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Continuous Distributions Counting Techniques Descriptive Statistics Discrete Distributions One-Sample Confidence Intervals One-Sample Tests 1
Index 1
Continuous Distributions Continuous distributions

Description

Plot pdf/cdf, calculate mean/variance/standard deviation, and compute probabilities for various continuous distributions

2 Continuous Distributions

Usage

```
# continuous Uniform Distribution
uniform.summary(a,b,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
uniform.prob(a,b,lb,ub)
uniform.quantile(a,b,q)
# Normal distribution
normal.summary(mu,sigma,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
normal.prob(mu, sigma, lb, ub)
normal.quantile(mu, sigma, q)
# Exponential distribution
exponential.summary(lambda,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))\\
exponential.prob(lambda, lb, ub)
exponential.quantile(lambda,q)
# Gamma distribution
gamma.summary(r,lambda,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
gamma.prob(r,lambda,lb,ub)
gamma.quantile(r,lambda,p)
# Weibull distribution
weibull.summary(beta,delta,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
weibull.prob(beta,delta,lb,ub)
weibull.quantile(beta,delta,p)
# Lognormal distribution
lognormal.summary(theta,omega,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
lognormal.prob(theta,omega,lb,ub)
lognormal.quantile(theta,omega,p)
#Beta distribution
beta.summary(alpha,beta,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
beta.prob(alpha,beta,lb,ub)
beta.quantile(alpha,beta,p)
```

Arguments

plotpdf	TRUE or FALSE, if TRUE, it plots the pmf
plotcdf	TRUE or FALSE, if TRUE, it polts the cdf
lb,ub	lower bound (lb) and upper bound (ub) in a probability statement; lb could be -Inf; ub could be Inf; ub cannot be less than lb
a,b	lower bound and upper bound of the support of a continuous uniform distribution
mu,sigma	mean and standard deviation of a normal distribution
lambda	parameter of an exponential distribution
r,lambda	shape and scale parameters of a gamma distribution
beta,delta	shape and scale parameters of a Weibull distribution
theta,omega	shape and scale parameters of a lognormal distribution
alpba,beta	parameters of a beta distribution

Continuous Distributions 3

Details

Plot the probability density function (pdf) and the cumulative distribution function (cdf) and calculate probability, mean, variable and standard deviation, and compute probabilities of various discrete distributions (continuous uniform distribution, normal distribution, exponential distribution, gamma distribution, Weibull distribution, lognormal distribution, beta distribution).

Value

uniform.summary

a list of the mean, variance, and standard deviation of a continuous uniform distribution, plot of the pdf/cdf or not

 ${\tt uniform.prob}$ probability of X between 1b and ub based on a continuous uniform distribution ${\tt uniform.quantile}$

quantile of a continuous uniform distribution

normal.summary a list of the mean, variance, and standard deviation of a normal distribution, plot of the pdf/cdf or not

 $\begin{array}{ll} {\sf normal.prob} & {\sf probability} \ of \ X \ between \ lb \ and \ ub \ based \ on \ a \ normal \ distribution \\ {\sf normal.quantile} \end{array}$

quantile of a normal distribution

exponential.summary

a list of the mean, variance, and standard deviation of an exponential distribution, plot of the pdf/cdf or not

exponential.prob

probability of X between lb and ub based on an exponential distribution

exponential.quantile

quantile of an exponential distribution

gamma . summary a list of the mean, variance, and standard deviation of a gamma distribution, plot of the pdf/cdf or not

gamma.prob probability of X between lb and ub based on a gamma distribution

gamma.quantile quantile of a gamma distribution

weibull.summary

a list of the mean, variance, and standard deviation of a Weibull distribution, plot of the pdf/cdf or not

weibull.prob probability of X between lb and ub based on a Weibull distribution weibull.quantile

quantile of a Weibull distribution

lognormal.summary

a list of the mean, variance, and standard deviation of a lognormal distribution, plot of the pdf/cdf or not

lognormal.prob probability of X between lb and ub based on a lognormal.quantile

normat.quartite

quantile of a lognormal distribution

beta. summary a list of the mean, variance, and standard deviation of a beta distribution, plot of the pdf/cdf or not

beta.prob probability of X between lb and ub based on a beta distribution

beta.quantile quantile of a beta distribution

Note

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References

Chapter 4 of the textbook "Applied Statistics and Probability for Engineers" 7th edition

