

Constructing Statistical Tolerance Limits for Non-Normal Data

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Statistical Tolerance Limits

- Consider a sample of n observations taken from a continuous population.

$$\{X_1, X_2, \dots, X_n\}$$

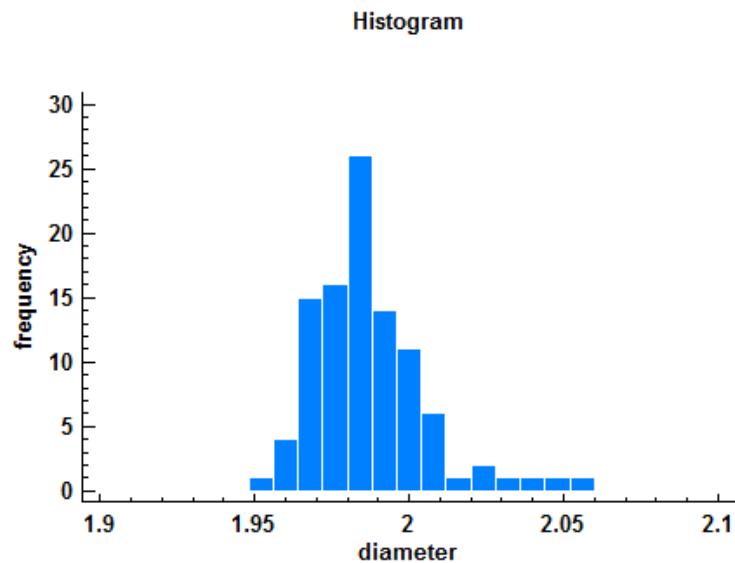
- Statistical tolerance limits create an interval that bounds a specified percentage of the population at a given level of confidence.

(such as 99% of the population with 95% confidence)

- These intervals are often used to demonstrate compliance with a set of requirements or specification limits.

Example: Medical Devices

- Consider the following measurements of the diameter of a sample $n = 100$ of medical devices:



- The specification limits are 2.0 ± 0.1

Assuming Normality

If we could assume that the data are a random sample from a normal distribution, then a 95% statistical tolerance interval for 99% of the underlying population would be given by:

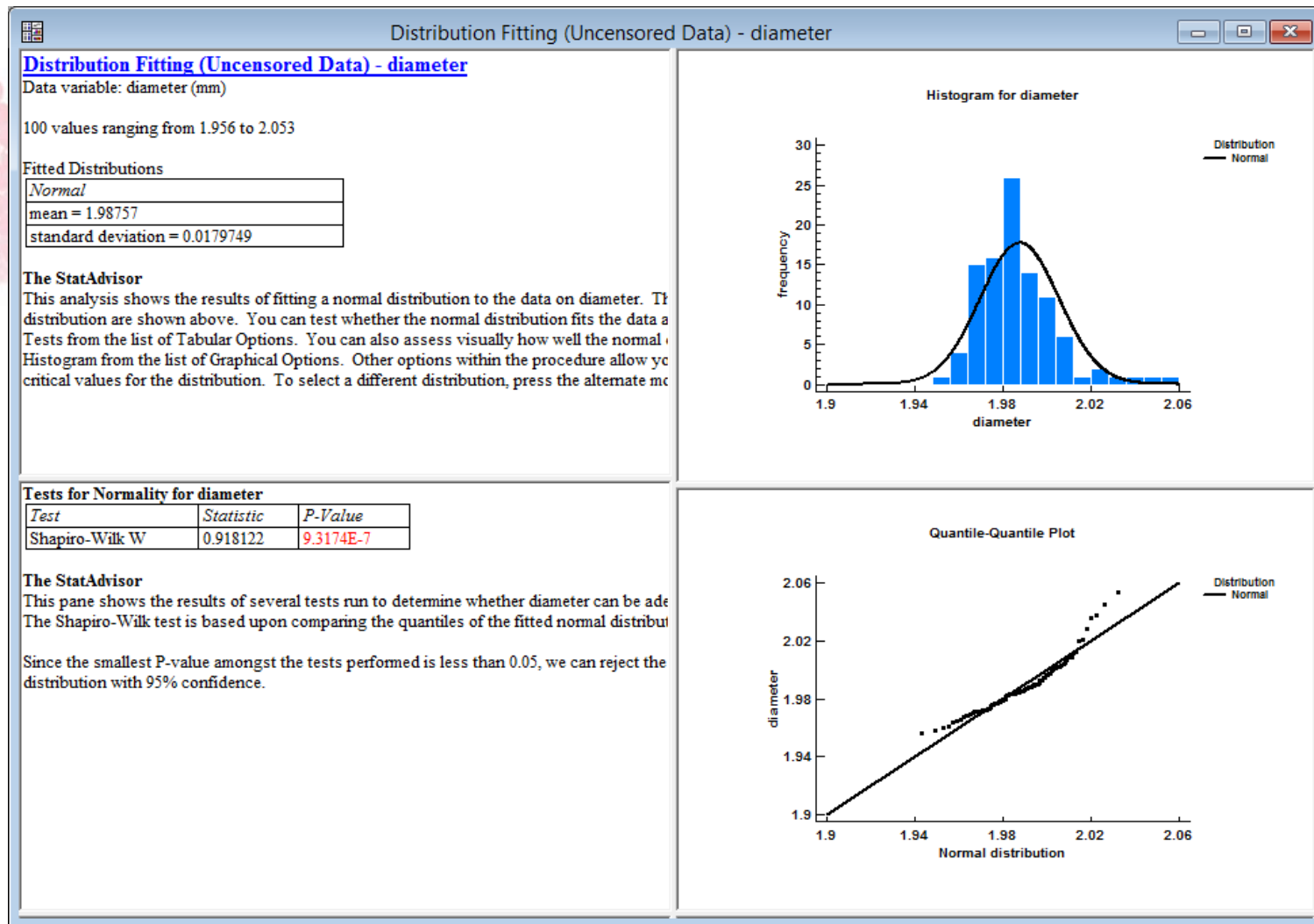
$$\bar{x} \pm Ks$$

K depends on the level of confidence $100(1-\alpha)\%$ and the percentage of the population P to be bound.

General Approach for Constructing Statistical Tolerance Limits

- Step 1: Test data for normality.
 - If normal dist. is tenable, calculate normal tolerance limits.
- Step 2: If not normal, search for a normalizing transformation.
 - If acceptable transformation is found, calculate normal tolerance limits for transformed data and invert the limits.
- Step 3: If transformation approach fails, try alternative distributions such as lognormal, extreme value or Weibull.
 - If a good fit is found, calculate tolerance limits using that distribution.
- Step 4: If all else fails, calculate nonparametric tolerance limits.

