

# Package: CRBHSF (via r-universe)

June 25, 2026

**Type** Package

**Title** Cyber-Resilient Bayesian Healthcare Surveillance Framework

**Version** 0.1.0

**Description** Provides methods for cyber-resilient healthcare performance surveillance using Bayesian risk estimation, latent organisational trust modelling, cyber-resilience assessment, external validation, decision-theoretic optimisation, and digital-twin deployment simulation. The package supports prospective deterioration monitoring, uncertainty-aware risk assessment, intervention prioritisation, ablation analysis, and operational evaluation for healthcare performance management and health system resilience research. The methodological framework is informed by contemporary guidance on prediction model development and validation (Efthimiou et al., 2024 <[doi:10.1136/bmj-2023-078276](https://doi.org/10.1136/bmj-2023-078276)>), transparent reporting of prediction models (Collins et al., 2024 <[doi:10.1136/bmj-2023-078378](https://doi.org/10.1136/bmj-2023-078378)>), and decision-analytic model evaluation (Vickers and Elkin, 2006 <[doi:10.1177/0272989X06295361](https://doi.org/10.1177/0272989X06295361)>).

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**Encoding** UTF-8

**LazyData** true

**Roxygen** list(markdown = TRUE)

**RoxygenNote** 7.3.3

**URL** <https://github.com/zerish12/CRBHSF>

**BugReports** <https://github.com/zerish12/CRBHSF/issues>

**Imports** dplyr, ggplot2, pROC, purrr, tibble

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

**Config/testthat/edition** 3

**Repository** <https://zerish12.r-universe.dev>

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clean_health_data	<i>Clean healthcare surveillance data</i>
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## Description

Clean healthcare surveillance data

## Usage

```
clean_health_data(data, provider_col, time_col)
```

## Arguments

data	A data frame.
provider_col	Character string giving the column name.
time_col	Character string giving the column name.

---

compare\_ml\_benchmarks *Compare simple machine-learning benchmarks*

---

**Description**

Compare simple machine-learning benchmarks

**Usage**

```
compare_ml_benchmarks(data, outcome_col, predictor_cols)
```

**Arguments**

data            A data frame.  
outcome\_col    Character string giving the column name.  
predictor\_cols Character vector of predictor column names.

---

compute\_crbhsf\_risk *Compute CRBHSF risk score*

---

**Description**

Compute CRBHSF risk score

**Usage**

```
compute_crbhsf_risk(  
  data,  
  bayes_risk_col = "risk_bayes",  
  trust_col = "latent_trust",  
  lambda = 0.2  
)
```

**Arguments**

data            A data frame.  
bayes\_risk\_col Character string giving the column name.  
trust\_col       Character string giving the column name.  
lambda          Numeric parameter.

---

compute_crpr	<i>Compute Cyber-Resilience Pressure Ratio</i>
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**Description**

Compute Cyber-Resilience Pressure Ratio

**Usage**

```
compute_crpr(data, bayes_risk_col = "risk_bayes", trust_col = "latent_trust")
```

**Arguments**

data	A data frame.
bayes_risk_col	Character string giving the column name.
trust_col	Character string giving the column name.

---

create_deterioration_outcome	<i>Create future deterioration outcome</i>
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**Description**

Create future deterioration outcome

**Usage**

```
create_deterioration_outcome(  
  data,  
  provider_col,  
  time_col,  
  value_col,  
  threshold = 1  
)
```

**Arguments**

data	A data frame.
provider_col	Character string giving the column name.
time_col	Character string giving the column name.
value_col	Character string giving the column name.
threshold	Numeric parameter.

---

estimate_evib	<i>Estimate expected value of intervention benefit</i>
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**Description**

Estimate expected value of intervention benefit

**Usage**

```
estimate_evib(  
  data,  
  risk_col = "risk_crbhsf",  
  review_cost = 1,  
  miss_cost = 10,  
  threshold = 0.5  
)
```

**Arguments**

data	A data frame.
risk_col	Character string giving the column name.
review_cost	Numeric parameter.
miss_cost	Numeric parameter.
threshold	Numeric parameter.

