.0475	0.0465	0.0455
-1.5	0.0668	0.0655	0.0643	0.0630	0.0618	0.0606	0.0594	0.0582	0.0571	0.0559
-1.4	0.0808	0.0793	0.0778	0.0764	0.0749	0.0735	0.0721	0.0708	0.0694	0.0681
-1.3	0.0968	0.0951	0.0934	0.0918	0.0901	0.0885	0.0869	0.0853	0.0838	0.0823
-1.2	0.1151	0.1131	0.1112	0.1093	0.1075	0.1056	0.1038	0.1020	0.1003	0.0985
-1.1	0.1357	0.1335	0.1314	0.1292	0.1271	0.1251	0.1230	0.1210	0.1190	0.1170
-1.0	0.1587	0.1562	0.1539	0.1515	0.1492	0.1469	0.1446	0.1423	0.1401	0.1379
-0.9	0.1841	0.1814	0.1788	0.1762	0.1736	0.1711	0.1685	0.1660	0.1635	0.1611
-0.8	0.2119	0.2090	0.2061	0.2033	0.2005	0.1977	0.1949	0.1922	0.1894	0.1867
-0.7	0.2420	0.2389	0.2358	0.2327	0.2296	0.2266	0.2236	0.2206	0.2177	0.2148
-0.6	0.2743	0.2709	0.2676	0.2643	0.2611	0.2578	0.2546	0.2514	0.2483	0.2451

The probability found in the table is 0.2005. Since this is a two-tailed test, this value represents only the area in one tail (the area to the left of  $Z = -0.84$ ). To capture the total probability of being equally extreme--either less than  $-0.84$  or greater than  $+0.84$ --we must multiply this single-tail probability by 2. The final calculation for our P-value is:  $0.2005 * 2 = \mathbf{0.401}$ .

## Summary and Interpretation of Results

Performing the manual calculation of the P-value from the Z-score is an excellent way to reinforce understanding of fundamental statistical concepts, specifically **probability distribution** and the definition of critical regions. Once the P-value has been determined using the methods outlined above, the final step in hypothesis testing is comparing it against the predefined level of **significance**, typically denoted as alpha ( $\alpha$ ), often set at 0.05).

The interpretation rule is simple yet critical: If the calculated P-value is less than or equal to the significance level ( $\text{P-value} \leq \alpha$ ), the result is considered statistically significant, and we subsequently reject the null hypothesis. Conversely, if the P-value exceeds alpha, we fail to reject the null hypothesis, concluding there is insufficient evidence to support the alternative claim.

To ensure clarity and accuracy when calculating the P-value by hand, consistently follow these key steps:

Identify the precise type of test being conducted (i.e., left-tailed, right-tailed, or two-tailed).

Locate the calculated Z-score in the [Z-table](#) to retrieve the cumulative area (let's call this Area A).

If the test is **Left-Tailed**: P-value = Area A (the value read directly from the table).

If the test is **Right-Tailed**: P-value =  $1 - \text{Area A}$  (subtracting the cumulative area from 1).

If the test is **Two-Tailed**: P-value =  $2 * (\text{Tail Area})$ . Ensure you use the value representing the tail probability, which is either Area A (if Z is negative) or  $(1 - \text{Area A})$  (if Z is positive).

## Transitioning to Automated Tools for Efficiency

While manual calculation provides foundational conceptual understanding and is invaluable for educational purposes, modern statistical practice mandates the use of automated tools for maximizing speed and guaranteeing precision. This is particularly true when researchers are dealing with complex distributions, large datasets, or Z-scores that require interpolation beyond the precise values listed in a static Z-table.

Statistical software and programming languages automate the integration process beneath the [Normal Distribution](#) curve, providing immediate and highly accurate P-values without the reliance on lookup tables or manual subtraction/multiplication.

For those transitioning from manual methods to automated workflows, the following tutorials explore how to calculate P-values efficiently using specialized statistical platforms and functions:

Calculating P-values in **R programming** environments using built-in statistical functions.

Utilizing the NORMSDIST or NORM.S.DIST functions within **Microsoft Excel** or similar spreadsheet software to determine standard normal probabilities.

Implementing P-value calculation in **Python** using the robust SciPy statistical library functions.