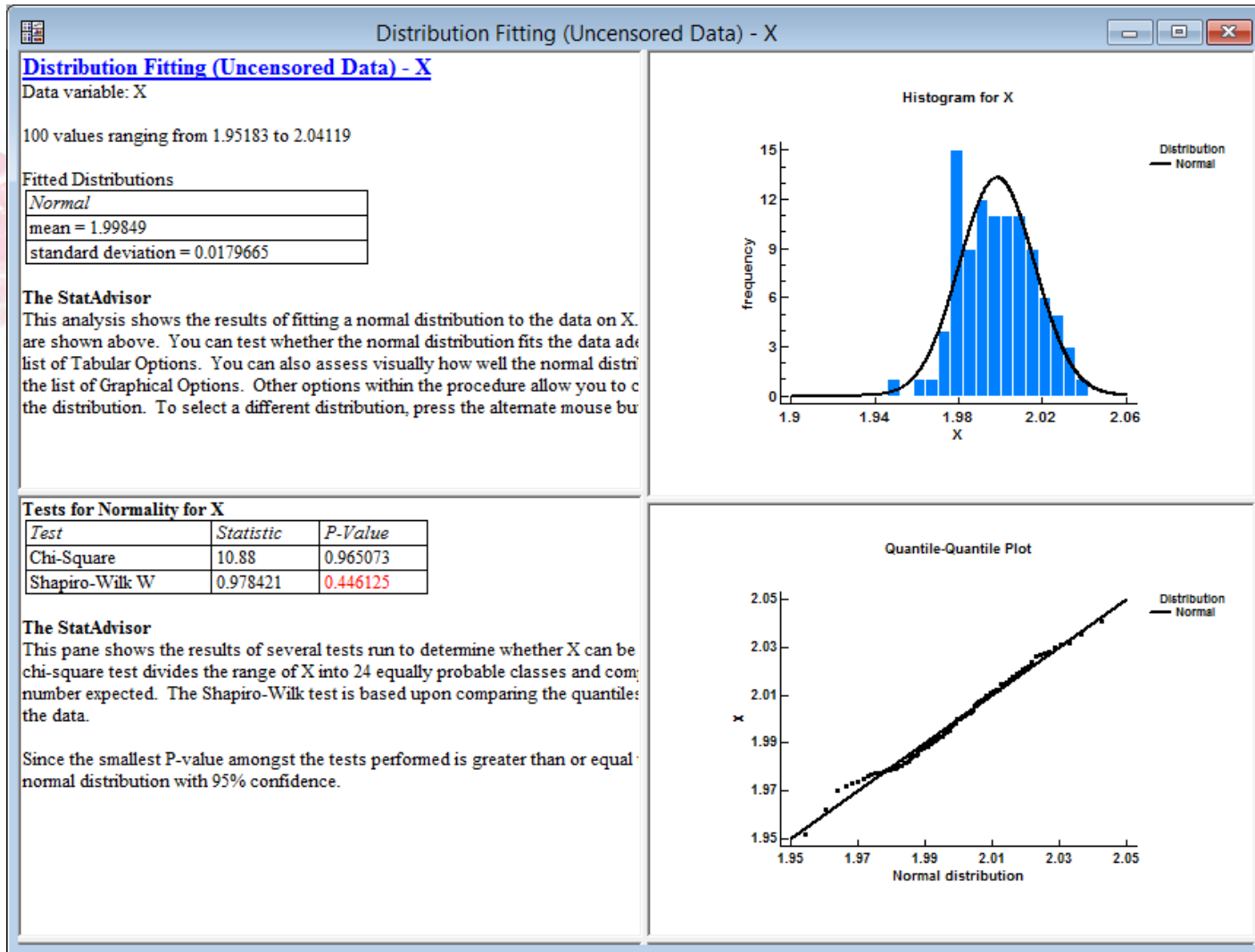
Step 1: Test for Normality



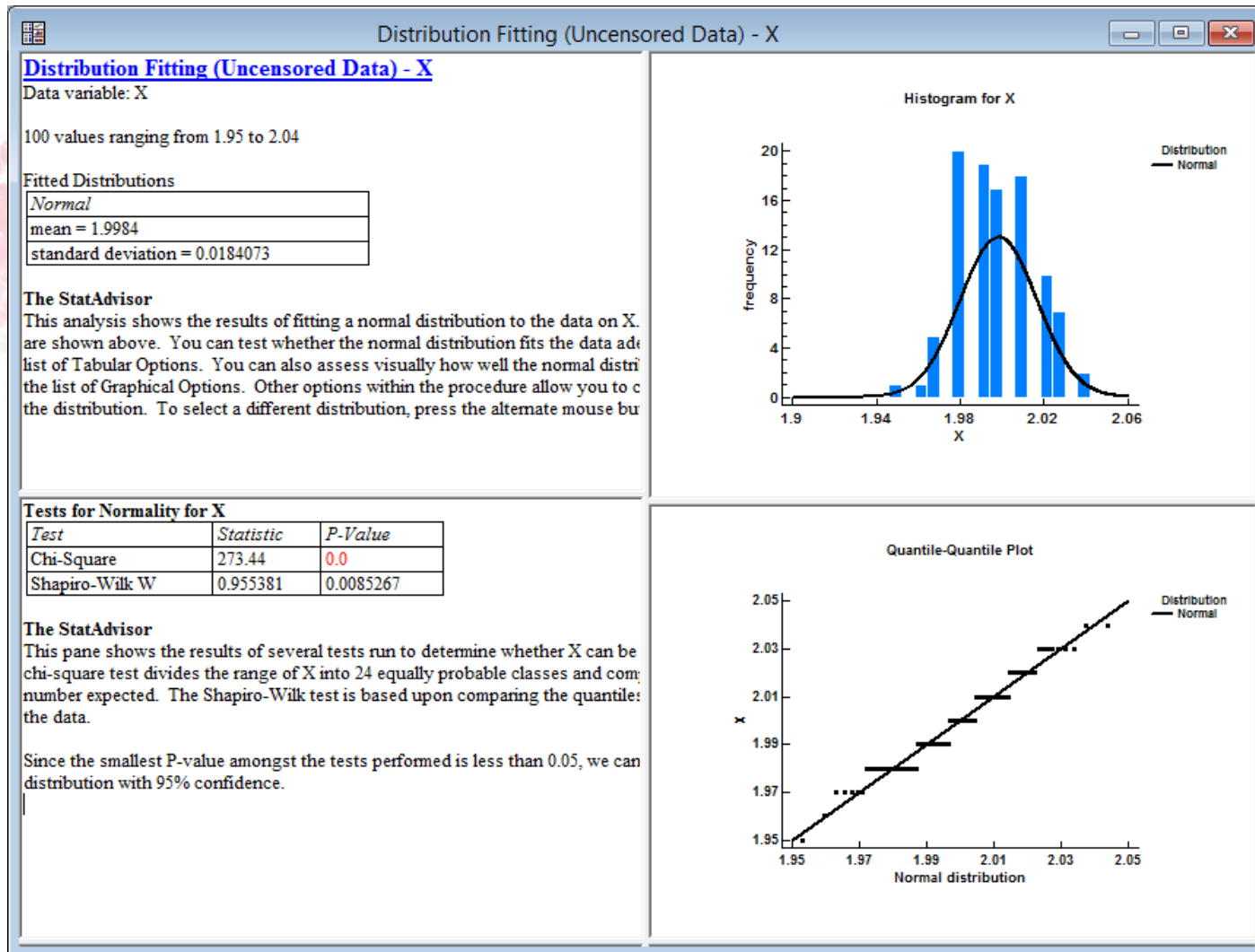
Tests for Normality

- Shapiro-Wilk test: recommended for sample sizes between 3 and 5000.
- Anderson-Darling test: recommended for sample sizes $n > 5000$.
- Chi-square test: recommended for heavily rounded data. NOTE: be sure to set up classes that match the rounding of the data.

Warning: Beware of rounded data



Warning: Beware of rounded data



Solution: bin the data

- Be sure classes match the rounding

Frequency Tabulation Options

Number of Classes:

Lower Limit:

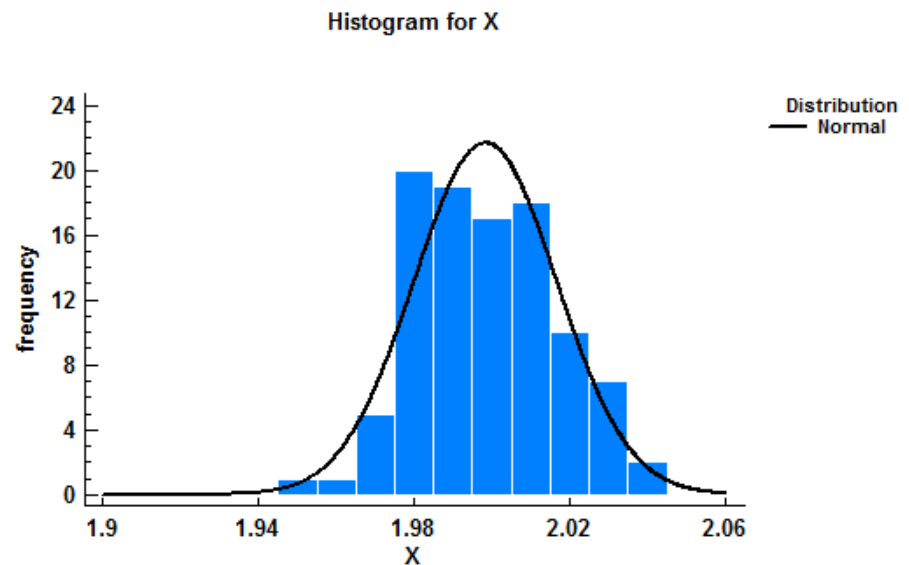
Upper Limit: Hold

Display

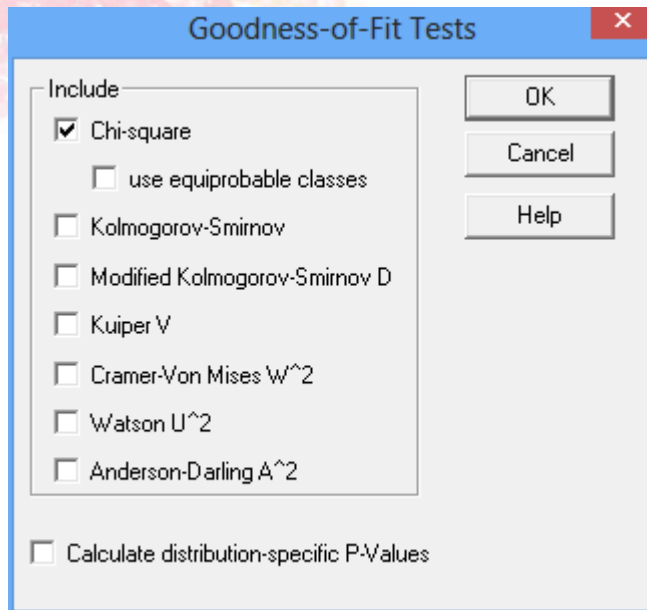
Histogram

Rootogram

Suspended rootogram



Run chi-square test



Goodness-of-Fit Tests for X
Chi-Square Test

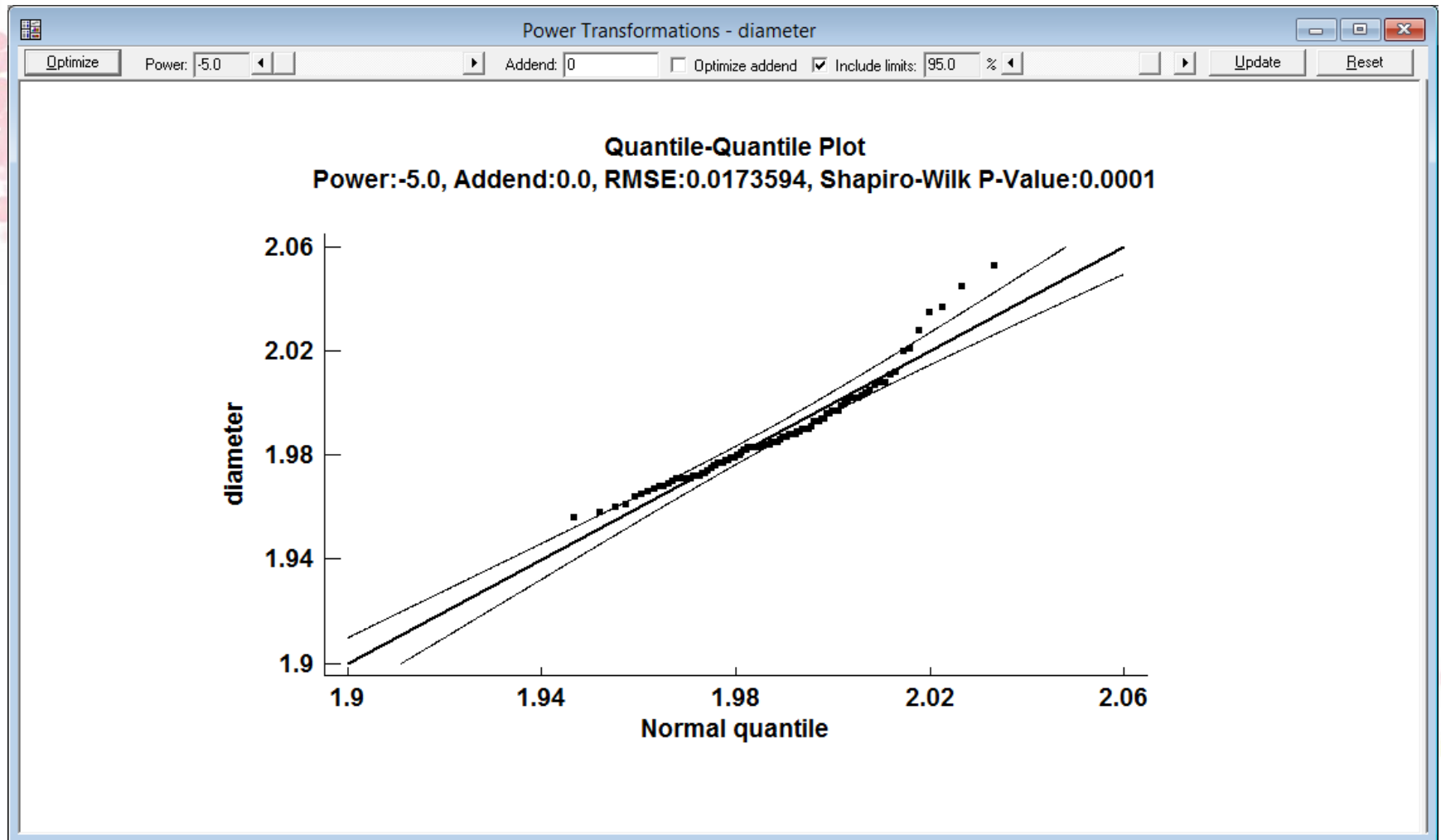
	<i>Lower</i>	<i>Upper</i>	<i>Observed</i>	<i>Expected</i>	
	<i>Limit</i>	<i>Limit</i>	<i>Frequency</i>	<i>Frequency</i>	<i>Chi-Square</i>
at or below		1.965	2	3.48	0.63
	1.965	1.975	5	6.70	0.43
	1.975	1.985	20	13.15	3.57
	1.985	1.995	19	19.34	0.01
	1.995	2.005	17	21.33	0.88
	2.005	2.015	18	17.64	0.01
	2.015	2.025	10	10.94	0.08
	2.025	2.035	7	5.08	0.72
above	2.035		2	2.34	0.05

Chi-Square = 6.37593 with 6 d.f. P-Value = 0.382421

Step 2: Search for normalizing transformation

- If X does not follow a normal distribution, it may be possible to find a power p such that X^p is normally distributed.
- If so, normal tolerance limits may be constructed for X^p and then inverted to create limits for X .
- The general method of Box and Cox may be used to find the best value of p .

