

Statistics Review

for the De Veaux/Velleman/Bock Series

Data

Categorical data are information about characteristics or qualities falling into different classifications.

Quantitative data are information about a quantity or measurement (with units).

The **W's** describe the data's context—*Who, What, Why, When, Where, and hoW.*

Displaying Categorical Data

A **bar chart** displays the distribution of a categorical variable, showing the count or percentage of values in each category.

A **pie chart** displays the distribution of a categorical variable by slicing a circle into pieces whose areas are proportional to the fraction of values in each category.

A **contingency table** displays the distribution of observations categorized on two variables.

Describing Categorical Data

A **distribution** shows counts or percentages of observations in each category.

A **conditional distribution** shows the distribution of one variable within a single category of another variable.

Independence exists when the distribution of one variable is the same in all categories of another variable; if the distribution depends on the category, we say there's an **association**.

Displaying Quantitative Data

A **histogram** displays the distribution of a quantitative variable in bars showing counts or percentages of observations falling in each interval.

A **stem-and-leaf display** records the actual data values falling in each interval by splitting the data into a stem (the tens digit, say) and a leaf (the units digit).

A **dotplot** displays dots (instead of bars or digits) for data values in each interval.

A **boxplot** displays a box spanning the middle 50% of the data (extending from the first to third quartile and showing the median), with whiskers extending to the lowest and highest nonoutlier data values, and outliers plotted.

Describing Quantitative Data

Always mention **shape** (*modes symmetry*), **center**, **spread**, and **unusual features** (*gaps, clusters, and outliers*).

Quantitative Data Statistics

Minimum is the smallest data value.

Maximum is the largest data value.

Range: $range = maximum - minimum$

The **median** is the middle value (half the data are larger, half smaller).

The **quartiles** divide each half of the data in half; 25% of the data are smaller than Q_1 (the first quartile) and 25% larger than Q_3 .

Interquartile range: $IQR = Q_3 - Q_1$

Outlier guideline: Data values that lie more than 1.5 IQRs below Q_1 or above Q_3 may be outliers.

5-number summary: $min, Q_1, median, Q_3, max$

Mean: $\bar{y} = \frac{\sum y}{n}$

Standard deviation: $s = \sqrt{\frac{\sum (y - \bar{y})^2}{n - 1}}$

Two Quantitative Variables

A **scatterplot** displays points corresponding to cases measured on two variables.

The **direction** is *positive* if higher values of one variable are generally associated with higher values of the other and *negative* if higher values of one variable are generally associated with lower values of the other.

Form: *linear* or *curved*

Strength: The less scatter, the stronger the association.

Unusual features: Look for *clusters, outliers, and influential points*.

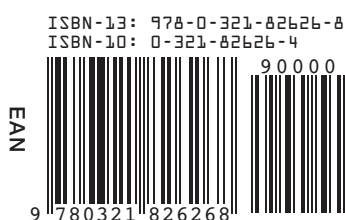
Correlation: r is a number between -1 and $+1$ describing the direction and strength of a linear relationship between two quantitative variables.

$$r = \frac{\sum z_x z_y}{n - 1}$$

Straight Enough Condition: If the pattern in the scatterplot looks reasonably straight, it's okay to fit a linear model.

The **regression line** is a model that predicts a value of y for each x ; $\hat{y} = b_0 + b_1x$, where $b_1 = \frac{r s_y}{s_x}$ and $b_0 = \bar{y} - b_1\bar{x}$; the line passes through (\bar{x}, \bar{y}) .

[more>](#)



Two Quantitative Variables (continued)

Residual: $e = y - \hat{y}$, the difference between the actual value of y and the value predicted by the model.

Least squares: The regression line minimizes the sum of the squared residuals.

Slope: The slope models the relationship as y -units per x -unit.

Intercept: The y -intercept is the starting value (the value of \hat{y} predicted when $x = 0$).

R-squared: R^2 is the fraction of the variability in y explained by the regression model.

Modeling Wisdom

Residual plots appear as randomly scattered points when the model is appropriate.

Influential points distort the model; if you are suspicious, try creating regression models both with and without them.

Cause and effect: A strong association is *not* evidence of causation.

Subsets: If a scatterplot shows distinct groups, it may be better to fit a model to each one separately.

Curvature: If the relationship is curved, **re-express** one or both variables to straighten the relationship. Possible approaches

include the **Ladder of Powers** (re-express y as $y^2, \sqrt{y}, \log y, \frac{-1}{\sqrt{y}}, \frac{-1}{y}$, etc.) or use $\log y$ and $\log x$.

