

# ‘RLumCarlo’: Tedious features - fine examples

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## 1 Scope

‘RLumCarlo’ is a collection of energy-band models to simulate luminescence signals in dosimetric materials using Monte-Carlo (MC) methods for various stimulation modes. This document aims at supplementing the package documentation and elaborating the package examples.

## 2 The models in ‘RLumCarlo’

### 2.1 Overview

TRANSITION	BASE MODEL	IRSL	OSL	LM-OSL	TL
Delocalised	OTOR	-	X	X	X
Localised	GOT	X	-	X	X
Excited state tunnelling	LTM	X	-	X	X

In the table above column headers refer to stimulation modes, which are infrared stimulated luminescence (IRSL), optically stimulated luminescence (OSL), LM-OSL (Bulur 1996), and thermally stimulated luminescence (short: TL). In the column ‘BASE MODEL’ OTOR refers to ‘One Trap-One Recombination Centre’, GOT to ‘General One Trap’, and LTM to ‘Localized Transition Model’ (Jain, Guralnik, and Andersen 2012; Pagonis et al. 2019). For general overview we refer to the excellent book by Chen and Pagonis (2011).

### 2.2 Where to find them

The following table lists models as implemented in ‘RLumCarlo’ along with the **R** function call and the corresponding R (\*.R) and C++ (\*.cpp) files. The modelling takes place in the C++ functions which are wrapped by the R functions with a similar name. If you, however, want to cross-check the code, you should inspect files with the ending `.cpp`.

MODEL_NAME	R_CALL	CORRESPONDING_FILES
MC_CW_IRSL_LOC	run_MC_CW_IRSL_LOC()	R/run_MC_CW_IRSL_LOC.R src/MC_C_MC_CW_IRSL_LOC.cpp
MC_CW_IRSL_TUN	run_MC_CW_IRSL_TUN()	R/run_MC_CW_IRSL_TUN.R src/MC_C_MC_CW_IRSL_TUN.cpp
MC_CW_OSL_DELOC	run_MC_CW_OSL_DELOC()	R/run_MC_CW_OSL_DELOC.R src/MC_C_MC_CW_OSL_DELOC.cpp
MC_ISO_DELOC	run_MC_ISO_DELOC()	R/run_MC_ISO_DELOC.R src/MC_C_MC_ISO_DELOC.cpp
MC_ISO_LOC	run_MC_ISO_LOC()	R/run_MC_ISO_LOC.R src/MC_C_MC_ISO_LOC.cpp
MC_ISO_TUN	run_MC_ISO_TUN()	R/run_MC_ISO_TUN.R src/MC_C_MC_ISO_TUN.cpp
MC_LM_OSL_DELOC	run_MC_LM_OSL_DELOC()	R/run_MC_LM_OSL_DELOC.R src/MC_C_MC_LM_OSL_DELOC.cpp
MC_LM_OSL_LOC	run_MC_LM_OSL_LOC()	R/run_MC_LM_OSL_LOC.R src/MC_C_MC_LM_OSL_LOC.cpp
MC_LM_OSL_TUN	run_MC_LM_OSL_TUN()	R/run_MC_LM_OSL_TUN.R src/MC_C_MC_LM_OSL_TUN.cpp
MC_TL_DELOC	run_MC_TL_DELOC()	R/run_MC_TL_DELOC.R src/MC_C_MC_TL_DELOC.cpp
MC_TL_LOC	run_MC_TL_LOC()	R/run_MC_TL_LOC.R src/MC_C_MC_TL_LOC.cpp
MC_TL_TUN	run_MC_TL_TUN()	R/run_MC_TL_TUN.R src/MC_C_MC_TL_TUN.cpp

Each model is run by calling one of the **R** functions starting with **run\_**. Currently, three different model types (TUN: tunnelling, LOC: localised transition, DELOC: delocalised transition) are implemented for the stimulation types TL, IRSL, LM-OSL, and ISO (isothermal). Please note that each model has different parameters and requirements.

### 3 ‘RLumCarlo’ model parameters and variables

The following table summarises the parameters used in the implemented MC models along with their physical meaning, units and the range of realistic values. This range represents just a rough guideline and might be exceeded for particular cases.