Statlets - Power Transformations

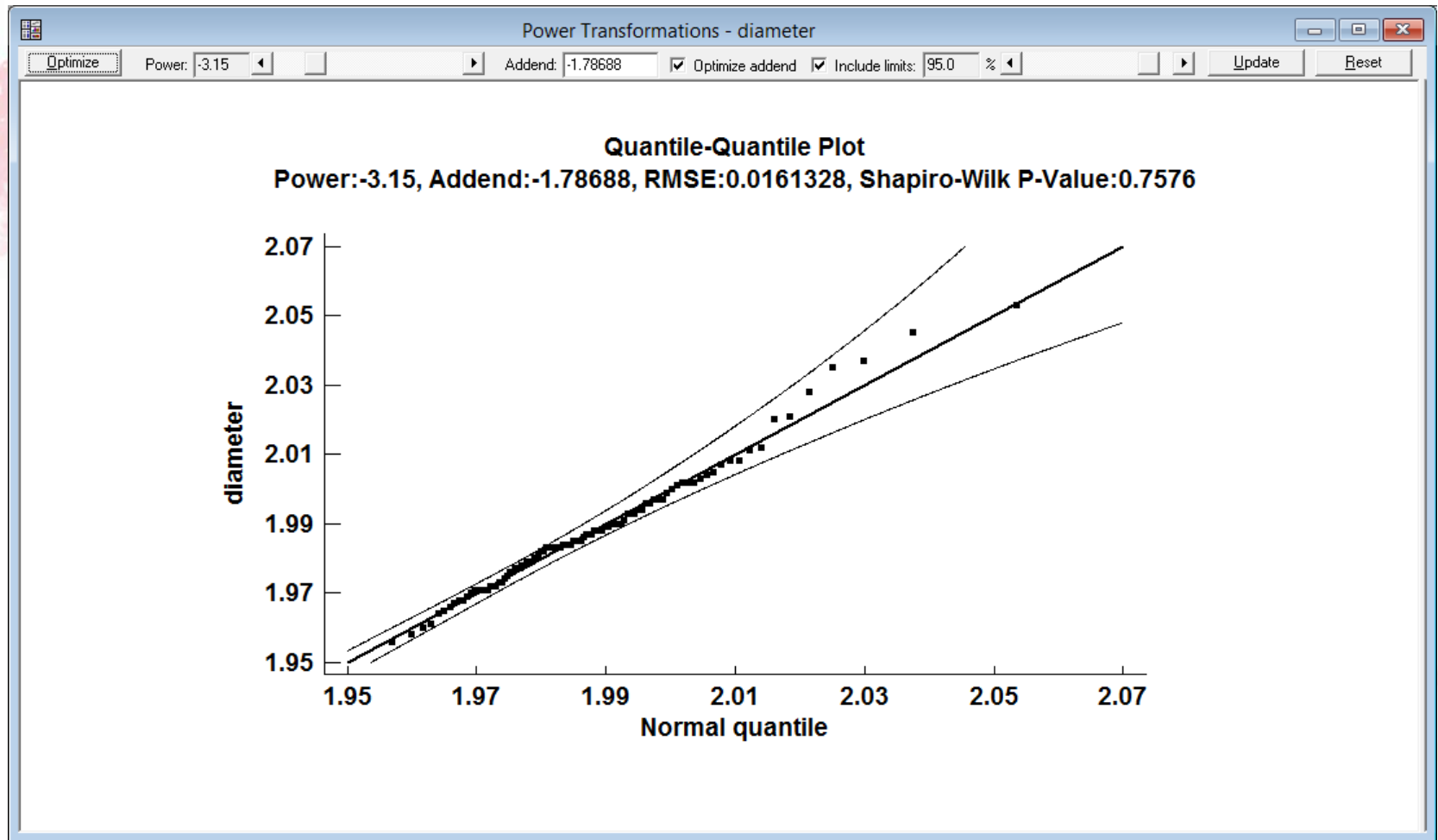


Power plus Addend

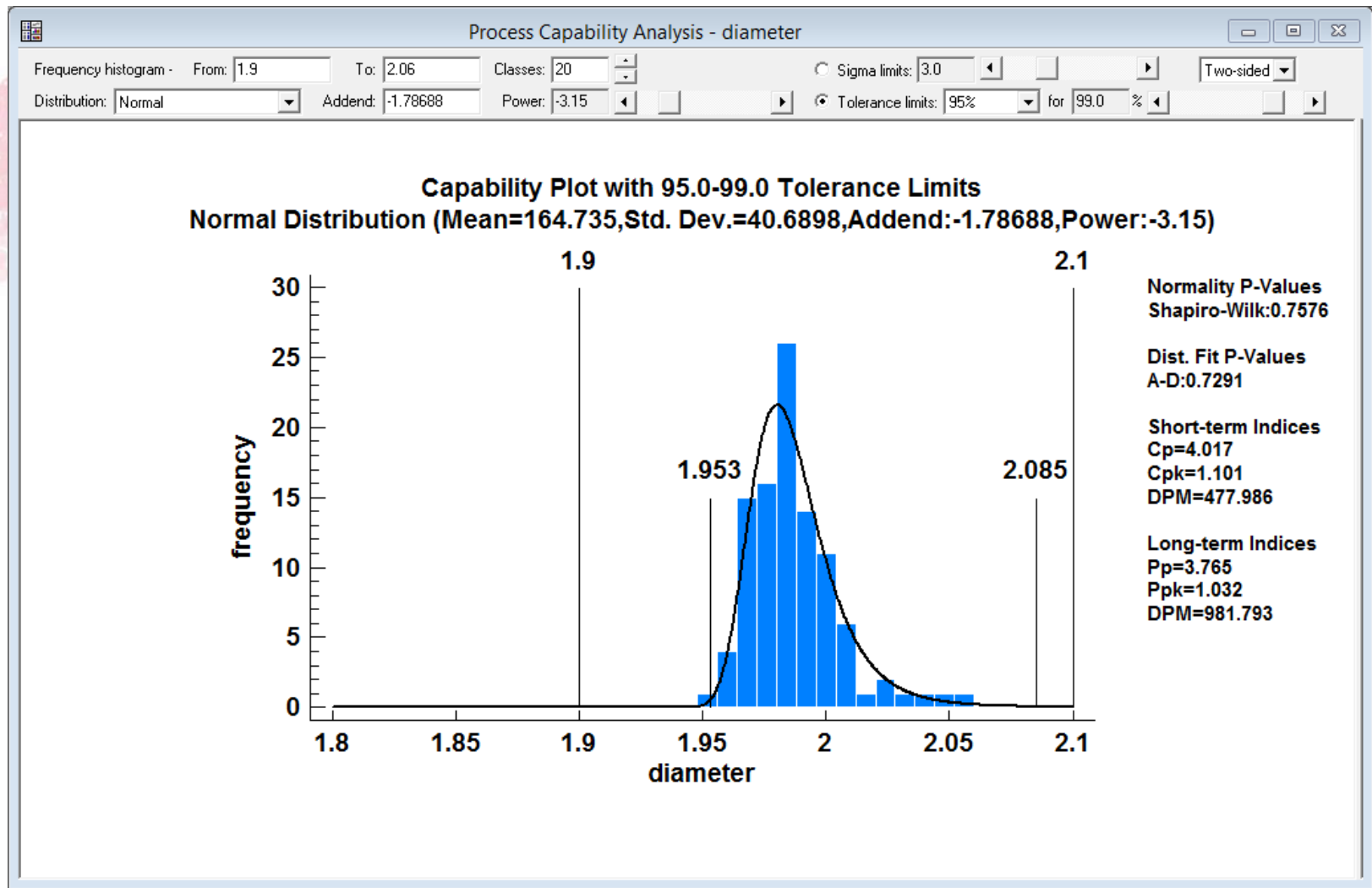
- If a simple power transformation does not work, an addend can be estimated as well:

$$X^t = (X + \Delta)^p$$

Power plus Addend



Tolerance Limits



Step 3: Find an Alternative Distribution

- Statgraphics will calculate statistical tolerance limits for the following distributions:
 - Normal
 - Lognormal
 - Weibull
 - Cauchy
 - Exponential
 - 2-parameter exponential
 - Gamma
 - Laplace
 - Largest extreme value
 - Pareto
 - Smallest extreme value