Simulations

Steps in a simulation: To investigate the distribution of outcomes for a situation of interest, create a simulation model based on random numbers.

1. **Model a component:** Explain how you will interpret random numbers to represent the most basic event of interest.
2. **Simulate a trial:** Explain how you will use random numbers to model one outcome.
3. Define your **response variable**.
4. **Run many trials**, recording the outcome for each.
5. **Analyze the response variable** by graphing the data and calculating summary statistics.
6. **State a conclusion** in the context of the original question.

Sampling

A **sample** is a subset of a **population** for which data are collected and analyzed in an effort to learn about unknown (unknowable) properties of the population. We use **sample statistics** to estimate **population parameters**.

Property	Statistic	Parameter
Proportion	\hat{p}	p
Mean	\bar{y}	μ
Standard deviation	s	σ
Slope	b_1	β_1

[more>](#)

Sampling (continued)

Sampling error is sample-to-sample variation in a statistic.

Bias is found in sampling methods that systematically misrepresent characteristics of the population.

- **Undercoverage** limits (or omits) some subpopulation.
- **Voluntary response** allows individuals to self-select their participation.
- **Nonresponse bias** occurs when many of those sampled elect not to participate.
- **Response bias** influences people's answers.

Random sampling gives each member of the population the same chance of being selected.

- In a **simple random sample**, each subset of size n is equally likely to be selected.
- A **stratified sample** draws random samples from each of several homogeneous subpopulations (strata).
- A **cluster sample** randomly selects entire heterogeneous subpopulations (clusters) from among many.
- A **systematic sample** selects (for example) every 12th individual from a list of the population starting from a randomly determined case.

Observational Studies

A **retrospective** study collects information looking into the past; a **prospective** study follows subjects over time.

Observational studies can spot associations between variables but can neither reach conclusions about populations nor establish cause and effect. A **lurking variable** that influences both x and y can make it appear that x causes y .

Experiments

An experiment applies treatments to randomly assigned subjects to observe the response.

A **factor** is a variable manipulated by the experimenter, applied at several different **levels**.

A **treatment** is the combination of factor levels applied to a subject.

The **response variable** is the (usually quantitative) outcome we measure to compare effects of the treatments.

A **control group** receives no treatment (or a null treatment) to provide a baseline for purposes of comparison.

Principles of Design:

- **Control** known sources of variability whenever possible.
- **Randomize** subjects to treatments to balance unknown sources of variability (subjects needn't be a random sample).
- **Replicate** each treatment on many subjects.
- **Block** subjects with respect to preexisting sources of variability we can't control.

Blinding is keeping people involved with the experiment unaware of treatment assignments, both (1) during the experiment (subjects and others in contact with them, often accomplished with **placebos**) and (2) during evaluation of the response. An experiment is **double blind** when both classes are kept unaware.

A **confounding variable** is a variable that influences the response variable in ways that we can't separate from the effects of the experimental factor.

Probability

Probability is the long-run frequency of an event's occurrence; $0 \leq P(A) \leq 1$.

A **sample space** is the set of all possible outcomes; $P(S) = 1$.

Complement Rule: $P(A^c) = 1 - P(A)$

Addition Rule: $P(A \text{ OR } B) = P(A) + P(B) - P(A \text{ AND } B)$

Multiplication Rule: $P(A \text{ AND } B) = P(A) \times P(B|A)$

Conditional probability: $P(B|A) = \frac{P(A \text{ AND } B)}{P(A)}$

Disjoint (mutually exclusive) events cannot both happen: $P(A \text{ AND } B) = 0$

Independent events: The occurrence of one event has no impact on the probability of the other: $P(A|B) = P(A)$

For **random variables:**

$$\mu = E(X) = \sum (x \times P(x))$$

$$\sigma^2 = \text{Var}(X) = \sum ((x - \mu)^2 P(x))$$

$$E(X + c) = E(X) + c$$

$$\text{Var}(X + c) = \text{Var}(X)$$

$$E(aX) = aE(X)$$

$$\text{Var}(aX) = a^2 \text{Var}(X)$$

$$E(X \pm Y) = E(X) \pm E(Y)$$

Pythagorean Theorem of Statistics:

If random variables X and Y are independent, then

$$\text{Var}(X \pm Y) = \text{Var}(X) + \text{Var}(Y)$$

$$SD(X \pm Y) = \sqrt{SD^2(X) + SD^2(Y)}$$

Normal model:

A Normal model, $N(\mu, \sigma)$, is unimodal, symmetric, and bell-shaped and is specified by its mean, μ , and standard deviation, σ .

