

# Package: PPQplan (via r-universe)

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**Type** Package

**Title** Process Performance Qualification (PPQ) Plans in Chemistry,  
Manufacturing and Controls (CMC) Statistical Analysis

**Version** 1.1.0

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**Depends** R (>= 3.2.0)

**Imports** ggplot2, plotly

**Description** Assessment for statistically-based PPQ sampling plan, including calculating the passing probability, optimizing the baseline and high performance cutoff points, visualizing the PPQ plan and power dynamically. The analytical idea is based on the simulation methods from the textbook Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). Statistical Methods for CMC Applications. In Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry (pp. 227-250). Springer, Cham.

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**URL** <https://allenzhuaz.github.io/PPQplan/>,  
<https://github.com/allenzhuaz/PPQplan>

**BugReports** <https://github.com/allenzhuaz/PPQplan/issues>

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**Repository** <https://allenzhuaz.r-universe.dev>

**RemoteUrl** <https://github.com/allenzhuaz/ppqplan>

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## Contents

heatmap_ly . . . . .	2
k_factor . . . . .	3
pi_ctplot . . . . .	4
pi_occurve . . . . .	5
pi_pp . . . . .	7
pp . . . . .	8
PPQ_ctplot . . . . .	9
PPQ_ggplot . . . . .	10
PPQ_occurve . . . . .	12
PPQ_pp . . . . .	13
rl_pp . . . . .	15
ti_ctplot . . . . .	16
ti_occurve . . . . .	17
ti_pp . . . . .	18

**Index** **20**

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heatmap_ly	<i>A General Heatmap for Dynamically Assessing Power of the Sampling Plan Using a General Specification Limit.</i>
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## Description

The function for dynamically plotting (ggplot) the heatmap to evaluate the sampling plan based on a general lower and/or upper specification limits.

## Usage

```
heatmap_ly(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, test.point, dynamic)
```

## Arguments

attr.name	(optional) user-defined attribute name for sampling plan assessment
attr.unit	(optional) user-defined attribute unit
Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
test.point	(optional) actual process data points for testing whether the processes pass PPQ
dynamic	logical; if TRUE, then convert the plain heatmap to dynamic graph using plotly.

**Value**

A Plain or Dynamic Heatmap for Sampling Plan Assessment.

**Author(s)**

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry. *Springer*.

**See Also**

pp and PPQ. occurve.

**Examples**

```
## Not run:
heatmap_ly(attr.name = "Thickness", attr.unit = "%", Llim = -0.2, Ulim = 0.2,
mu = seq(-0.2, 0.2, 0.001), sigma = seq(0,0.2, 0.001),
test.point=data.frame(c(0.1,-0.05),c(0.15,0.05)), n=2, dynamic = T)

## End(Not run)
```

---

k\_factor

*Estimating K-factors for Tolerance Intervals Based on Howe's Method*


---

**Description**

Estimates k-factors for tolerance intervals based on Howe's method with normality assumption.

**Usage**

```
k_factor(n, alpha = 0.05, P = 0.99, side = 1)
```

**Arguments**

n	Sample size
alpha	The level chosen such that (1-alpha) is the confidence level.
P	The proportion of the population to be covered by the tolerance interval.
side	Whether a 1-sided or 2-sided tolerance interval is required (determined by side = 1 or side = 2, respectively).

**Value**

The estimated k-factor for tolerance intervals assuming normality.

**Note**

This function is a simplified version of `tolerance::K.factor()`, only considering Howe's method.

**See Also**

`ti_pp`

**Examples**

```
k_factor(10, P = 0.95, side = 2)
```

---

pi_ctplot	<i>Heatmap/Contour Plot for Assessing Power of the CQA PPQ Plan Using Prediction Interval.</i>
-----------	--

---

**Description**

The function for plotting the heatmap to evaluate the PPQ plan based on the specification test, given lower and upper specification limits.

**Usage**

```
pi_ctplot(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, n.batch, alpha, test.point)
```

**Arguments**

attr.name	user-defined attribute name for PPQ assessment
attr.unit	user-defined attribute unit
Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
alpha	significant level for constructing the prediction interval.
test.point	(optional) actual process data points for testing whether the processes pass PPQ

**Value**

Heatmap (or Contour Plot) for PPQ Assessment.

**Author(s)**

Yalin Zhu

## References

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

## See Also

pi\_pp and pi\_occurve.