Stimulation mode	Parameter	Parameter description	Unit	Realistic values
<b>Delocalized TL</b>	E	Thermal activation energy of the trap	eV	0.5–3
	s	Frequency factor of the trap	1/s	1E8–1E16
	times	Sequence of time steps for simulation (heating rate is 1 K/s)	s	0–700
	clusters	Number of MC runs	1	1E1–1E4
	N <sub>e</sub>	Total number of electron traps available	1	2–1E5
	n <sub>filled</sub>	Number of filled electron traps at the beginning of the simulation	1	1–1E5
	R	Delocalized retrapping ratio	1	0–1
<b>Delocalized CW-IRSL</b>	A	Optical excitation rate from trap to conduction band	1/s	1E–3–1
	times	Sequence of time steps for simulation	s	0–500
	clusters	Number of MC runs	1	1E1–1E4

<b>Delocalized ISO</b>	N_e	Total number of electron traps available	1	2–1E5
	n_filled	Number of filled electron traps at the beginning of the simulation	1	1–1E5
	R	Delocalized retrapping ratio	1	0–1
	E	Thermal activation energy of the trap	eV	0.5–3
	s	Frequency factor of the trap	1/s	1E8–1E16
	T	Temperature of the isothermal process	°C	20–300
	times	Sequence of time steps for simulation	s	0–1000
	clusters	Number of MC runs	1	1E1–1E4
<b>Delocalized LM-OSL</b>	N_e	Number of electrons	1	2–1E5
	n_filled	Number of filled electron traps at the beginning of the simulation	1	1–1E5
	R	Delocalized retrapping ratio	1	0–1
	A	Optical excitation rate from trap to conduction band	1/s	1E–3–1
	times	Sequence of time steps for simulation	s	0–3000
	clusters	Number of MC runs	1	1E1–1E4
	N_e	Total number of electron traps available	1	2–1E5
	n_filled	Number of filled electron traps at the beginning of the simulation	1	1–1E5
<b>Localized TL</b>	R	Delocalized retrapping ratio	1	0–1
	E	Thermal activation energy of the trap	eV	0.5–3
	s	Frequency factor of the trap	1/s	1E8–1E16
	times	Sequence of time steps for simulation (heating rate 1 K/s)	s	0–700
	clusters	Number of MC runs	1	1E1–1E4
	n_filled	Number of filled electron traps at the beginning of the simulation	1	1–1E5
	r	Localized retrapping ratio	1	0–1E5
	A	Optical excitation rate from ground state of the trap to the excited state	1/s	1E–3–1
<b>Localized CW-IRSL</b>	times	Sequence of time steps for simulation	s	0–500
	clusters	Number of MC runs	1	1E1–1E4
	n_filled	Number of filled electron traps at the beginning of the simulation	1	1–1E5
	r	Localized retrapping ratio	1	0–1E5
	E	Thermal activation energy of the trap	eV	0.5–3
	s	Frequency factor of the trap	1/s	1E8–1E16
	T	Temperature of the isothermal process	°C	20–300
	times	Sequence of time steps for simulation	s	0–1000
<b>Localized ISO</b>	clusters	Number of MC runs	1	1E1–1E4
	n_filled	Number of filled electron traps at the beginning of the simulation	1	1–1E5
	r	Localized retrapping ratio	1	0–1E5
	A	Optical excitation rate from ground state of the trap to the excited state	1/s	1E–3–1
	times	Sequence of time steps for simulation	s	0–3000