68% of values lie within $\mu \pm 1\sigma$;

95% of values lie within $\mu \pm 2\sigma$;

99.7% lie within $\mu \pm 3\sigma$.

Bernoulli trials:

- two outcomes (success, failure)
- known probability of success p
- trials are independent

Geometric model:

X = number of Bernoulli trials until the first success

$$P(x) = q^{x-1}p$$

$$E(X) = \frac{1}{p}$$

Binomial model:

X = number of successes in n Bernoulli trials

$$P(x) = \binom{n}{x} p^x q^{n-x}$$

$$E(X) = np \quad SD(X) = \sqrt{npq}$$

Normal approximation: If we expect at least 10 successes and 10 failures, then binomial probabilities may be approximated using the Normal model $N(np, \sqrt{npq})$.

Sampling Distribution Models

For a **sample proportion:** Provided that the sampled values are independent and the sample size is large enough, the sampling distribution of \hat{p} can be modeled by a Normal model with

$$\mu(\hat{p}) = p \text{ and } SD(\hat{p}) = \sqrt{\frac{pq}{n}}$$

For a **sample mean:** The **Central Limit Theorem** If a random sample of size n is drawn from a population with mean μ and standard deviation σ , then as n increases the sampling distribution of the sample mean, \bar{y} , approaches the Normal model $N(\mu, \frac{\sigma}{\sqrt{n}})$, regardless of the shape of the population.

Confidence Intervals

If the appropriate assumptions and conditions are met, we can have a specified level of confidence that the interval

$$estimate \pm (critical\ value) \times SE(estimate)$$

captures the value of a population parameter.

Hypothesis Tests

The four steps:

Hypotheses: Write a null hypothesis for the value of the population parameter and specify the alternative hypothesis (upper tail, lower tail, or two-tailed).

Model: Check assumptions and conditions, then specify the type of test and the sampling model.

Mechanics: Calculate the test statistic and find the P-value.

Conclusion: Link the P-value to your decision (reject or fail to reject H_0) and state your conclusion in the proper context.

The **null hypothesis** (H_0) specifies a parameter and a hypothesized value for that parameter.

The **alternative hypothesis** (H_A) is a statement indicating what values of the parameter are of interest (different from, smaller than, or larger than that specified in H_0).

The **P-value** is the probability that results at least as extreme as those we observed could have occurred if the null hypothesis were true.

Type I error is rejecting the null hypothesis when it is true.

Type II error is failing to reject the null hypothesis when it is false.

Power is the probability the test rejects a false null hypothesis.

Effect size is the difference between the hypothesized value of the parameter and its true value.

Assumptions for Inference

PROPORTIONS (z)

- **One sample**
 1. Individuals are independent.
 2. Sample is sufficiently large.
- **Two groups**
 1. Groups are independent.
 2. Data in each group are independent.
 3. Both groups are sufficiently large.

MEANS (t)

- **One Sample** ($df = n - 1$)
 1. Individuals are independent.
 2. Population has a Normal model.
- **Matched pairs** ($df = n - 1$)
 1. Data are matched.
 2. Individuals are independent.
 3. Population of differences is Normal.
- **Two independent samples** (df from technology)
 1. Groups are independent.
 2. Data in each group are independent.
 3. Both populations are Normal.

And the Conditions That Support or Override Them

1. SRS and $n < 10\%$ of the population.
 2. Successes and failures each ≥ 10 .
1. (Think about how the data were collected.)
 2. Both are SRSs and $n < 10\%$ of populations OR random allocation.
 3. Successes and failures each ≥ 10 for both groups.
1. (Think about the design.)
 2. SRS and $n < 10\%$ of the population.
 3. Histogram is unimodal and symmetric.*
1. (Think about the design.)
 2. SRSs and $n < 10\%$ OR random allocation.
 3. Histogram of differences is unimodal and symmetric.*
1. (Think about the design.)
 2. SRSs and $n < 10\%$ OR random allocation.
 3. Both histograms are unimodal and symmetric.*

DISTRIBUTIONS/ASSOCIATION (χ^2)

- **Goodness-of-fit** ($df = \#$ of cells $- 1$; one variable, one sample compared with population model)
 1. Data are counts.
 2. Data in sample are independent.
 3. Sample is sufficiently large.
- **Homogeneity** [$df = (r - 1)(c - 1)$; many groups compared on one variable]
 1. Data are counts.
 2. Data in groups are independent.
 3. Groups are sufficiently large.
- **Independence** [$df = (r - 1)(c - 1)$; sample from one population classified on two variables]
 1. Data are counts.
 2. Data are independent.
 3. Sample is sufficiently large.