## Examples

```
## Not run:
## Example verifying simulation results in the textbook page 249
mu <- seq(95, 105, 0.1)
sigma <- seq(0.2, 3.5, 0.1)
pi_ctplot(attr.name = "Composite Assay", attr.unit = "%LC",
mu = mu, sigma = sigma, Llim=95, Ulim=105)
mu <- seq(90, 110, 0.5)
pi_ctplot(attr.name = "Composite Assay", attr.unit = "%LC",
mu = mu, sigma = sigma, Llim=90, Ulim=110)

mu <- seq(95,105,0.1)
sigma <- seq(0.1,2.5,0.1)
pi_ctplot(attr.name = "Sterile Concentration Assay", attr.unit = "%",
mu = mu, sigma = sigma, Llim=95, Ulim=105)
test <- data.frame(mean=c(97,98.3,102.5), sd=c(0.55, 1.5, 1.2))
pi_ctplot(attr.name = "Sterile Concentration Assay", attr.unit = "%", Llim=95, Ulim=105,
mu = mu, sigma = sigma, test.point=test)

## End(Not run)
```

---

pi\_occurve

*Operating Characteristic (OC) Curves for the CQA PPQ Plan using Prediction Interval.*

---

## Description

The function for plotting the OC curves and optimizing the baseline and high performance PPQ plans, given lower and upper specification limits.

## Usage

```
pi_occurve(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, n.batch, alpha, add.reference)
```

**Arguments**

attr.name	user-defined attribute name
attr.unit	user-defined attribute unit
Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
alpha	significant level for constructing the prediction interval.
add.reference	logical; if TRUE, then add reference OC curves (Baseline and High Performance) in the plot.

**Value**

OC curves for specification test and PPQ plan.

**Author(s)**

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

**See Also**

pi\_pp and rl\_pp.

**Examples**

```
## Not run:
pi_ocurve(attr.name = "Total Protein", attr.unit = "mg/mL",
  sigma = seq(0.01,1,0.01))
pi_ocurve(attr.name = "Total Protein", attr.unit = "mg/mL",
  sigma = seq(0.01,1,0.01), n.batch=3)
# Baseline curve
pi_ocurve(attr.name = "Total Protein", attr.unit = "mg/mL",
  sigma = seq(0.01,1,0.01), alpha = 0.1135434)
# High performance curve
pi_ocurve(attr.name = "Total Protein", attr.unit = "mg/mL",
  sigma = seq(0.01,1,0.01), alpha = 0.0225518)

# 95% with reference curves
pi_ocurve(attr.name = "Total Protein", attr.unit = "mg/mL",
```

```

sigma = seq(0.01,1,0.01), add.reference=TRUE)
pi_ocurve(attr.name = "Composite Assay", attr.unit = "%",
mu = 100, sigma = seq(0.1,6,0.1), Llim=95, Ulim=105, n.batch=1, add.reference=TRUE)

pi_ocurve(attr.name = "Sterile Concentration Assay", attr.unit="%",
mu=97, sigma=seq(0.1, 10, 0.1), Llim=95, Ulim=105, n=10, add.reference=TRUE)

pi_ocurve(attr.name = "Sterile Concentration Assay", attr.unit="%",
mu=100, sigma=seq(0.1, 10, 0.1), Llim=95, Ulim=105, n=10, add.reference=TRUE)

pi_ocurve(attr.name = "Sterile Concentration Assay", attr.unit="%",
mu=seq(95,105,0.1), sigma=1, Llim=95, Ulim=105, n=10, add.reference=TRUE)

pi_ocurve(attr.name = "Protein Concentration", attr.unit="%",
mu=seq(90, 110, 0.1), sigma=1.25, Llim=90, Ulim=110, add.reference=TRUE)

## End(Not run)

```

---

pi\_pp

---

*Probability of Passing PPQ Test using Prediction Interval*


---

### Description

The function for calculating the probability of passing critical quality attributes (CQA) PPQ test .

### Usage

```
pi_pp(Llim, Ulim, mu, sigma, n, n.batch, alpha)
```

### Arguments

Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
alpha	significant level for constructing the prediction interval.

### Value

A numeric value of the passing/acceptance probability

### Author(s)

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

**See Also**

r1\_pp.