	clusters	Number of MC runs	1	1E1–1E4
	n_filled	Number of filled electron traps at the beginning of the simulation	1	1–1E5
	r	Localized retrapping ratio	1	0–1E5
<b>TL with tunneling recombination</b>	E	Thermal activation energy of the trap	eV	0.5–3
	s	Effective frequency factor of the tunneling process	1/s	1E8–1E16
	rho	Dimensionless density of recombination centers (defined as $\rho'$ in Huntley 2006)	1	1E-7–1E-4
	r_c	Critical distance (>0) that is to be inserted if the sample has been thermally and/or optically pretreated, so that the electron-hole pairs within r_c have already recombined	1	0–2
	times	Sequence of time steps for simulation (heating rate 1 K/s)	s	0–700
	clusters	Number of MC runs	1	1E1–1E4
	N_e	Total number of electron traps available	1	2–1E5
<b>CW-IRSL with tunneling recombination</b>	delta.r	Increments of the unitless distance parameter $r'$	1	1E-3–1E-1
	A	Effective optical excitation rate of the tunneling process	1/s	1E-3–1
	rho	Dimensionless density of recombination centers (defined as $\rho'$ in Huntley 2006)	1	1E-7–1E-4
	times	Sequence of time steps for simulation	s	0–500
	clusters	Number of MC runs	1	1E1–1E4
	N_e	Total number of electron traps available	1	2–1E5
	r_c	Critical distance (>0) that is to be inserted if the sample has been thermally and/or optically pretreated, so that the electron-hole pairs within r_c have already recombined	1	0–2
<b>ISO with tunneling recombination</b>	delta.r	Increments of the unitless distance parameter $r'$	1	1E-3–1E-1
	E	Thermal activation energy of the trap	eV	0.5–3
	s	Effective frequency factor of the tunneling process	1/s	1E8–1E16
	T	Temperature of the isothermal process	°C	20–300
	rho	Dimensionless density of recombination centers (defined as $\rho'$ in Huntley 2006)	1	1E-7–1E-4
	times	Sequence of time steps for simulation	s	0–1000
	clusters	Number of MC runs	1	1E1–1E4
<b>LM-OSL with tunneling recombination</b>	N_e	Total number of electron traps available	1	2–1E5
	r_c	Critical distance (>0) that is to be inserted if the sample has been thermally and/or optically pretreated, so that the electron-hole pairs within r_c have already recombined	1	0–2
	delta.r	Increments of the unitless distance parameter $r'$	1	1E-3–1E-1
	A	Effective optical excitation rate of the tunneling process	1/s	1E-3–1
	rho	Dimensionless density of recombination centers (defined as $\rho'$ in Huntley 2006)	1	1E-7–1E-4
	times	Sequence of time steps for simulation	s	0–3000
	clusters	Number of MC runs	1	1E1–1E4

N_e	Total number of electron traps available	1	2-1E5
r_c	Critical distance (>0) that is to be inserted if the sample has been thermally and/or optically pretreated, so that the electron-hole pairs within r_c have already recombined	1	0-2
delta.r	Increments of the unitless distance parameter r'	1	1E-3-1E-1

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## 4 Examples

The following examples illustrate the capacity of ‘RLumCarlo’, by using code-snippets deploying longer simulation times than allowed for the standard package examples, which aim at a functionality test.

### 4.1 Example 1: A first example

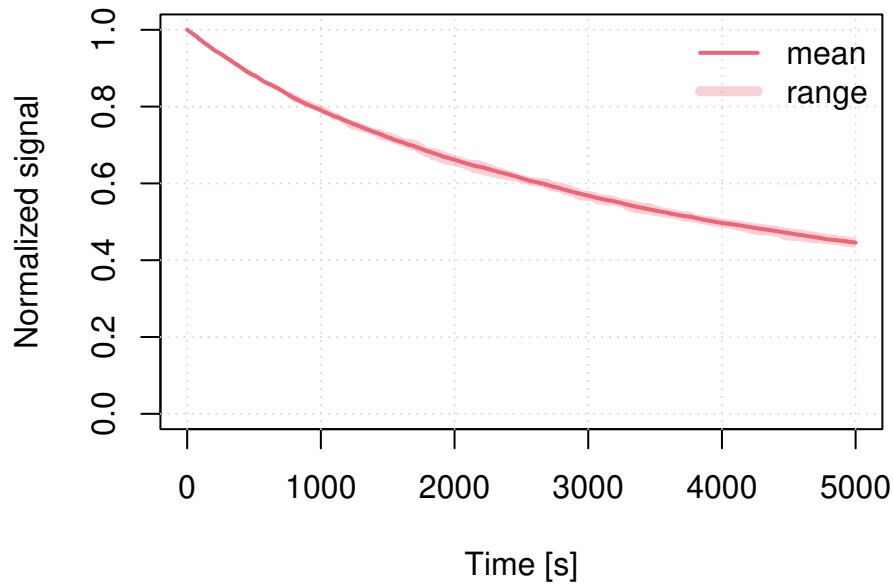
The first example is an iso-thermal decay curve using the tunnelling model (other models work similarly). Returned are either the simulated signal or the estimated remaining trapped charge carriers. The Function `plot_RLumCarlo()` provides an easy way to visualise the modelling results and is here called using the tee operator `%T>%` from the package `magrittr` (which is imported by ‘RLumCarlo’). Simulation results are stored in the object `results` while, at the same time, piped to the function `plot_RLumCarlo()` for the output visualisation.