Examples

```
# Continuous uniform distribution
a=4.9;b=5.1;uniform.summary(a,b)
uniform.prob(a,b,4.95,5) # P(4.95 < X < 5)
uniform.quantile(a,b,0.9) # x such that P(X>x)=0.1
#Normal distribution
normal.summary(10,2) #mu=10,sigma=2,variance=4
normal.prob(10,2,9,11)
normal.quantile(10,2,0.98)
#Exponential distribution
exponential.summary(25) #lambda=25
exponential.prob(25,0.1,Inf) #P(X>0.1)
exponential.quantile(25,0.1) # x such that P(X>x)=0.9
#Gamma distribution
gamma.summary(10,.5) #r=10,lambda=0.5
gamma.prob(10,0.5,25,Inf) #P(X>25)
gamma.quantile(10,0.5,0.95) # x such that P(X<x)=0.95
#Weibull distribution
weibull.summary(2,5000) #beta=2,delta=5000
weibull.prob(2,5000,6000,Inf) \#P(X>6000)
weibull.quantile(2,5000,1-0.05) # x such that P(X>x)=0.05
#Lognormal distribution
lognormal.summary(10,1.5) #theta=10,omega=1.5
lognormal.prob(10,1.5,10000,Inf) #P(X>10000)
lognormal.quantile(10,1.5,1-0.99) # x such that P(X>x)=0.99
#Beta distribution
beta.summary(2.5,1) #alpha=2.5, beta=1
beta.prob(2.5,1,0.7,Inf) # P(X>0.7)
beta.quantile(2.5,1,0.99) # x such that P(X<x)=0.99
```

Counting Techniques

Sample Space Characterization

Description

Use permutations and combinations to count outcomes

Counting Techniques 5

Usage

```
nPr(n,r): permute r items from a set of n distinct items  nCr(n,r) \colon \text{ select r items from a set of n items where order does not matter } \\ SimPerm(n\_vec) \colon \text{ permute n items when they are not totally distinct}
```

Arguments

```
n total n itmes  \begin{tabular}{ll} $r$ & select $r$ \\  \begin{tabular}{ll} $n\_vec$ & a vector of $(n\_1,\,n\_2,...,n\_r)$ \\ \end{tabular}
```

Details

Permutation, Combination, and Permutation of similar items

Value

nPr(n,r) provides the number of different ways to permute r items from a set of n distinct items.

nCr(n,r) provides the number of different ways to select r items from a set of n items where order does not matter.

 $SimPerm(n_vec)$ provides the number of different ways to permute n items when they are not totally distinct.

Note

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References

Section 2.2 of the textbook "Applied Statistics and Probability for Engineers" 7th edition

Examples

```
nPr(8,4)
nCr(3,2)
nCr(47,4)
SimPerm(c(2,3,2))
```

6 Descriptive Statistics

Descriptive Statistics

Descriptive Statistics

Description

Calculate and plot descriptive statistics

Usage

```
data.summary(x)
```

Arguments

x the sample

plot True of False; plot or not the stem and leaf diagram, scatterplot, histogram,

boxplot, QQ-plot

Details

Calculate various numerical descritive statistics and plot the stem and leaf diagram, scatterplot, histogram, boxplot, QQ-plot

Value

data.summary

a table of numerical descritive statistics (std stands for standard deviation), a stem and leaf diagram, and a figure consisting of scatterplot, histogram, boxplot, QQ-plot of the sample

Note

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References

Chapter 6 of the textbook "Applied Statistics and Probability for Engineers" 7th edition

Examples

 $x = scan("https://raw.githubusercontent.com/Harrindy/StatEngine/master/Data/CompressiveStrength.csv") \\ data.summary(x)$

Discrete Distributions 7

Discrete Distributions

Discrete distributions

Description

Plot pmf/cdf, calculate mean/variance/standard deviation, and compute probabilities for various discrete distributions