**Examples**

```
## Not run:
pi_pp(sigma=0.5, mu=2.5, n=10, n.batch=1, Llim=1.5, Ulim=3.5, alpha=0.05)

sapply(X=c(0.1,0.5, 1,2,3,4,5,10), FUN = pi_pp, mu=97, n=10, Llim=95, Ulim=105,
n.batch=1, alpha=0.05)
sapply(X=c(0.1,0.5, 1,2,3,4,5,10), FUN = pi_pp, mu=100, n=10, Llim=95, Ulim=105,
n.batch=1, alpha=0.05)

## End(Not run)
```

---

pp	<i>Probability of Passing General Upper and/or Lower Specification Limit</i>
----	--

---

**Description**

The function for calculating the probability of passing a general upper and/or lower boundary.

**Usage**

```
pp(Llim, Ulim, mu, sigma, n)
```

**Arguments**

Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations)

**Value**

A numeric value of the passing/acceptance probability

**Author(s)**

Yalin Zhu



**See Also**

r1\_pp and PPQ\_pp.

---

PPQ_ctplot	<i>Heatmap/Contour Plot for Assessing Power of the CQA PPQ Plan Using General Multiplier.</i>
------------	---

---

**Description**

The function for plotting the heatmap to evaluate the PPQ plan based on the specification test, given lower and upper specification limits.

**Usage**

```
PPQ_ctplot(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, n.batch, k, test.point)
```

**Arguments**

attr.name	(optional) user-defined attribute name for PPQ assessment
attr.unit	(optional) user-defined attribute unit
Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
k	general multiplier for constructing the specific interval
test.point	(optional) actual process data points for testing whether the processes pass PPQ

**Value**

Heatmap (or Contour Plot) for PPQ Assessment.

**Author(s)**

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

**See Also**

PPQ\_pp and PPQ\_occurve.

**Examples**

```
## Not run:
mu <- seq(1.6,3.4,0.05)
sigma <- seq(0.05,0.8,0.01)
PPQ_ctplot(attr.name = "Total Protein", attr.unit = "mg/mL", Llim=1.5, Ulim=3.5,
mu = mu, sigma = sigma, k=2.373)

## Example verifying simulation results in the textbook page 249
mu <- seq(95, 105, 0.1)
sigma <- seq(0.2, 5, 0.1)
PPQ_ctplot(attr.name = "Composite Assay", attr.unit = "%LC", Llim=95, Ulim=105,
mu = mu, sigma = sigma, k=2.373)
mu <- seq(90, 110, 0.5)
PPQ_ctplot(attr.name = "Composite Assay", attr.unit = "%LC", Llim=90, Ulim=110,
mu = mu, sigma = sigma, k=2.373)

mu <- seq(95,105,0.1)
sigma <- seq(0.1,2.5,0.1)
PPQ_ctplot(attr.name = "Sterile Concentration Assay", attr.unit = "%", Llim=95, Ulim=105,
mu = mu, sigma = sigma, k=2.373)
test <- data.frame(mean=c(97,98.3,102.5), sd=c(0.55, 1.5, 1.2))
PPQ_ctplot(attr.name = "Sterile Concentration Assay", attr.unit = "%", Llim=95, Ulim=105,
mu = mu, sigma = sigma, k=2.373, test.point=test)

## End(Not run)
```

---

PPQ\_ggplot

*Heatmap/Contour Plot for Dynamically Assessing Power of the CQA  
PPQ Plan Using General Multiplier.*

---

**Description**

The function for dynamically plotting (ggplot) the heatmap to evaluate the PPQ plan based on the specification test, given lower and upper specification limits.

**Usage**

```
PPQ_ggplot(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, n.batch, k,
test.point, dynamic)
```

**Arguments**

attr.name        (optional) user-defined attribute name for PPQ assessment  
attr.unit        (optional) user-defined attribute unit

Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
k	general multiplier for constructing the specific interval
test.point	(optional) actual process data points for testing whether the processes pass PPQ
dynamic	logical; if TRUE, then convert the heatmap ggplot to dynamic graph using plotly.

**Value**

Dynamic Heatmap (or Contour Plot) for PPQ Assessment.

**Author(s)**

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

**See Also**

PPQ\_pp and PPQ\_occurve.