#### 4.1.1 Model the signal

The most obvious modelling output is the luminescence signal itself, our example below simulates an iso-thermal (ITL) signal for a temperature (T) of 200 °C over 5,000 s using a tunnelling transition model. Trap parameters are  $E = 1.2$  eV for the trap depth and a frequency factor of  $1 \times 10^{10}$  (1/s). The parameter `rho` ( $\rho'$ ) defines the recombination centre density.

```
results <- run_MC_ISO_TUN(
  E = 1.2,
  s = 1e10,
  T = 200,
  N_e = 200,
  rho = 0.007,
  clusters = 10,
  times = seq(0, 5000)
) %T>%
plot_RLumCarlo(norm = TRUE,
               legend = TRUE,
               main = "Iso-thermal decay (TUN)")
```

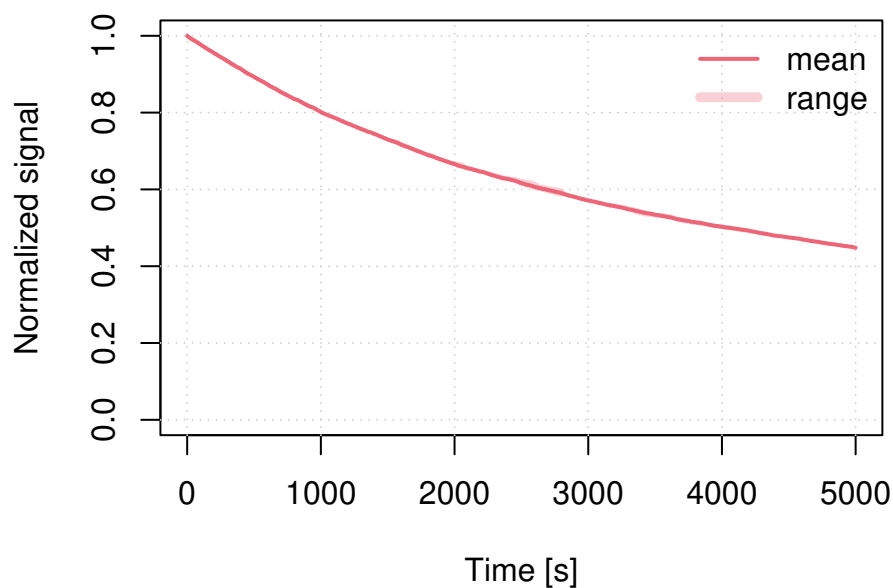
## Iso-thermal decay (TUN)



In the example above  $N_e$  is a scalar, which means that all clusters start with the same number of electrons (here 200). However, 'RLumCarlo' supports different starting conditions with regard to the initial number of electrons. For example, one could assume that the number of initial electrons vary randomly between 190 and 210. Such a situation is created in the next example. Generally, 'RLumCarlo' supports such an input for the parameters  $N_e$  and  $n_{\text{filled}}$ .

```
results <- run_MC_ISO_TUN(  
  E = 1.2,  
  s = 1e10,  
  T = 200,  
  N_e = sample(190:210,10,TRUE),  
  rho = 0.007,  
  clusters = 10,  
  times = seq(0, 5000)  
) %T>%  
plot_RLumCarlo(norm = TRUE,  
               legend = TRUE,  
               main = "Iso-thermal decay (TUN) for varying N_e")
```

