

# Package: PRA (via r-universe)

September 4, 2024

**Type** Package

**Title** Project Risk Analysis

**Version** 0.3.0

**Description** Data analysis for Project Risk Management via the Second Moment Method, Monte Carlo Simulation, Contingency Analysis, Sensitivity Analysis, Earned Value Management, Learning Curves, Design Structure Matrices, and more.

**Imports** mc2d, minpack.lm, stats

**License** MIT + file LICENSE

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.3.2

**URL** <https://paulgovan.github.io/PRA/>, <https://github.com/paulgovan/PRA>

**BugReports** <https://github.com/paulgovan/PRA/issues>

**Suggests** ggplot2, knitr, rmarkdown, testthat (>= 3.0.0)

**VignetteBuilder** knitr

**Config/testthat/edition** 3

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

**RemoteUrl** <https://github.com/paulgovan/pr>

**RemoteRef** HEAD

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ac	<i>Actual Cost (AC).</i>
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### Description

Actual Cost (AC).

### Usage

```
ac(actual_costs, time_period)
```

### Arguments

actual_costs	Vector of actual costs incurred at each time period.
time_period	Current time period.

### Value

The function returns the Actual Cost (AC) of work completed.

### Examples

```
# Set the actual costs and current time period for a toy project.
actual_costs <- c(9000, 18000, 36000, 70000, 100000)
time_period <- 3

# Calculate the AC and print the results.
ac <- ac(actual_costs, time_period)
cat("Actual Cost (AC):", ac, "\n")
```

---

contingency	<i>Contingency Calculation.</i>
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**Description**

Contingency Calculation.

**Usage**

```
contingency(sims, phigh = 0.95, pbase = 0.5)
```

**Arguments**

sims	List of results from a Monte Carlo simulation.
phigh	Percentile level for contingency calculation. Default is 0.95.
pbase	Base level for contingency calculation. Default is 0.5

**Value**

The function returns the value of calculated contingency.

**Examples**

```
# Set the number of simulations and the task distributions for a toy project.
num_sims <- 10000
task_dists <- list(
  list(type = "normal", mean = 10, sd = 2), # Task A: Normal distribution
  list(type = "triangular", a = 5, b = 10, c = 15), # Task B: Triangular distribution
  list(type = "uniform", min = 8, max = 12) # Task C: Uniform distribution
)

# Set the correlation matrix for the correlations between tasks.
cor_mat <- matrix(c(
  1, 0.5, 0.3,
  0.5, 1, 0.4,
  0.3, 0.4, 1
), nrow = 3, byrow = TRUE)

# Run the Monte Carlo simulation.
results <- mcs(num_sims, task_dists, cor_mat)

# Calculate the contingency and print the results.
contingency <- contingency(results, phigh = 0.95, pbase = 0.50)
cat("Contingency based on 95th percentile and 50th percentile:", contingency)
```

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cor_matrix	<i>Generate Correlation Matrix.</i>
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**Description**

Generate Correlation Matrix.

**Usage**

```
cor_matrix(num_samples = 100, num_vars = 5, dists = dists)
```

**Arguments**

num_samples	The number of samples to generate.
num_vars	The number of distributions to sample.
dists	A list describing each distribution.

**Value**

The function returns the correlation matrix for the distributions.

**Examples**

```
# List of probability distributions
dists <- list(
  normal = function(n) rnorm(n, mean = 0, sd = 1),
  uniform = function(n) runif(n, min = 0, max = 1),
  exponential = function(n) rexp(n, rate = 1),
  poisson = function(n) rpois(n, lambda = 1),
  binomial = function(n) rbinom(n, size = 10, prob = 0.5)
)

# Generate correlation matrix
cor_matrix <- cor_matrix(num_samples = 100, num_vars = 5, dists = dists)

# Print correlation matrix
print(cor_matrix)
```

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cpi	<i>Cost Performance Index (CPI).</i>
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**Description**

Cost Performance Index (CPI).

**Usage**

```
cpi(ev, ac)
```

**Arguments**

ev	Earned Value.
ac	Actual Cost.

**Value**

The function returns the Cost Performance Index (CPI) of work completed.