Usage

```
# For use-defined discrete distribution
discrete.plotpdf(x,fx)
discrete.plotcdf(x,fx)
discrete.summary(x,fx,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE"))
discrete.prob(x,fx,lb)
discrete.prob(x,fx,lb,ub,inclusive=c("none","left","right","both"))
# Discrete Uniform Distribution
duniform.summary(range,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
duniform.prob(range, lb)
duniform.prob(range,lb,ub,inclusive=c("none","left","right","both"))
# Binomial distribution
binomial.summary(n,p,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
binomial.prob(n,p,lb)
binomial.prob(n,p,lb,ub,inclusive=c("none","left","right","both"))
# Geometric distribution
geometric.summary(p,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
geometric.prob(p,lb)
geometric.prob(p,lb,ub,inclusive=c("none","left","right","both"))
# Negative Binomial distribution
negbinom.summary(r,p,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
negbinom.prob(r,p,lb)
negbinom.prob(r,p,lb,ub,inclusive=c("none","left","right","both"))
# Hypergeometric distribution
hypergeo.summary(N,K,n,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
hypergeo.prob(N,K,n,lb)
hypergeo.prob(N,K,n,lb,ub,inclusive=c("none","left","right","both"))
# Poisson distribution
poisson.summary(lambda,L,plotpdf=c("TRUE","FALSE"), plotcdf=c("TRUE","FALSE")))
poisson.prob(lambda,L,lb)
poisson.prob(lambda,L,lb,ub,inclusive=c("none","left","right","both"))
```

Arguments

x possible values of a user defined discrete random variable

8 Discrete Distributions

fx probabilities of X=x, the order of entries in fx must matches the order of entries in x. plotpdf TRUE or FALSE, if TRUE, it plots the pmf TRUE or FALSE, if TRUE, it polts the cdf plotcdf 1b,ub lower bound (lb) and upper bound (ub) in a probability statement; lb could be -Inf; ub could be Inf; ub cannot be less than lb "none": lb<X<ub; "left": lb<=X<ub; "right": lb<X<=ub; "both": lb<=X<=ub inclusive contains all possible values of a discrete uniform random variable Range parameter p of a Binomial distribution/Geometric distribution/Negtive Binomial p distribution parameter n of a Binomial distribution n parameter r of a negative Binomial distribution N,K,nparameters N, K, and n of a hypergeometric distribution

Details

lambda, L

Plot the probability mass function (pmf) and the cumulative distribution function (cdf) and calculate probability, mean, variable and standard deviation, and compute probabilities of various discrete distributions (a self-defined discrete distribution, discrete uniform distribution, binomial distribution, geometric distribution, negative binomial distribution, hypergeometric distribution, and Poisson distribution).

parameters lambda and L of a Poisson distribution. Default: L=1

Value

discrete.plotpdf

a figure of the pmf of the user-defined discrete distribution

discrete.plotcdf

a figure of the cdf of the user-defined discrete distribution

discrete.summary

a list of the mean, variance, and standard deviation of the user-defined discrete

distribution, plot of the pmf/cdf or not

discrete.prob probability of X between lb and ub based on the user-defined discrete distribu-

tion

duniform.summary

a list of the mean, variance, and standard deviation of a discrete uniform distri-

bution, plot of the pmf/cdf or not

duniform.prob probability of X between lb and ub based on a uniform distribution

binomial.summary

a list of the mean, variance, and standard deviation of a binomial distribution,

plot of the pmf/cdf or not

binomial.prob probability of X between lb and ub based on a binomial distribution

geometric.summary

a list of the mean, variance, and standard deviation of a geometric distribution,

plot of the pmf/cdf or not

 $\label{eq:constraints} \mbox{geometric.prob} \ \ \mbox{probability of X between 1b and ub based on a geometric distribution}$

negbinom.summary

a list of the mean, variance, and standard deviation of a negative binomial distribution, plot of the pmf/cdf or not