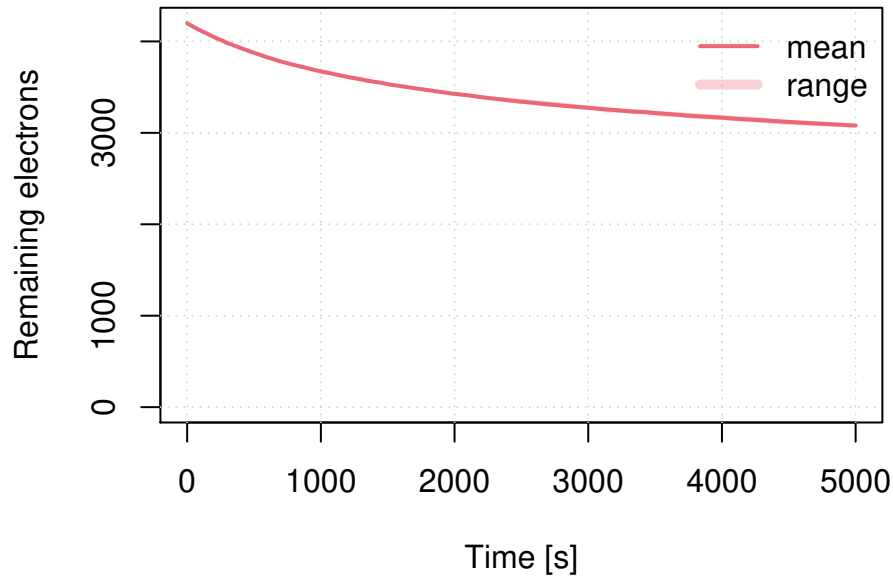
### Iso-thermal decay (TUN) for varying N<sub>e</sub>



#### 4.1.2 Model remaining charges

The first example can be slightly altered to provide alternative insight. Instead of the luminescence signal, the variant below returns the number of remaining electrons in the trap.

```
results <- run_MC_ISO_TUN(  
  E = 1.2,  
  s = 1e10,  
  T = 200,  
  rho = 0.007,  
  times = seq(0, 5000),  
  output = "remaining_e"  
) %T>%  
plot_RLumCarlo(  
  legend = TRUE,  
  ylab = "Remaining electrons"  
)
```



#### 4.1.3 Understanding the numerical output

In both cases the modelling output is an object of class `RLumCarlo_Model_Output`, which is basically a list consisting of an array and a numeric (vector).

```
str(results)
```

```
## List of 2
## $ signal: num [1:5001, 1:21, 1:10] 200 200 199 198 198 198 198 197 196 196 ...
##   ..- attr(*, "dimnames")=List of 3
##     .. ..$ : NULL
##     .. ..$ : NULL
##     .. ..$ : NULL
## $ time   : int [1:5001] 0 1 2 3 4 5 6 7 8 9 ...
## - attr(*, "class")= chr "RLumCarlo_Model_Output"
## - attr(*, "model")= chr "run_MC_ISO_TUN"
```

While this represents the full modelling output results, its interpretation might be less straight forward, and the user may want to condense the information via `summary()`. The function `summary()` is also used internally by the function `plot_RLumCarlo()` to simplify the data before there are plotted.

```
df <- summary(results)
```

```
##      time      mean      y_min      y_max      sd
## Min.   : 0      Min.   :3080    Min.   :3060    Min.   :3100    Min.   : 0.7379
## 1st Qu.:1250    1st Qu.:3188    1st Qu.:3159    1st Qu.:3219    1st Qu.:10.3392
## Median :2500    Median :3341    Median :3320    Median :3370    Median :14.0637
## Mean   :2500    Mean   :3423    Mean   :3402    Mean   :3445    Mean   :13.7298
## 3rd Qu.:3750    3rd Qu.:3598    3rd Qu.:3581    3rd Qu.:3619    3rd Qu.:16.9509
## Max.   :5000    Max.   :4199    Max.   :4198    Max.   :4200    Max.   :20.9149
##      var      sum
## Min.   : 0.5444    Min.   :30800
## 1st Qu.:106.9000    1st Qu.:31880
## Median :197.7889    Median :33410
## Mean   :205.2451    Mean   :34231
## 3rd Qu.:287.3333    3rd Qu.:35983
## Max.   :437.4333    Max.   :41991
```