---

estimate_latent_trust	<i>Estimate latent organisational trust</i>
-----------------------	---

---

**Description**

Estimate latent organisational trust

**Usage**

```
estimate_latent_trust(  
  data,  
  anomaly_col = NULL,  
  corruption_col = NULL,  
  cyber_col = NULL,  
  missing_col = NULL,  
  delay_col = NULL  
)
```

**Arguments**

data	A data frame.
anomaly_col	Character string giving the column name.
corruption_col	Character string giving the column name.
cyber_col	Character string giving the column name.
missing_col	Character string giving the column name.
delay_col	Character string giving the column name.

---

fit\_bayesian\_surveillance

*Fit lightweight Bayesian surveillance model*

---

**Description**

Fit lightweight Bayesian surveillance model

**Usage**

```
fit_bayesian_surveillance(data, y_col, n_col, alpha0 = 2, beta0 = 2)
```

**Arguments**

data	A data frame.
y_col	Character string giving the column name.
n_col	Character string giving the column name.
alpha0	Numeric parameter.
beta0	Numeric parameter.

---

generate\_surveillance\_report

*Generate simple surveillance report*

---

**Description**

Generate simple surveillance report

**Usage**

```
generate_surveillance_report(validation_results, deployment_results = NULL)
```

**Arguments**

validation_results	Validation results returned by validate_surveillance().
deployment_results	Optional deployment results returned by simulate_digital_twin().

---

`plot_ablation_auc`      *Plot ablation AUC*

---

### **Description**

Plot ablation AUC

### **Usage**

```
plot_ablation_auc(ablation_results)
```

### **Arguments**

`ablation_results`  
Function argument.

---

`plot_deployment_impact`  
*Plot digital-twin deployment impact*

---

### **Description**

Plot digital-twin deployment impact

### **Usage**

```
plot_deployment_impact(deployment_results)
```

### **Arguments**

`deployment_results`  
Optional deployment results returned by `simulate_digital_twin()`.

---

plot\_risk\_distribution  
*Plot risk distribution*

---

**Description**

Plot risk distribution

**Usage**

```
plot_risk_distribution(data, risk_col = "risk_crbhsf")
```

**Arguments**

data	A data frame.
risk_col	Character string giving the column name.

---

run\_ablation\_study     *Run CRBHSF ablation study*

---

**Description**

Evaluates the incremental predictive contribution of multiple surveillance risk scores by calculating validation metrics for each score against a common binary outcome.

**Usage**

```
run_ablation_study(
  data,
  outcome_col,
  score_cols = c("risk_bayes", "latent_trust", "crpr", "risk_crbhsf")
)
```

**Arguments**

data	A data frame containing the outcome and score columns.
outcome_col	Character string giving the binary outcome column.
score_cols	Character vector giving the surveillance score columns to evaluate. Defaults to risk_bayes, latent_trust, crpr, and risk_crbhsf.

**Value**

A tibble containing validation metrics for each score column, including sample size, event count, event rate, AUC, and Brier score.

---

simulate\_digital\_twin *Simulate digital-twin deployment*

---

**Description**

Simulate digital-twin deployment

**Usage**

```
simulate_digital_twin(  
  data,  
  outcome_col,  
  score_col,  
  capacity = 0.1,  
  intervention_effect = 0.6,  
  miss_cost = 15,  
  review_cost = 1  
)
```

**Arguments**

data	A data frame containing outcome and score columns.
outcome_col	Character string giving the binary deterioration outcome column.
score_col	Character string giving the risk score column used for prioritisation.
capacity	Proportion of observations selected for review.
intervention_effect	Proportion of reviewed deterioration events assumed preventable.
miss_cost	Cost assigned to a missed deterioration event.
review_cost	Cost assigned to each review.

**Value**

A tibble summarising reviews, prevented deteriorations, losses, and loss reduction.

---

validate\_surveillance *Validate surveillance score*

---

**Description**

Validate surveillance score

**Usage**

```
validate_surveillance(data, outcome_col, score_col)
```

**Arguments**

<code>data</code>	A data frame containing the outcome and score columns.
<code>outcome_col</code>	Character string giving the binary outcome column.
<code>score_col</code>	Character string giving the risk score column.

**Value**

A tibble containing sample size, event count, event rate, AUC, and Brier score.

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