Discrete Distributions 9

negbinom.prob probability of X between lb and ub based on a negative binomial distribution hypergeo.summary

a list of the mean, variance, and standard deviation of a hypergeometric distribution, plot of the pmf/cdf or not

hypergeo.prob probability of X between lb and ub based on a hypergeometric distribution poisson.summary

a list of the mean, variance, and standard deviation of a Poisson distribution, plot of the pmf/cdf or not

probability of X between lb and ub based on a Poisson distribution

Note

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poisson.prob

References

Chapter 3 of the textbook "Applied Statistics and Probability for Engineers" 7th edition

Examples

```
x=c(0,1,2,3,4); fx=c(0.6561,0.2916,0.0486,0.0036,0.0001);
discrete.summary(x,fx)
\label{linear_discrete} \\ discrete.summary(\texttt{x},\texttt{fx},\texttt{plotpdf=FALSE},\texttt{plotcdf=TRUE}) \\
discrete.summary(x,fx,plotpdf=TRUE,plotcdf=FALSE)
discrete.prob(x, fx, 2) #P(X=2)
discrete.prob(x,fx,2,4,inclusive="none") #P(2<X<4)</pre>
discrete.prob(x,fx,2,4,inclusive="left") \#P(2\leq X\leq 4)
discrete.prob(x,fx,2,4,inclusive="right") #P(2<X<=4)</pre>
discrete.prob(x, fx, 2, 4, inclusive="both") #P(2<=X<=4)
discrete.prob(x,fx,2,Inf,inclusive="none") #P(2<X)</pre>
discrete.prob(x,fx,2,Inf,inclusive="left") #P(2<=X)</pre>
discrete.prob(x,fx,2,Inf,inclusive="right") #P(2<X)</pre>
discrete.prob(x,fx,2,Inf,inclusive="both") #P(2<=X)</pre>
discrete.prob(x,fx,-Inf,4,inclusive="none") #P(X<4)</pre>
discrete.prob(x,fx,-Inf,4,inclusive="left") #P(X<4)</pre>
discrete.prob(x,fx,-Inf,4,inclusive="right") #P(X<=4)</pre>
discrete.prob(x,fx,-Inf,4,inclusive="both") \#P(X\leq 4)
#X~Discrete Uniform(1,2,4,5,8,10)
range=c(1,2,4,5,8,10)
duniform.summary(range)
duniform.prob(range,2) #P(X=2)
duniform.prob(range,2,4,inclusive="left") #P(2<=X<4)</pre>
#X~Binomial(n=5,p=0.3)
binomial.summary(5,0.3)
```

```
binomial.prob(5,0.3,2) \#P(X=2)
binomial.prob(5,0.3,2,4,inclusive="left") #P(2<=X<4)
#X~Geometric(p=0.3)
geometric.summary(0.3)
geometric.prob(0.3,2) #P(X=2)
geometric.prob(0.3,2,4,inclusive="left") #P(2<=X<4)</pre>
#X~Negative Binomial(r=3,p=0.3)
negbinom.summary(3,0.3)
negbinom.prob(3,0.3,5) #P(X=5)
negbinom.prob(3,0.3,5,7,inclusive="left") \#P(5 \le X < 7)
#X~Hypergeometric(N=300,K=100,n=4)
hypergeo.summary(300,100,4)
hypergeo.prob(300,100,4,2) #P(X=2)
hypergeo.prob(300,100,4,2,4,inclusive="left") \#P(2 \le X \le 4)
#X~Poisson(lambda=2.3,L=5)
poisson.summary(2.3,5)
poisson.prob(2.3,5,10) #P(X=10)
poisson.prob(2.3,5,1,Inf,inclusive="left") #P(1<=X)</pre>
```

```
One-Sample Confidence Intervals One\text{-}Sample \ \textit{Tests}
```

Description

Conduct hypothesis testing on population mean, population variance, and population proportion. Compute power of a test, and calculate the required sample size for a desired power.