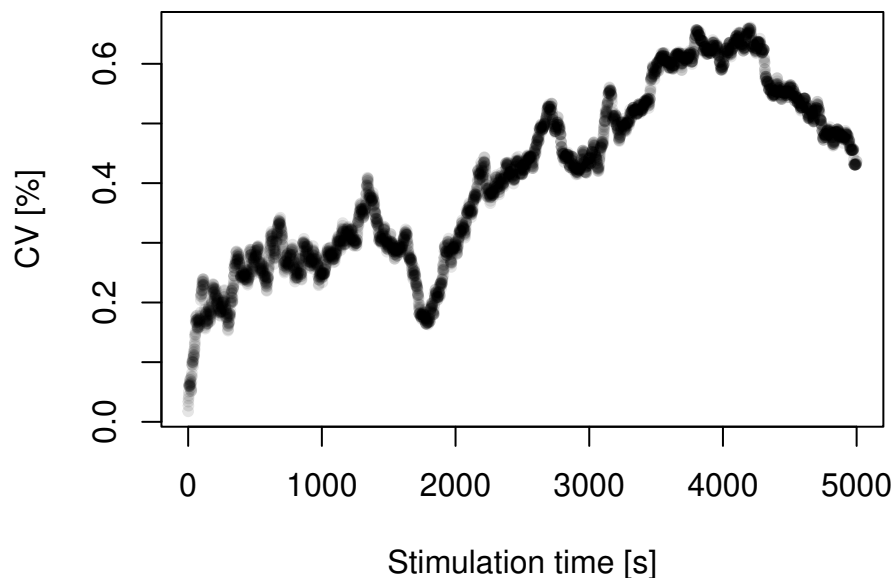


```
head(df)
```

```
##   time  mean y_min y_max      sd      var  sum
## 1    0 4199.1  4198  4200 0.7378648 0.5444444 41991
## 2    1 4197.7  4195  4200 1.4181365 2.0111111 41977
## 3    2 4196.8  4195  4198 1.1352924 1.2888889 41968
## 4    3 4195.3  4193  4197 1.6363917 2.6777778 41953
## 5    4 4194.6  4191  4197 1.8378732 3.3777778 41946
## 6    5 4193.6  4190  4197 2.1186998 4.4888889 41936
```

The call summarises the modelling results and returns a terminal output and a `data.frame` with, e.g., the mean or the standard deviation, which can be used to create plots for further insight. For instance, the stimulation time against coefficient of variation (CV in %):

```
plot(
  x = df$time,
  y = (df$sd / df$mean) * 100,
  pch = 20,
  col = rgb(0,0,0,.1),
  xlab = "Stimulation time [s]",
  ylab = "CV [%]"
)
```