REGRESSION (t, $df = n - 2$)

- **Association** of each quantitative variable ($\beta = 0?$)
 1. Form of relationship is linear.
 2. Errors are independent.
 3. Variability of errors is constant.
 4. Errors have a Normal model.
1. Scatterplot looks approximately linear.
 2. No apparent pattern in residuals plot.
 3. Residuals plot has consistent spread.
 4. Histogram of residuals is approximately unimodal and symmetric, or Normal probability plot reasonably straight.*
- (*less critical as n increases)

Quick Guide to Inference

Think		Show					
Inference about?	One group or two?	Procedure	Model	Parameter	Estimate	SE	
PROPORTIONS	One sample	1-Proportion z-Interval	z	p	\hat{p}	$\sqrt{\frac{\hat{p}\hat{q}}{n}}$	
		1-Proportion z-Test				$\sqrt{\frac{p_0q_0}{n}}$	
	Two independent groups	2-Proportion z-Interval	z	$p_1 - p_2$	$\hat{p}_1 - \hat{p}_2$	$\sqrt{\frac{\hat{p}_1\hat{q}_1}{n_1} + \frac{\hat{p}_2\hat{q}_2}{n_2}}$	
		2-Proportion z-Test				$\sqrt{\frac{\hat{p}\hat{q}}{n_1} + \frac{\hat{p}\hat{q}}{n_2}}, \hat{p} = \frac{y_1 + y_2}{n_1 + n_2}$	
MEANS	One sample	t-Interval t-Test	t $df = n - 1$	μ	\bar{y}	$\frac{s}{\sqrt{n}}$	
	Two independent groups	2-Sample t-Test 2-Sample t-Interval	t df from technology	$\mu_1 - \mu_2$	$\bar{y}_1 - \bar{y}_2$	$\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$	
	Matched pairs	Paired t-Test Paired t-Interval	t $df = n - 1$	μ_d	\bar{d}	$\frac{s_d}{\sqrt{n}}$	
DISTRIBUTIONS (one categorical variable)	One sample	Goodness-of-Fit	χ^2 $df = \text{cells} - 1$				
	Many independent groups	Homogeneity χ^2 Test	χ^2 $df = (r - 1)(c - 1)$			$\sum \frac{(Obs - Exp)^2}{Exp}$	
INDEPENDENCE (two categorical variables)	One sample	Independence χ^2 Test	χ^2 $df = (r - 1)(c - 1)$				
	One sample	Linear Regression t-Test or Confidence Interval for β		β_1	b_1	$\frac{s_e}{s_x \sqrt{n - 1}}$ (compute with technology)	
ASSOCIATION (two quantitative variables)	One sample	*Confidence Interval for μ_v	t $df = n - 2$	μ_v	\hat{y}_v	$\sqrt{SE^2(b_1) \times (x_v - \bar{x})^2 + \frac{s_e^2}{n}}$	
		*Prediction Interval for y_v		y_v	\hat{y}_v	$\sqrt{SE^2(b_1) \times (x_v - \bar{x})^2 + \frac{s_e^2}{n} + s_e^2}$	
Inference about?	One group or two?	Procedure	Model	Parameter	Estimate	SE	Chapter