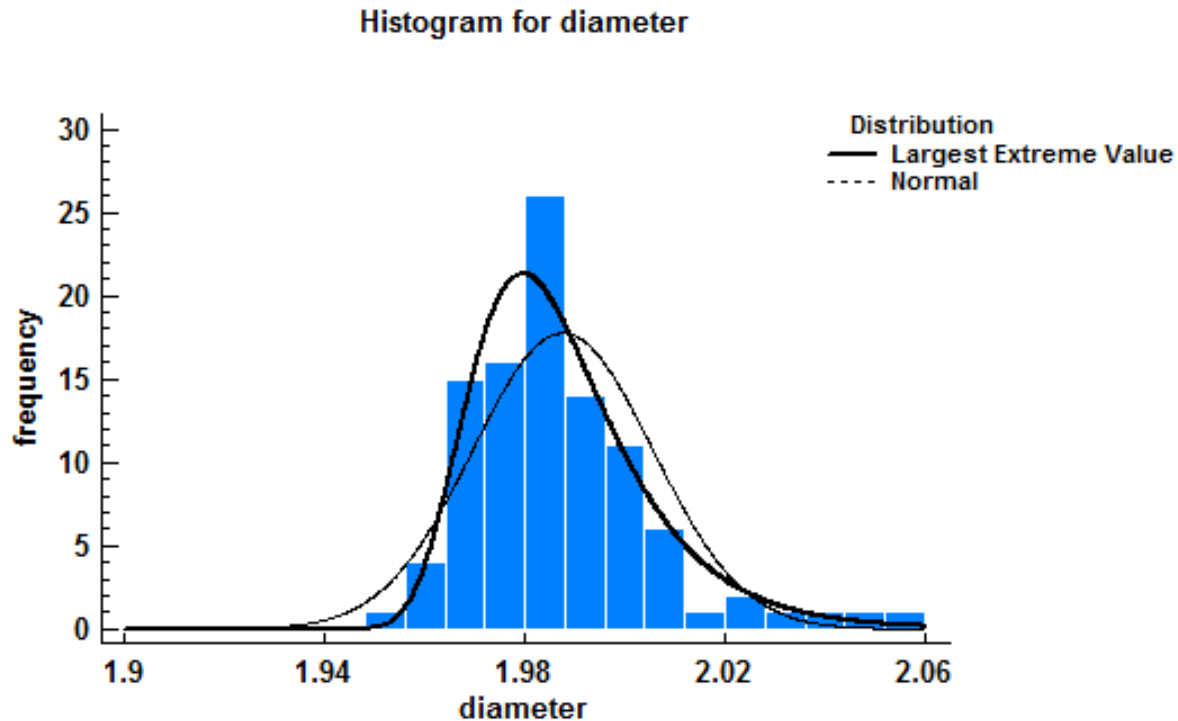
Distribution Fitting

- Comparison of alternative distributions

Comparison of Alternative Distributions

<i>Distribution</i>	<i>Est. Parameters</i>	<i>Log Likelihood</i>	<i>A²</i>
Largest Extreme Value	2	270.696	0.309344
Laplace	2	266.126	0.925591
Loglogistic	2	265.245	0.844609
Logistic	2	264.939	0.881328
Inverse Gaussian	2	261.022	1.83356
Lognormal	2	261.02	1.85076
Gamma	2	260.846	1.86193
Normal	2	260.484	1.93765
Cauchy	2	257.124	1.37669
Smallest Extreme Value	2	234.823	6.71728
Weibull	2	161.746	
Pareto	1	-131.127	44.7816
Exponential	1	-168.691	45.1176

Fitted Distributions



Goodness-of-Fit

- To test goodness-of-fit for alternative distributions, Anderson-Darling test is popular.
- Be sure to use “modified” form of the test, which adjusts the test statistics and P-value to account for estimated parameters.

Goodness-of-Fit Tests for diameter
Anderson-Darling A²

	<i>Largest Extreme Value</i>	<i>Normal</i>
A ²	0.309344	1.93765