**Examples**

```
# Set the BAC and actual % complete for an example project.
bac <- 100000
actual_per_complete <- 0.35

# Calculate the EV
ev <- ev(bac, actual_per_complete)

# Set the actual costs and current time period and calculate the AC.
actual_costs <- c(9000, 18000, 36000, 70000, 100000)
time_period <- 3
ac <- ac(actual_costs, time_period)

# Calculate the CPI and print the results.
cpi <- cpi(ev, ac)
cat("Cost Performance Index (CPI):", cpi, "\n")
```

---

cv

*Cost Variance (CV).*

---

**Description**

Cost Variance (CV).

**Usage**

```
cv(ev, ac)
```

**Arguments**

ev	Earned Value.
ac	Actual Cost.

**Value**

The function returns the Cost Variance (CV) of work completed.

**Examples**

```
# Set the BAC and actual % complete for an example project.
bac <- 100000
actual_per_complete <- 0.35

# Calculate the EV
ev <- ev(bac, actual_per_complete)

# Set the actual costs and current time period and calculate the AC.
actual_costs <- c(9000, 18000, 36000, 70000, 100000)
time_period <- 3
ac <- ac(actual_costs, time_period)

# Calculate the CV and print the results.
cv <- cv(ev, ac)
cat("Cost Variance (CV):", cv, "\n")
```

---

ev

*Earned Value (EV).*


---

**Description**

Earned Value (EV).

**Usage**

```
ev(bac, actual_per_complete)
```

**Arguments**

```
bac          Budget at Completion (BAC) (total planned budget).
actual_per_complete
              Actual work completion percentage.
```

**Value**

The function returns the Earned Value (EV) of work completed.

**Examples**

```
# Set the BAC and actual % complete for a toy project.
bac <- 100000
actual_per_complete <- 0.35

# Calculate the EV and print the results.
ev <- ev(bac, actual_per_complete)
cat("Earned Value (EV):", ev, "\n")
```

---

fit_sigmoidal	<i>Fit a Sigmoidal Model.</i>
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## Description

Fit a Sigmoidal Model.

## Usage

```
fit_sigmoidal(data, x_col, y_col, model_type)
```

## Arguments

data	A data frame containing the time (x_col) and completion (y_col) vectors.
x_col	The name of the time vector.
y_col	The name of the completion vector.
model_type	The name of the sigmoidal model (Pearl, Gompertz, or Logistic).

## Value

The function returns a list of results for the sigmoidal model.

## Examples

```
# Set up a data frame of time and completion percentage data
data <- data.frame(time = 1:10, completion = c(5, 15, 40, 60, 70, 75, 80, 85, 90, 95))

# Fit a logistic model to the data.
fit <- fit_sigmoidal(data, "time", "completion", "logistic")

# Use the model to predict future completion times.
predictions <- predict_sigmoidal(fit, seq(min(data$time), max(data$time),
  length.out = 100), "logistic")

# Plot the results.
p <- ggplot2::ggplot(data, ggplot2::aes_string(x = "time", y = "completion")) +
  ggplot2::geom_point() +
  ggplot2::geom_line(data = predictions, ggplot2::aes(x = x, y = pred), color = "red") +
  ggplot2::labs(title = "Fitted Logistic Model", x = "time", y = "completion %") +
  ggplot2::theme_minimal()
p
```

---

grandparent\_dsm      *Risk-based 'Grandparent' Design Structure Matrix (DSM).*

---

**Description**

Risk-based 'Grandparent' Design Structure Matrix (DSM).

**Usage**

```
grandparent_dsm(S, R)
```

**Arguments**

S                      Resource-Task Matrix 'S' giving the links (arcs) between resources and tasks.  
R                      Risk-Resource Matrix 'R' giving the links (arcs) between risks and resources.

**Value**

The function returns the Risk-based 'Grandparent' DSM 'G' giving the number of risks shared between each task.

**Examples**

```
# Set the S and R matrices and print the results.  
S <- matrix(c(1, 1, 0, 0, 1, 0, 0, 1, 1), nrow = 3, ncol = 3)  
R <- matrix(c(1, 1, 1, 1, 0, 0), nrow = 2, ncol = 3)  
cat("Resource-Task Matrix:\n")  
print(S)  
cat("\nRisk-Resource Matrix:\n")  
print(R)  
# Calculate the Risk-based Grandparent Matrix and print the results.  
risk_dsm <- grandparent_dsm(S, R)  
cat("\nRisk-based 'Grandparent' DSM:\n")  
print(risk_dsm)
```

---

mcs                      *Monte Carlo Simulation.*

---

**Description**

Monte Carlo Simulation.