**Examples**

```
## Not run:
mu <- seq(95, 105, 0.1)
sigma <- seq(0.1, 1.7, 0.1)
PPQ_ggplot(attr.name = "Sterile Concentration Assay", attr.unit = "%", Llim=95, Ulim=105,
mu = mu, sigma = sigma, k=2.373, dynamic = FALSE)
test <- data.frame(mu=c(97,98.3,102.5), sd=c(0.55, 1.5, 0.2))
PPQ_ggplot(attr.name = "Sterile Concentration Assay", attr.unit = "%", Llim=95, Ulim=105,
mu = mu, sigma = sigma, k=2.373, test.point = test)

## End(Not run)
```

---

PPQ\_occurve                      *Operating Characteristic (OC) Curves for the CQA PPQ Plan Using General Multiplier.*

---

### Description

The function for plotting the OC curve to show the PPQ plan, given lower and upper specification limits.

### Usage

```
PPQ_occurve(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, n.batch, k, add.reference)
```

### Arguments

attr.name	(optional) user-defined attribute name
attr.unit	(optional) user-defined attribute unit
Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
k	general multiplier for constructing the specific interval
add.reference	logical; if TRUE, then add reference OC curves (Baseline and High Performance) in the plot.

### Value

OC curves for specification test and PPQ plan.

### Author(s)

Yalin Zhu

Yalin Zhu

### References

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

### See Also

PPQ\_pp and r1\_pp.

**Examples**

```
## Not run:
PPQ_occurve(attr.name = "Sterile Concentration Assay", attr.unit="%", Llim=95, Ulim=105,
mu=97, sigma=seq(0.1, 10, 0.1), n=10, k=2.373, add.reference=TRUE)
PPQ_occurve(attr.name = "Sterile Concentration Assay", attr.unit="%", Llim=95, Ulim=105,
mu=100, sigma=seq(0.1, 10, 0.1), n=10, k=2.373, add.reference=TRUE)
PPQ_occurve(attr.name = "Sterile Concentration Assay", attr.unit="%", Llim=95, Ulim=105,
mu=seq(95,105,0.1), sigma=1, n=10, k=2.373)
PPQ_occurve(attr.name = "Sterile Concentration Assay", attr.unit="%", Llim=95, Ulim=105,
mu=seq(95,105,0.1), sigma=1, n=10, k=2.373, add.reference=TRUE)

PPQ_occurve(attr.name = "Protein Concentration", attr.unit="%", Llim=90, Ulim=110,
mu=seq(90, 110, 0.1), sigma=1.25, k=2.373)

## Only display referece curves, leave k as NULL by default
PPQ_occurve(attr.name = "Sterile Concentration Assay", attr.unit="%LC", Llim=95, Ulim=105,
mu=98, sigma=seq(0.1, 10, 0.1), n=10, add.reference=TRUE)

## End(Not run)
```

---

PPQ\_pp

---

*Probability of Passing PPQ Test Using General Multiplier*


---

**Description**

The function for calculating the probability of passing critical quality attributes (CQA) PPQ test .

**Usage**

```
PPQ_pp(Llim, Ulim, mu, sigma, n, n.batch, k)
```

**Arguments**

Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
k	general multiplier for constructing the specific interval

**Value**

A numeric value of the passing/acceptance probability

**Author(s)**

Yalin Zhu

## References

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

## See Also

r1\_pp.