Modified Form	0.315531	1.95262
P-Value	$\geq 0.10^*$	0.0000564085*

*Indicates that the P-Value has been compared to tables of critical values specially constructed for fitting the selected distribution. Other P-values are based on general tables and may be very conservative (except for the Chi-Square Test).

Tolerance Limits

Statistical Tolerance Limits Options

Distribution

Normal

Lognormal

Weibull

Normal after transformation

Power:

Nonparametric (specified confidence)

Interval Depth:

Nonparametric (specified proportion)

Cauchy

Exponential

Exponential (2-parameter)

Gamma

Laplace

Largest extreme value

Pareto

Smallest extreme value

Type of Limits

Two-sided

Lower limit only

Upper limit only

Confidence Level: %

Population Proportion: %

Threshold:

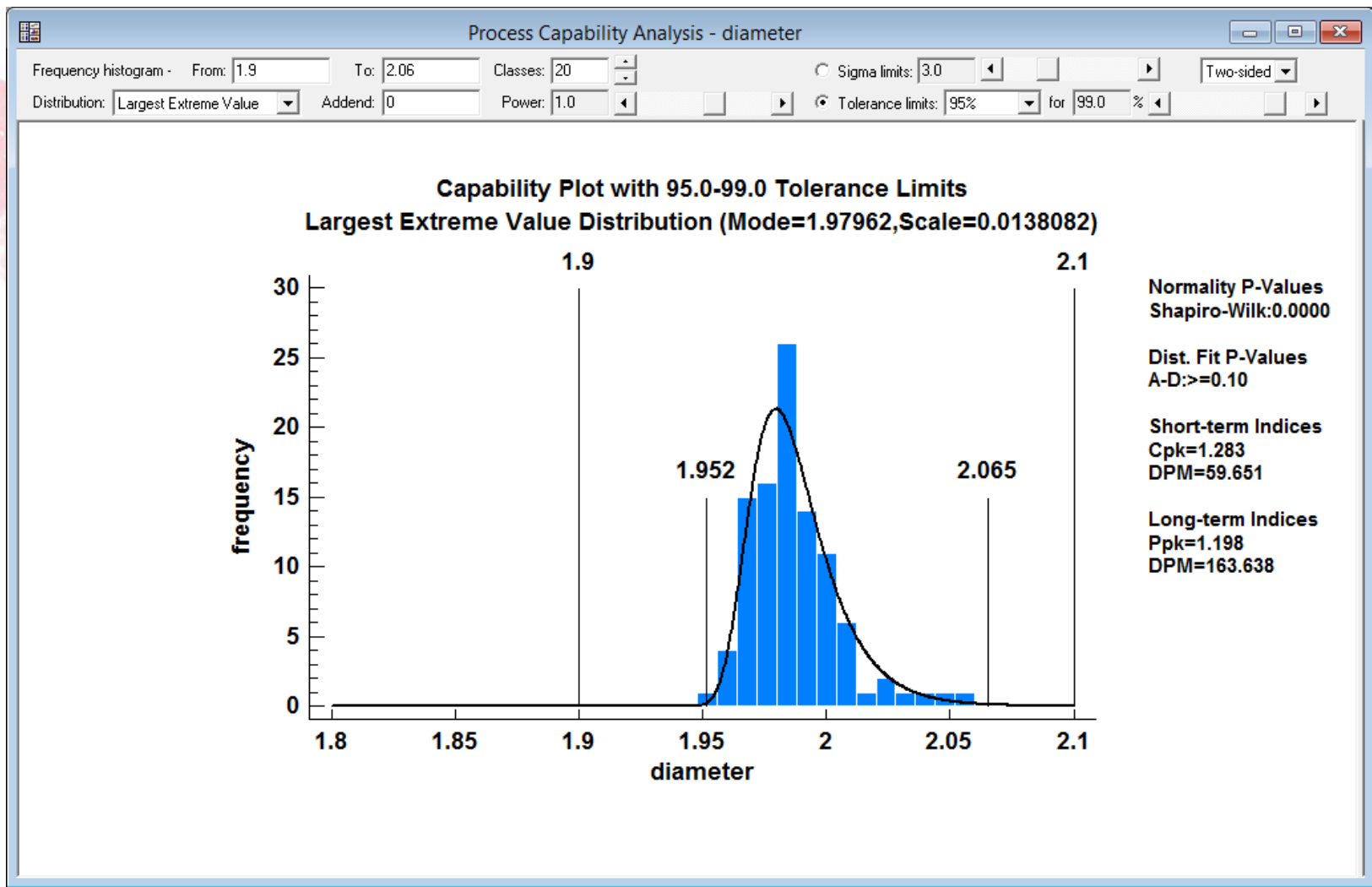
Pareto Threshold:

OK

Cancel

Help

Tolerance Limits



Step 4: Nonparametric Limits

- Can specify either the population percentage or the confidence level, but not both.