Usage

```
#Test on pupulation mean of a normal distribution when the population variance is known:
Ztest(mu0,H1,alpha,sigma,sample) # if sample available
Ztest(mu0,h1,alpha,sigma,n,barx) # if statistics are provided
# Power calculation
Ztest.power(H1,alpha,sigma,n,delta)
# Sample size calculation
sample.size.Ztest(H1,sigma,alpha,beta,delta)
#Test on pupulation mean of a normal distribution when the population variance is unknown:
Ttest(mu0=?,H1=?,alpha=?,sample=?) #If data are available
Ttest(mu0=?,H1=?,alpha=?,n=?,barx=?,s=?) #If statistics are provided s=sd(data)
# Power calculation
Ttest.power(H1=?,est.sigma=sn,alpha=?,n=?,delta=?)
# Sample size calculation
sample.size.Ttest(H1=?,est.sigma=sn, beta=?,delta=?,alpha=?
#One-Sample Proportion Z-tests on a population proportion:
Proptest(p0=?,H1=?,alpha=?,n=?,X=?)
# Power calculation
```

```
Proptest.power(H1=?,alpha=?,p0=?,p1=?,n=?)
# Sample size calculation
sample.size.Proptest(H1=?,beta=?,alpha=?,p0=?,p1=?)
```

Arguments

mu0 the hypothesized value of the population mean in Z-tests and T-tests

H1 type of alternative hypothesis: "two", "left", or "right"

alpha the significance level

sample a vector of the observed sample

sigma the known population standard deviation

n the sample size

barx the observed sample mean

delta the delta value in power/sample size calculation of Z-test/T-test

beta the desired power for sample size calculation
s the observed sample standard deviation

est.sigma an estimate of the population standard deviation in power/sample size calcula-

tion of T-test

sigma0 the hypothesized value of the population standard deviation in Chi-square-tests

lambda the lambda value in power/sample size calculation of Chi-square-tests

p0 the hypothesized value of the population proportion X number of observations belongs to a class of interest

p1 the p1 value in power/sample size calculation of One-sample Prportion Z-tests

Details

Conduct hypothesis testing on population mean, population variance, and population proportion. Compute power of a test, and calculate the required sample size for a desired power.

Value

test As long as the function has "test", it produces the test results of using three

approaches.

power As long as the function has "test", it computes the power of the test.

sample.size As long as the function has "sample.size", the outcome is the minimum sample

size required to reach the given power.

Note

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Author(s)

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References

Chapter 9 of the textbook "Applied Statistics and Probability for Engineers" 7th edition

12 One-Sample Tests

Examples

```
#Ztest
#must include the = sign
Ztest(mu0=3,H1="two",alpha=0.05,sigma=0.9,n=15,barx=2.78)
Ztest.power(H1="two",alpha=0.05,sigma=0.9,n=15,delta=3.25-3)
sample.size.Ztest(H1="two",sigma=0.9,alpha=0.05,beta=1-0.9,delta=3.75-3)
#Ttest
#must include the = sign
x=c(131.15, 130.69, 130.91, 129.54, 129.64, 128.77, 130.72, 128.33,
128.24, 129.65, 130.14, 129.29, 128.71, 129.00, 129.39, 130.42,
129.53, 130.12, 129.78, 130.92)
data.summary(x)
Ttest(mu0=130,H1="two",alpha=0.05,sample=x)
Ttest.power(H1="two",est.sigma=sd(x),alpha=0.05, n=length(x),delta=0.5)
sample.size.Ttest(H1="two",est.sigma =sd(x),beta=0.25,delta=0.1,alpha=0.05)
#Chi-square test
#must include the = sign
Chi2test(sigma0=sqrt(0.01),H1="right",alpha=0.05,n=20,s=sqrt(0.0153))
Chi2test.power(H1="right",alpha=0.05,n=20,lambda=1.25)
sample.size.Chi2test(H1="right", beta=0.2,lambda=1.25,alpha=0.05)
Chi2test(sigma0=0.01,H1="right",alpha=0.01,n=15,s=0.008)
Chi2test.power(H1="right",alpha=0.01,n=15,lambda=1.5)
sample.size.Chi2test(H1="right", beta=0.2, lambda=1.25, alpha=0.01)
#One-sample proportion Z-test
#must include the = sign
Proptest(p0=0.05, H1="left",alpha=0.05,n=300,X=7)
Proptest.power(H1="left",alpha=0.05,p0=0.05,p1=0.03,n=300)
sample.size.Proptest(H1="left",beta=0.1,alpha=0.05,p0=0.05,p1=0.03)
```

One-Sample Tests

One-Sample Confidence Intervals

Tinterval(level, sample) # if sample available

Description

Compute confidence intervals on the population mean, population variance, and population proportion. In addition, it computes prediction interval for a single future observation from a normal distribution.