## Examples

```
## Not run:
PPQ_pp(Llim = 90, Ulim = 110, mu=105, sigma=1.5, n=10, k=3.1034)

# One-sided tolerance interval with k=0.753 (95/67.5 one-sided tolerance interval LTL)
PPQ_pp(sigma=0.03, mu=1.025, n=40, Llim=1, Ulim=Inf, k=0.753)

sapply(X=c(0.1,0.5, 1,2,3,4,5,10), FUN = PPQ_pp, mu=97, n=10, Llim=95, Ulim=105, k=2.373)
sapply(X=seq(0.1,10,0.1), FUN = PPQ_pp, mu=97, n=10, Llim=95, Ulim=105, k=2.373)

sapply(X=c(0.1,0.5, 1,2,3,4,5,10), FUN = PPQ_pp, mu=100, n=10, Llim=95, Ulim=105, k=2.373)

sigma <- seq(0.1, 4, 0.1)
pp1 <- sapply(X=sigma, FUN = PPQ_pp, mu=97, n=10, Llim=95, Ulim=105, k=2.373)
pp2 <- sapply(X=sigma, FUN = PPQ_pp, mu=98, n=10, Llim=95, Ulim=105, k=2.373)
pp3 <- sapply(X=sigma, FUN = PPQ_pp, mu=99, n=10, Llim=95, Ulim=105, k=2.373)
pp4 <- sapply(X=sigma, FUN = PPQ_pp, mu=100, n=10, Llim=95, Ulim=105, k=2.373)
plot(sigma, pp1, xlab="Standard Deviation", main="LSL=95, USL=105, k=2.373, n=10",
ylab="Probability of Passing", type="o", pch=1, col=1, lwd=1, ylim=c(0,1))
lines(sigma, pp2, type="o", pch=2, col=2)
lines(sigma, pp3, type="o", pch=3, col=3)
lines(sigma, pp4, type="o", pch=4, col=4)
legend("topright", legend=paste0(rep("mu=",4),c(97,98,99,100)), bg="white",
col=c(1,2,3,4), pch=c(1,2,3,4), lty=1, cex=0.8)

mu <- seq(95, 105, 0.1)
pp5 <- sapply(X=mu, FUN = PPQ_pp, sigma=0.5, n=10, Llim=95, Ulim=105, k=2.373)
pp6 <- sapply(X=mu, FUN = PPQ_pp, sigma=1, n=10, Llim=95, Ulim=105, k=2.373)
pp7 <- sapply(X=mu, FUN = PPQ_pp, sigma=1.5, n=10, Llim=95, Ulim=105, k=2.373)
pp8 <- sapply(X=mu, FUN = PPQ_pp, sigma=2, n=10, Llim=95, Ulim=105, k=2.373)
pp9 <- sapply(X=mu, FUN = PPQ_pp, sigma=2.5, n=10, Llim=95, Ulim=105, k=2.373)
plot(mu, pp5, xlab="Mean Value", main="LSL=95, USL=105, k=2.373, n=10",
ylab="Probability of Passing", type="o", pch=1, col=1, lwd=1, ylim=c(0,1))
lines(mu, pp6, type="o", pch=2, col=2)
lines(mu, pp7, type="o", pch=3, col=3)
lines(mu, pp8, type="o", pch=4, col=4)
lines(mu, pp9, type="o", pch=5, col=5)
legend("topright", legend=paste0(rep("sigma=",5),seq(0.5,2.5,0.5)), bg="white",
col=c(1,2,3,4,5), pch=c(1,2,3,4,5), lty=1, cex=0.8)

## End(Not run)
```

---

`r1_pp`*Probability of Passing Specification Test for a Release Batch*

---

**Description**

The function for calculating the probability of passing critical quality attributes (CQA) specification test .

**Usage**

```
r1_pp(Llim, Ulim, mu, sigma, NV)
```

**Arguments**

Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
NV	nominal volume for the specification test.

**Value**

A numeric value of the passing/acceptance probability

**Author(s)**

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

**See Also**

PPQ\_pp, pi\_pp and ti\_pp.

**Examples**

```
r1_pp(Llim=1.5, Ulim=3.5, mu=2.5, sigma=0.8)
```

---

ti_ctplot	<i>Heatmap/Contour Plot for Assessing Power of the PPQ Plan using Tolerance Interval.</i>
-----------	---

---

### Description

The function for plotting the heatmap to evaluate the PPQ plan based on the specification test, given lower and upper specification limits.

### Usage

```
ti_ctplot(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, n.batch,
alpha, coverprob, side, test.point)
```

### Arguments

attr.name	user-defined attribute name for PPQ assessment
attr.unit	user-defined attribute unit
Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
alpha	significant level for constructing the tolerance interval.
coverprob	coverage probability for constructing the tolerance interval
side	whether a 1-sided or 2-sided tolerance interval is required (determined by side = 1 or side = 2, respectively).
test.point	(optional) actual process data points for testing whether the processes pass PPQ

### Value

Heatmap (or Contour Plot) for PPQ Assessment.

### Author(s)

Yalin Zhu

### References

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.



**See Also**

ti\_pp and ti\_occurve.

**Examples**

```
## Not run:
mu <- seq(95,105,0.1)
sigma <- seq(0.1,2.5,0.1)
ti_ctplot(attr.name = "Sterile Concentration Assay", attr.unit = "%",
mu = mu, sigma = sigma, Llim=95, Ulim=105)

ti_ctplot(attr.name = "Extractable Volume", attr.unit = "% of NV=1mL",
Llim = 100, Ulim = Inf, mu=seq(100, 110, 0.5), sigma=seq(0.2, 15 ,0.5), n=40,
alpha = 0.05, coverprob = 0.675, side=1)

## End(Not run)
```

---

ti_occurve	<i>Operating Characteristic (OC) Curves for the PPQ Plan using Tolerance Interval.</i>
------------	--

---

**Description**

The function for plotting the OC curve to show the PPQ plan based on the specification test, given lower and upper specification limits.