Statistical Tolerance Limits Options

Distribution

- Normal
- Lognormal
- Weibull
- Normal after transformation
- Power:
- Nonparametric (specified confidence)
- Interval Depth:
- Nonparametric (specified proportion)

- Cauchy
- Exponential
- Exponential (2-parameter)
- Gamma
- Laplace
- Largest extreme value
- Pareto
- Smallest extreme value

Type of Limits

- Two-sided
- Lower limit only
- Upper limit only

Confidence Level: %

Population Proportion: %

Threshold:

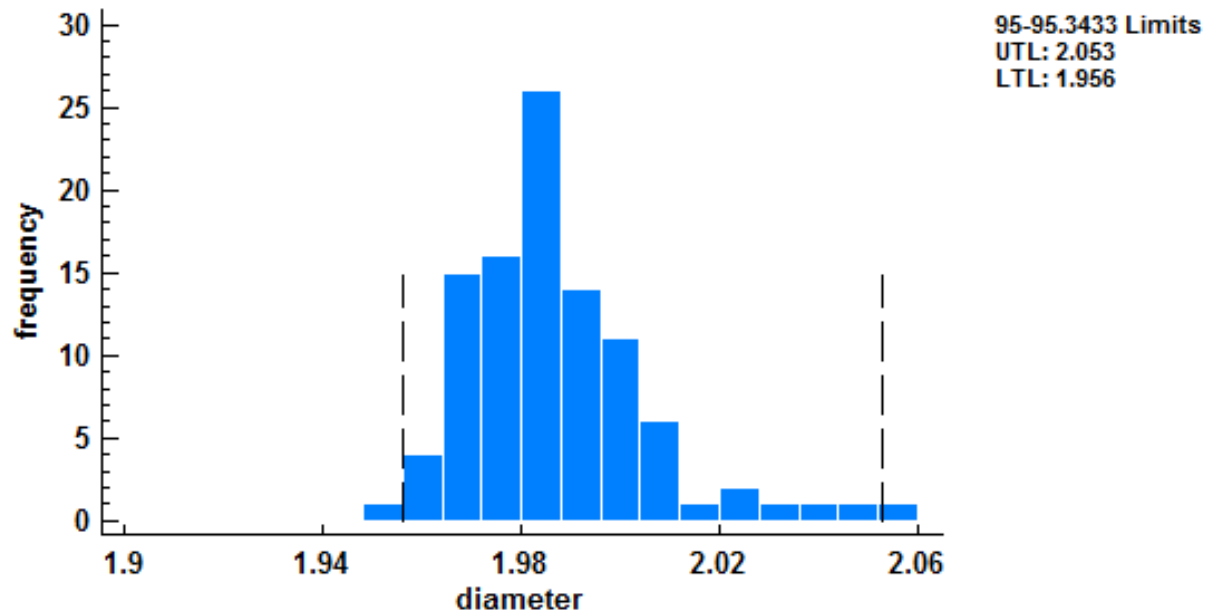
Pareto Threshold:

OK Cancel Help

- Tolerance interval is $[X_{(d)}, X_{(n-d+1)}]$

Nonparametric Limits

Nonparametric Tolerance Limits



Sample Size Determination – Nonparametric Limits

- How large a sample is needed so that the range [min, max] forms a 95-99 tolerance interval?

Sample Size Determination - Statistical Tolerance Limits

Distribution

Normal
Mean: Sigma:

Lognormal
Mean: Sigma: Threshold:

Weibull
Shape: Scale: Threshold:

Nonparametric

Type of Limits

Two-sided
 Lower limit only
 Upper limit only

Confidence Level: %
Population Proportion: %

Lower Spec. Limit: Upper Spec. Limit: Allowance: %

OK
Cancel
Help

Sample Size Determination (Statistical Tolerance Intervals)

<i>Conf. Level</i>	<i>Pop. Percentage</i>	<i>Distribution</i>
95.0%	99.0%	

The required sample size is 473.

<i>Lower tolerance limit</i>	<i>Upper tolerance limit</i>
smallest observation	largest observation

Sample Size Determination – Parametric Limits

- How large should n be when fitting a normal or some other distribution?
- Different approaches to the problem.
 1. Choose n so that the probability of including P^* % or more of the population in the tolerance interval is small, where $P^* > P$.
 2. Choose n so that the probability of the entire tolerance interval being within the specification limits is large.

Example: LEV Distribution

Sample Size Determination - Statistical Tolerance Limits

Distribution

Cauchy

Exponential

Exponential (2-parameter)

Gamma

Laplace

Largest extreme value

Lognormal

Normal

Normal after transformation

Pareto

Smallest extreme value

Weibull

Nonparametric

Mode

Scale

2.0

0.015

Type of Limits

Two-sided

Lower limit only

Upper limit only

Confidence Level:

95.0 %

Population Proportion:

99.0 %

Lower Spec. Limit:

1.9

Upper Spec. Limit:

2.1

Simulation

Inclusion Percentage:

90.0 %

Number of Trials:

10000

Maximum n:

30000

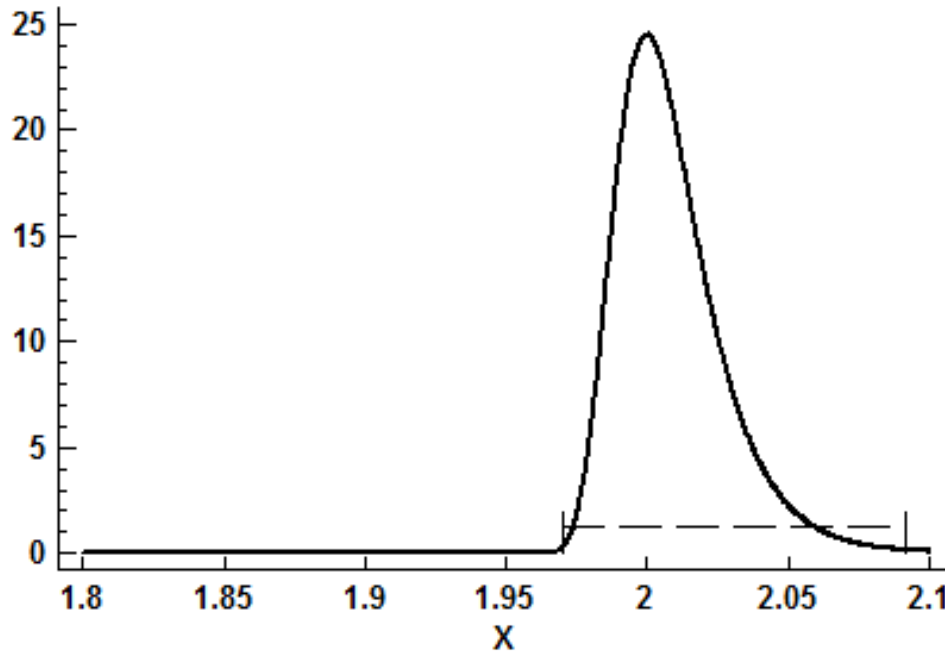
OK

Cancel

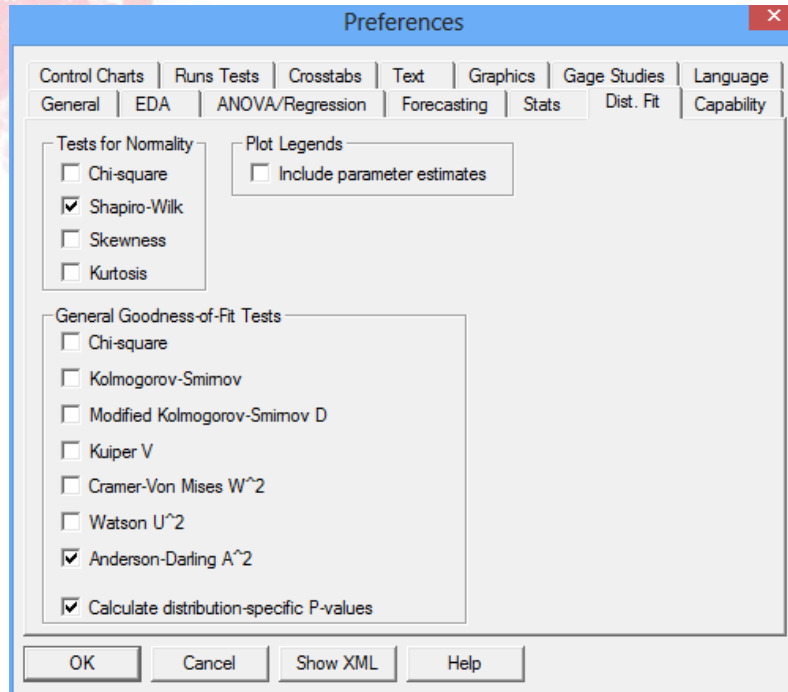
Help

Requires $n=120$ Observations

Largest extreme value distribution ($n=120$)
Mode=2.0, Scale=0.015



System Preferences



Preferences

Control Charts | Runs Tests | Crosstabs | Text | Graphics | Gage Studies | Language
General | EDA | ANOVA/Regression | Forecasting | Stats | Dist. Fit | Capability

Tests for Normality

- Chi-square
- Shapiro-Wilk
- Skewness
- Kurtosis

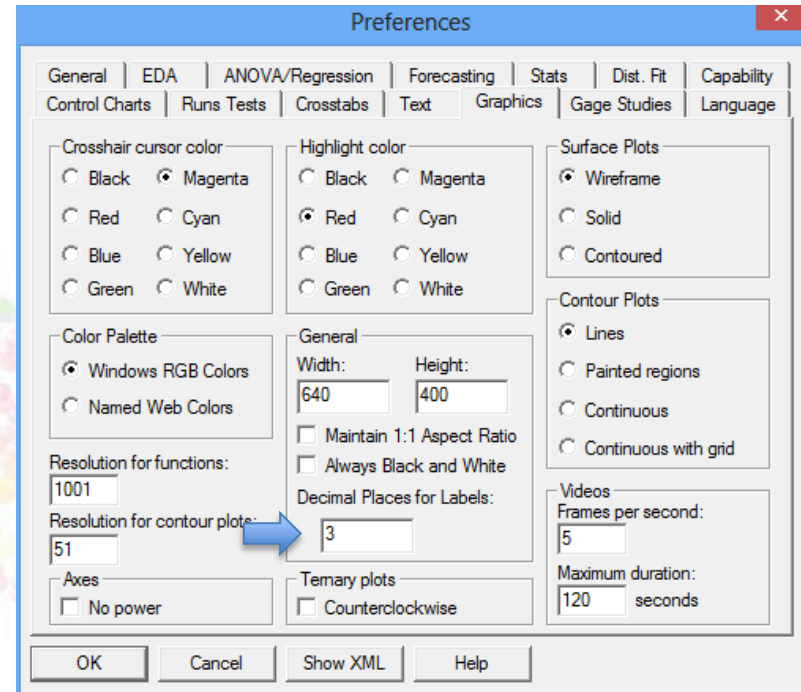
Plot Legends

- Include parameter estimates

General Goodness-of-Fit Tests

- Chi-square
- Kolmogorov-Smimov
- Modified Kolmogorov-Smimov D
- Kuiper V
- Cramer-Von Mises W^2
- Watson U^2
- Anderson-Darling A^2
- Calculate distribution-specific P-values

OK Cancel Show XML Help



Preferences

General | EDA | ANOVA/Regression | Forecasting | Stats | Dist. Fit | Capability
Control Charts | Runs Tests | Crosstabs | Text | Graphics | Gage Studies | Language

Crosshair cursor color

- Black
- Magenta
- Red
- Cyan
- Blue
- Yellow
- Green
- White

Highlight color

- Black
- Magenta
- Red
- Cyan
- Blue
- Yellow
- Green
- White

Color Palette

- Windows RGB Colors
- Named Web Colors

General

Width: 640 Height: 400

- Maintain 1:1 Aspect Ratio
- Always Black and White

Resolution for functions: 1001

Resolution for contour plots: 51

Decimal Places for Labels: 3

Surface Plots

- Wireframe
- Solid
- Contoured

Contour Plots

- Lines
- Painted regions
- Continuous
- Continuous with grid

Videos

Frames per second: 5

Maximum duration: 120 seconds

Axes

- No power

Ternary plots

- Counterclockwise

OK Cancel Show XML Help

References

- Video, slides and sample data may be found at www.statgraphics.com/webinars.
- Hahn, G.J., Meeker, W.Q. and Escobar, L.A. (2017) Statistical Intervals: A Guide for Practitioners and Researchers, second edition. Wiley, New York.