**Usage**

```
mcs(num_sims, task_dists, cor_mat = NULL)
```



**Arguments**

num\_sims            The number of simulations.  
 task\_dists         A list of lists describing each task distribution.  
 cor\_mat            The correlation matrix for the tasks (Optional).

**Value**

The function returns a list of the total mean, variance, standard deviation, and percentiles for the project.

**Examples**

```
# Set the number of simulations and task distributions for a toy project.
num_sims <- 10000
task_dists <- list(
  list(type = "normal", mean = 10, sd = 2), # Task A: Normal distribution
  list(type = "triangular", a = 5, b = 10, c = 15), # Task B: Triangular distribution
  list(type = "uniform", min = 8, max = 12) # Task C: Uniform distribution
)

# Set the correlation matrix for the correlations between tasks.
cor_mat <- matrix(c(
  1, 0.5, 0.3,
  0.5, 1, 0.4,
  0.3, 0.4, 1
), nrow = 3, byrow = TRUE)

# Run the Monte Carlo simulation and print the results.
results <- mcs(num_sims, task_dists, cor_mat)
cat("Mean Total Duration:", results$total_mean, "\n")
cat("Variance of Total Variance:", results$total_variance, "\n")
cat("Standard Deviation of Total Duration:", results$total_sd, "\n")
cat("5th Percentile:", results$percentiles[1], "\n")
cat("Median (50th Percentile):", results$percentiles[2], "\n")
cat("95th Percentile:", results$percentiles[3], "\n")
hist(results$total_distribution, breaks = 50, main = "Distribution of Total Project Duration",
  xlab = "Total Duration", col = "skyblue", border = "white")
```

---

parent\_dsm

*Resource-based 'Parent' Design Structure Matrix (DSM).*


---

**Description**

Resource-based 'Parent' Design Structure Matrix (DSM).

**Usage**

```
parent_dsm(S)
```

**Arguments**

`S` Resource-Task Matrix 'S' giving the links (arcs) between resources and tasks.

**Value**

The function returns the Resource-based 'Parent' DSM 'P' giving the number of resources shared between each task.

**Examples**

```
# Set the S matrix for a toy project and print the results.
s <- matrix(c(1, 1, 0, 0, 1, 0, 0, 1, 1), nrow = 3, ncol = 3)
cat("Resource-Task Matrix:\n")
print(s)

# Calculate the Resource-based Parent DSM and print the results.
resource_dsm <- parent_dsm(s)
cat("\nResource-based 'Parent' DSM:\n")
print(resource_dsm)
```

---

predict\_sigmoidal      *Predict a Sigmoidal Function.*

---

**Description**

Predict a Sigmoidal Function.

**Usage**

```
predict_sigmoidal(fit, x_range, model_type)
```

**Arguments**

`fit`                    A list containing the results of a sigmoidal model.  
`x_range`                A vector of time values for the prediction.  
`model_type`            The type of model (Pearl, Gompertz, or Logistic) for the prediction.

**Value**

The function returns a table of results containing the time and predicted values.

**Examples**

```
# Set up a data frame of time and completion percentage data
data <- data.frame(time = 1:10, completion = c(5, 15, 40, 60, 70, 75, 80, 85, 90, 95))

# Fit a logistic model to the data.
fit <- fit_sigmoidal(data, "time", "completion", "logistic")

# Use the model to predict future completion times.
predictions <- predict_sigmoidal(fit, seq(min(data$time), max(data$time),
  length.out = 100), "logistic")

# Plot the results.
p <- ggplot2::ggplot(data, ggplot2::aes_string(x = "time", y = "completion")) +
  ggplot2::geom_point() +
  ggplot2::geom_line(data = predictions, ggplot2::aes(x = x, y = pred), color = "red") +
  ggplot2::labs(title = "Fitted Logistic Model", x = "time", y = "completion %") +
  ggplot2::theme_minimal()
p
```

pv

*Planned Value (PV).***Description**

Planned Value (PV).

**Usage**

```
pv(bac, schedule, time_period)
```

**Arguments**

bac	Budget at Completion (BAC) (total planned budget).
schedule	Vector of planned work completion (in terms of percentage) at each time period.
time_period	Current time period.

**Value**

The function returns the Planned Value (PV) of work completed.