Usage

```
#CI for pupulation mean of a normal distribution when the population variance is known:
Zinterval(level,sigma,sample) # if sample available
Zinterval(level,sigma,n,barx) # if stats are provided
# Choice of sample size for estimating the population mean when error is specified
sample.size.Zinterval(level,sigma,E)
#CI for pupulation mean when the population variance is unknown and the distribution is normal
#or when the sample size is smaller than 25:
```

One-Sample Tests 13

```
Tinterval(level,n,barx,s) # if stats are provided

#Large-sample CI for pupulation mean:
AZinterval(level,sample) # if sample available
AZinterval(level,n,barx,s) # if stats are provided

#CI for pupulation variance (or standard deviation) of a normal distribution:
Chi2interval(level,sample) # if sample available
Chi2interval(level,n,s) # if stats are provided

#Large-sample CI for a pupulation proportion:
Propinterval(level,n,X)

# Choice of sample size for estimating a population proportion when error is specified sample.size.Propinterval(level,ini.p,E) # using an intial guess
sample.size.Propinterval(level,ini.p=0.5,E) # using the conservative apporach

# Prediction interval of a single future observation form a normal distribution:
Predinterval(level,sample) # if sample available
Predinterval(level,n,barx,s) # if stats are provided
```

Arguments

level	the confidence level
sample	a vector of the observed sample
sigma	the known population standard deviation
S	the observed sample standard deviation
barx	the observed sample mean
n	the sample size
Χ	number of observations belongs to a class of interest
E	specified error in sample size calculation
ini.p	A initial estimate of the populatin proportion. Default is 0.5 which corresponds to the conservative approach
df	the degrees of freedom of a t or chi.square distribution
q	a quantile value

Details

Compute CIs for the population mean, population variance, and population proportaion and PI for a single future observation from a normal distribution.

Value

interval	As long as the function has "interval", the outcome contains a two-sided CI (or PI) and the two one-sided confidence bounds.
sample.size	As long as the function has "sample.size", the outcome is the minimum sample size required to control the error to be no larger than E.
t.quantile	quantile of a t distribution
Chi2.quantile	quantile of a chi.square distribution

14 One-Sample Tests

Note

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Author(s)

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References

Chapter 8 of the textbook "Applied Statistics and Probability for Engineers" 7th edition

Examples

```
#Zinterval
\#must include the = sign
x=c(64.1, 64.7, 64.5, 64.6, 64.5, 64.3, 64.6, 64.8, 64.2, 64.3)
Zinterval(level=0.95, sigma=1, sample=x)
Zinterval(level=0.99, sigma=1, sample=x)
sample.size.Zinterval(E=0.5, sigma=1, level=0.95)
# Using stats, must include the = sign
Zinterval(level=0.95, sigma=2, n=9, barx=98)
#Tinterval
\#must include the = sign
Tinterval(level=0.95, n=10, barx=1000, s=20)
Tinterval(level=0.95,n=25,barx=1000,s=20)
Tinterval(level=0.99, n=10, barx=1000, s=20)
Tinterval(level=0.99, n=25, barx=1000, s=20)
#Large-sample Zinterval
#must include the = sign
x=scan("https://raw.githubusercontent.com/Harrindy/StatEngine/master/Data/Mercury.csv")
AZinterval(level=0.95, sample=x)
#Chi.square interval for variance/standard deviation
#must include the = sign
Chi2interval(level=0.95, n=20, s=0.01532)
#CIs for a porpulation proportion
#must include the = sign
Propinterval(level=0.95, n=85, X=10)
sample.size.Propinterval(level=0.95,ini.p=0.12,E=0.05)
sample.size.Propinterval(level=0.95,ini.p=0.5,E=0.05)
#Prediction interval for normal distribution
#must include the = sign
x=c(19.8, 10.1, 14.9, 7.5, 15.4, 15.4, 15.4, 18.5, 7.9, 12.7, 11.9, 11.4, 11.4,
14.1, 17.6, 16.7, 15.8, 19.5, 8.8, 13.6, 11.9, 11.4)
Tinterval(level=0.95,sample=x)
Predinterval(level=0.95, sample=x)
```