**Usage**

```
ti_occurve(attr.name, attr.unit, Llim, Ulim, mu, sigma, n, n.batch, alpha,
coverprob, side, add.reference, NV)
```

**Arguments**

attr.name	user-defined attribute name
attr.unit	user-defined attribute unit
Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
alpha	significant level for constructing the tolerance interval.
coverprob	coverage probability for constructing the tolerance interval
side	whether a 1-sided or 2-sided tolerance interval is required (determined by side = 1 or side = 2, respectively).

add.reference logical; if TRUE, then add reference OC curves (Baseline and High Performance) in the plot.

NV nominal volume for the specification test.

**Value**

OC curves for specification test and PPQ plan.

**Author(s)**

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

**See Also**

ti\_pp and rl\_pp.

**Examples**

```
## Not run:
ti_ocurve(attr.name = "Sterile Concentration Assay", attr.unit="",
mu=97, sigma=seq(0.1, 10, 0.1), Llim=95, Ulim=105, n=10, add.reference=TRUE)

ti_ocurve(attr.name = "Sterile Concentration Assay", attr.unit="",
mu=100, sigma=seq(0.1, 10, 0.1), Llim=95, Ulim=105, n=10, add.reference=TRUE)

ti_ocurve(attr.name = "Extractable Volume", attr.unit = "% of NV=3mL",
Llim = 100, Ulim = Inf, mu=102.5, sigma=seq(0.2, 6 ,0.05), n=40,
alpha = 0.05, coverprob = 0.97, side=1, NV=3)

ti_ocurve(attr.name = "Extractable Volume", attr.unit = "% of NV=3mL",
Llim = 100, Ulim = Inf, mu=102.5, sigma=seq(0.2, 6 ,0.05), n=40,
alpha = 0.05, coverprob = 0.992, side=1, NV=3)

## End(Not run)
```

---

ti\_pp

*Probability of Passing PPQ Test using Tolerance Interval*

---

**Description**

The function for calculating the probability of passing critical quality attributes (CQA) PPQ test .

**Usage**

```
ti_pp(Llim, Ulim, mu, sigma, n, n.batch, alpha, coverprob, side)
```

**Arguments**

Llim	lower specification limit
Ulim	upper specification limit
mu	hypothetical mean of the attribute
sigma	hypothetical standard deviation of the attribute
n	sample size (number of locations) per batch
n.batch	number of batches for passing PPQ during validation
alpha	significant level for constructing the tolerance interval
coverprob	coverage probability for constructing the tolerance interval
side	whether a 1-sided or 2-sided tolerance interval is required (determined by side = 1 or side = 2, respectively).

**Value**

A numeric value of the passing/acceptance probability

**Author(s)**

Yalin Zhu

**References**

Burdick, R. K., LeBlond, D. J., Pfahler, L. B., Quiroz, J., Sidor, L., Vukovinsky, K., & Zhang, L. (2017). *Statistical Applications for Chemistry, Manufacturing and Controls (CMC) in the Pharmaceutical Industry*. Springer.

**See Also**

r1\_pp.

**Examples**

```
ti_pp(sigma=0.5, mu=2.5, n=10, n.batch=1, Llim=1.5, Ulim=3.5, alpha=0.05)

sapply(X=c(0.1,0.5, 1,2,3,4,5,10), FUN = ti_pp, mu=97, n=10, Llim=95, Ulim=105,
n.batch=1, alpha=0.05)
sapply(X=c(0.1,0.5, 1,2,3,4,5,10), FUN = ti_pp, mu=100, n=10, Llim=95, Ulim=105,
n.batch=1, alpha=0.05)
```

# Index

heatmap\_ly, 2

k\_factor, 3

pi\_ctplot, 4

pi\_occurve, 5

pi\_pp, 7

pp, 8

PPQ\_ctplot, 9

PPQ\_ggplot, 10

PPQ\_occurve, 12

PPQ\_pp, 13

r1\_pp, 15

ti\_ctplot, 16

ti\_occurve, 17

ti\_pp, 18