Table Z
Areas under the standard normal curve

z	0.09	0.08	0.07	0.06	0.05	0.04	0.03	0.02	0.01	0.00	z
-3.8	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	-3.8
-3.7	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	-3.7
-3.6	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0002	0.0002	-3.6
-3.5	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	-3.5
-3.4	0.0002	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	-3.4
-3.3	0.0003	0.0004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0005	0.0005	0.0005	-3.3
-3.2	0.0005	0.0005	0.0005	0.0006	0.0006	0.0006	0.0006	0.0006	0.0007	0.0007	-3.2
-3.1	0.0007	0.0007	0.0008	0.0008	0.0008	0.0008	0.0009	0.0009	0.0009	0.0010	-3.1
-3.0	0.0010	0.0010	0.0011	0.0011	0.0011	0.0012	0.0012	0.0013	0.0013	0.0013	-3.0
-2.9	0.0014	0.0014	0.0015	0.0015	0.0016	0.0016	0.0017	0.0018	0.0018	0.0019	-2.9
-2.8	0.0019	0.0020	0.0021	0.0021	0.0022	0.0023	0.0023	0.0024	0.0025	0.0026	-2.8
-2.7	0.0026	0.0027	0.0028	0.0029	0.0030	0.0031	0.0032	0.0033	0.0034	0.0035	-2.7
-2.6	0.0036	0.0037	0.0038	0.0039	0.0040	0.0041	0.0043	0.0044	0.0045	0.0047	-2.6
-2.5	0.0048	0.0049	0.0051	0.0052	0.0054	0.0055	0.0057	0.0059	0.0060	0.0062	-2.5
-2.4	0.0064	0.0066	0.0068	0.0069	0.0071	0.0073	0.0075	0.0078	0.0080	0.0082	-2.4
-2.3	0.0084	0.0087	0.0089	0.0091	0.0094	0.0096	0.0099	0.0102	0.0104	0.0107	-2.3
-2.2	0.0110	0.0113	0.0116	0.0119	0.0122	0.0125	0.0129	0.0132	0.0136	0.0139	-2.2
-2.1	0.0143	0.0146	0.0150	0.0154	0.0158	0.0162	0.0166	0.0170	0.0174	0.0179	-2.1
-2.0	0.0183	0.0188	0.0192	0.0197	0.0202	0.0207	0.0212	0.0217	0.0222	0.0228	-2.0
-1.9	0.0233	0.0239	0.0244	0.0250	0.0256	0.0262	0.0268	0.0274	0.0281	0.0287	-1.9
-1.8	0.0294	0.0301	0.0307	0.0314	0.0322	0.0329	0.0336	0.0344	0.0351	0.0359	-1.8
-1.7	0.0367	0.0375	0.0384	0.0392	0.0401	0.0409	0.0418	0.0427	0.0436	0.0446	-1.7
-1.6	0.0455	0.0465	0.0475	0.0485	0.0495	0.0505	0.0516	0.0526	0.0537	0.0548	-1.6
-1.5	0.0559	0.0571	0.0582	0.0594	0.0606	0.0618	0.0630	0.0643	0.0655	0.0668	-1.5
-1.4	0.0681	0.0694	0.0708	0.0721	0.0735	0.0749	0.0764	0.0778	0.0793	0.0808	-1.4
-1.3	0.0823	0.0838	0.0853	0.0869	0.0885	0.0901	0.0918	0.0934	0.0951	0.0968	-1.3
-1.2	0.0985	0.1003	0.1020	0.1038	0.1056	0.1075	0.1093	0.1112	0.1131	0.1151	-1.2
-1.1	0.1170	0.1190	0.1210	0.1230	0.1251	0.1271	0.1292	0.1314	0.1335	0.1357	-1.1
-1.0	0.1379	0.1401	0.1423	0.1446	0.1469	0.1492	0.1515	0.1539	0.1562	0.1587	-1.0
-0.9	0.1611	0.1635	0.1660	0.1685	0.1711	0.1736	0.1762	0.1788	0.1814	0.1841	-0.9
-0.8	0.1867	0.1894	0.1922	0.1949	0.1977	0.2005	0.2033	0.2061	0.2090	0.2119	-0.8
-0.7	0.2148	0.2177	0.2206	0.2236	0.2266	0.2296	0.2327	0.2358	0.2389	0.2420	-0.7
-0.6	0.2451	0.2483	0.2514	0.2546	0.2578	0.2611	0.2643	0.2676	0.2709	0.2743	-0.6
-0.5	0.2776	0.2810	0.2843	0.2877	0.2912	0.2946	0.2981	0.3015	0.3050	0.3085	-0.5
-0.4	0.3121	0.3156	0.3192	0.3228	0.3264	0.3300	0.3336	0.3372	0.3409	0.3446	-0.4
-0.3	0.3483	0.3520	0.3557	0.3594	0.3632	0.3669	0.3707	0.3745	0.3783	0.3821	-0.3
-0.2	0.3859	0.3897	0.3936	0.3974	0.4013	0.4052	0.4090	0.4129	0.4168	0.4207	-0.2
-0.1	0.4247	0.4286	0.4325	0.4364	0.4404	0.4443	0.4483	0.4522	0.4562	0.4602	-0.1
0.0	0.4641	0.4681	0.4721	0.4761	0.4801	0.4840	0.4880	0.4920	0.4960	0.5000	0.0

For $z \leq -3.90$, the areas are 0.0000 to four decimal places.

Table Z (cont.)
Areas under the standard normal curve

z	0.00	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
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981