Index

AZinterval (One-Sample Tests), 12	geometric.prob(Discrete
	Distributions), 7
beta.prob (Continuous Distributions), 1	<pre>geometric.summary(Discrete</pre>
beta.quantile (Continuous	Distributions), 7
Distributions), 1	
beta.summary(Continuous	hypergeo.prob(Discrete Distributions),
Distributions), 1	7
binomial.prob(Discrete Distributions),	hypergeo.summary(Discrete
7	Distributions), 7
binomial.summary(Discrete	1 1 (0 1)
Distributions), 7	lognormal.prob(Continuous
	Distributions), 1
Chi2.quantile (One-Sample Tests), 12	lognormal.quantile (Continuous
Chi2interval (One-Sample Tests), 12	Distributions), 1
Chi2test (One-Sample Confidence	lognormal.summary(Continuous
Intervals), 10	Distributions), 1
Continuous Distributions, 1	accor (Counting Techniques) A
Counting Techniques, 4	nCr (Counting Techniques), 4
DataSummary (Descriptive Statistics) 6	negbinom.prob(Discrete Distributions),
DataSummary (Descriptive Statistics), 6 Descriptive Statistics, 6	/
Discrete Distributions, 7	negbinom.summary(Discrete
·	Distributions), 7
discrete.plotcdf(Discrete Distributions), 7	normal.prob(Continuous Distributions),
discrete.plotpdf (Discrete	1
	normal.quantile (Continuous
Distributions), 7	Distributions), 1
discrete.summary(Discrete	normal.summary (Continuous
Distributions), 7	Distributions), 1
duniform.prob(Discrete Distributions),	nPr (Counting Techniques), 4
duniform cummary (Discrete	One-Sample Confidence Intervals, 10
duniform.summary(Discrete	One-Sample Tests, 12
Distributions), 7	one-sample rests, 12
exponential.prob(Continuous	<pre>poisson.prob (Discrete Distributions), 7</pre>
Distributions), 1	poisson.summary (Discrete
exponential.quantile (Continuous	Distributions), 7
Distributions), 1	Predinterval (One-Sample Tests), 12
exponential.summary (Continuous	Propinterval (One-Sample Tests), 12
Distributions), 1	Proptest (One-Sample Confidence
51301130010113), 1	Intervals), 10
gamma.prob(Continuous Distributions), 1	
gamma.quantile (Continuous	<pre>sample.size.Chi2test (One-Sample</pre>
Distributions), 1	Confidence Intervals), 10
gamma.summary (Continuous	<pre>sample.size.Propinterval(One-Sample</pre>
Distributions), 1	Tests), 12

16 INDEX

```
sample.size.Proptest(One-Sample
        Confidence Intervals), 10
sample.size.Ttest (One-Sample
        Confidence Intervals), 10
sample.size.Zinterval (One-Sample
        Tests), 12
sample.size.Ztest(One-Sample
        Confidence Intervals), 10
SimPerm (Counting Techniques), 4
t.quantile (One-Sample Tests), 12
Tinterval (One-Sample Tests), 12
Ttest (One-Sample Confidence
        Intervals), 10
uniform.prob(Continuous
        Distributions), 1
uniform.quantile (Continuous
        Distributions), 1
uniform.summary(Continuous
        Distributions), 1
weibull.prob(Continuous
        Distributions), 1
weibull.quantile (Continuous
        Distributions), 1
weibull.summary(Continuous
        Distributions), 1
Zinterval (One-Sample Tests), 12
Ztest (One-Sample Confidence
        Intervals), 10
```