**Examples**

```
# Set the BAC, schedule, and current time period for a toy project.
bac <- 100000
schedule <- c(0.1, 0.2, 0.4, 0.7, 1.0)
time_period <- 3

# Calculate the PV and print the results.
pv <- pv(bac, schedule, time_period)
cat("Planned Value (PV):", pv, "\n")
```

---

sensitivity

*Sensitivity Analysis.*


---

**Description**

Sensitivity Analysis.

**Usage**

```
sensitivity(task_dists, cor_mat = NULL)
```

**Arguments**

`task_dists`      A list of lists describing each task distribution.  
`cor_mat`          The correlation matrix for the tasks (Optional).

**Value**

The function returns a vector of sensitivity results with respect to each task.

**Examples**

```
# Set the task distributions for a toy project.
task_dists <- list(
  list(type = "normal", mean = 10, sd = 2), # Task A: Normal distribution
  list(type = "triangular", a = 5, b = 15, c = 10), # Task B: Triangular distribution
  list(type = "uniform", min = 8, max = 12) # Task C: Uniform distribution
)

# Set the correlation matrix between the tasks.
cor_mat <- matrix(c(
  1, 0.5, 0.3,
  0.5, 1, 0.4,
  0.3, 0.4, 1
), nrow = 3, byrow = TRUE)

# Calculate the sensitivity of each task and print the results
sensitivity_results <- sensitivity(task_dists, cor_mat)
cat("Sensitivity of the variance in total cost with respect to the variance in each task cost:\n")
print(sensitivity_results)

# Build a vertical barchart and display the results.
data <- data.frame(
  Tasks = c('A', 'B', 'C'),
  Sensitivity = sensitivity_results
)
barplot(height=data$Sensitivity, names=data$Tasks, col='skyblue',
        horiz=TRUE, xlab = 'Sensitivity', ylab = 'Tasks')
```

---

smm                      *Second Moment Analysis.*

---

**Description**

Second Moment Analysis.

**Usage**

```
smm(mean, var, cor_mat = NULL)
```

**Arguments**

mean	The mean vector.
var	The variance vector.
cor_mat	The correlation matrix (optional).

**Value**

The function returns a list of the total mean, variance, and standard deviation for the project.

**Examples**

```
# Set the mean vector, variance vector, and correlation matrix for a toy project.
mean <- c(10, 15, 20)
var <- c(4, 9, 16)
cor_mat <- matrix(c(
  1, 0.5, 0.3,
  0.5, 1, 0.4,
  0.3, 0.4, 1
), nrow = 3, byrow = TRUE)

# Use the Second Moment Method to estimate the results for the project.
result <- smm(mean, var, cor_mat)
print(result)
```

---

spi                      *Schedule Performance Index (SPI).*

---

**Description**

Schedule Performance Index (SPI).

**Usage**

```
spi(ev, pv)
```

**Arguments**

ev                Earned Value.  
pv                Planned Value.

**Value**

The function returns the Schedule Performance Index (SPI) of work completed.

**Examples**

```
# Set the BAC, schedule, and current time period for an example project.
bac <- 100000
schedule <- c(0.1, 0.2, 0.4, 0.7, 1.0)
time_period <- 3

# Calculate the PV.
pv <- pv(bac, schedule, time_period)

# Set the actual % complete and calculate the EV.
actual_per_complete <- 0.35
ev <- ev(bac, actual_per_complete)

# Calculate the SPI and print the results.
spi <- spi(ev, pv)
cat("Schedule Performance Index (SPI):", spi, "\n")
```

---

sv                                *Schedule Variance (SV).*

---

**Description**

Schedule Variance (SV).

**Usage**

```
sv(ev, pv)
```

**Arguments**

ev                Earned Value.  
pv                Planned Value.

**Value**

The function returns the Schedule Variance (SV) of work completed.

**Examples**

```
# Set the BAC, schedule, and current time period for an example project.
bac <- 100000
schedule <- c(0.1, 0.2, 0.4, 0.7, 1.0)
time_period <- 3

# Calculate the PV.
pv <- pv(bac, schedule, time_period)

# Set the actual % complete and calculate the EV.
actual_per_complete <- 0.35
ev <- ev(bac, actual_per_complete)

# Calculate the SV and print the results.
sv <- sv(ev, pv)
cat("Schedule Variance (SV):", sv, "\n")
```

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