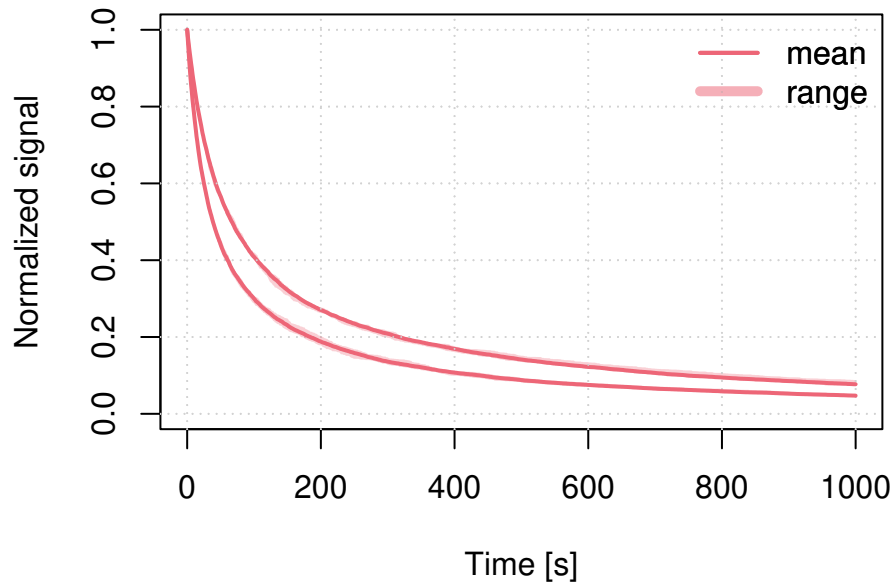
## 4.2 Example 2: Combining two plots

The following examples use again the tunnelling model but for continuous wave (CW) infrared light stimulation (IRSL), and they combine two plots in one single plot window.

```
## set time vector
times <- seq(0, 1000)

## Run MC simulation
run_MC_CW_IRSL_TUN(A = 0.12, rho = 0.003, times = times) %>%
  plot_RLumCarlo(norm = TRUE, legend = TRUE)

run_MC_CW_IRSL_TUN(A = 0.21, rho = 0.003, times = times) %>%
  plot_RLumCarlo(norm = TRUE, add = TRUE)
```



### 4.3 Example 3: Testing different parameters

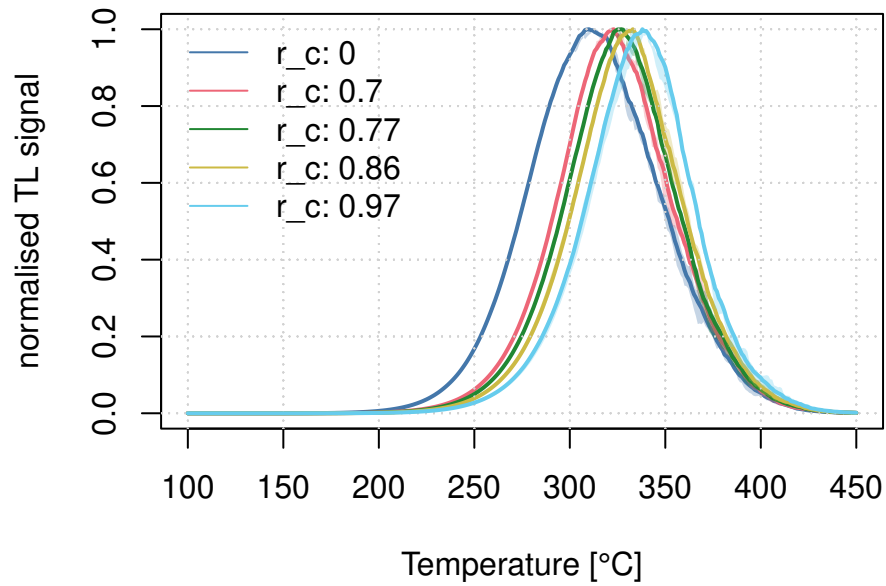
The example above can be further extended to test the effect of different parameters. Contrary to the example above, here the results are stored in a list and `plot_RLumCarlo()` is called only one time and it will then iterate automatically over the results to create a combined plot.

```
s <- 3.5e12
rho <- 0.015
E <- 1.45
r_c <- c(0,0.7,0.77,0.86, 0.97)
times <- seq(100, 450) # here time = temperature
results <- lapply(r_c, function(x) {
  run_MC_TL_TUN(
    s = s,
    E = E,
    rho = rho,
    r_c = x,
    times = times
  )
})
```

The plot output can be highly customised to provide a better visual experience, e.g., the manual setting of the colours and the legend.

```
## plot curves, but without legend
plot_RLumCarlo(
  object = results,
  ylab = "normalised TL signal",
  xlab = "Temperature [\u00b0C]",
  plot_uncertainty = "range",
  col = khroma::colour("bright")(length(r_c)),
  legend = FALSE,
  norm = TRUE
)
```

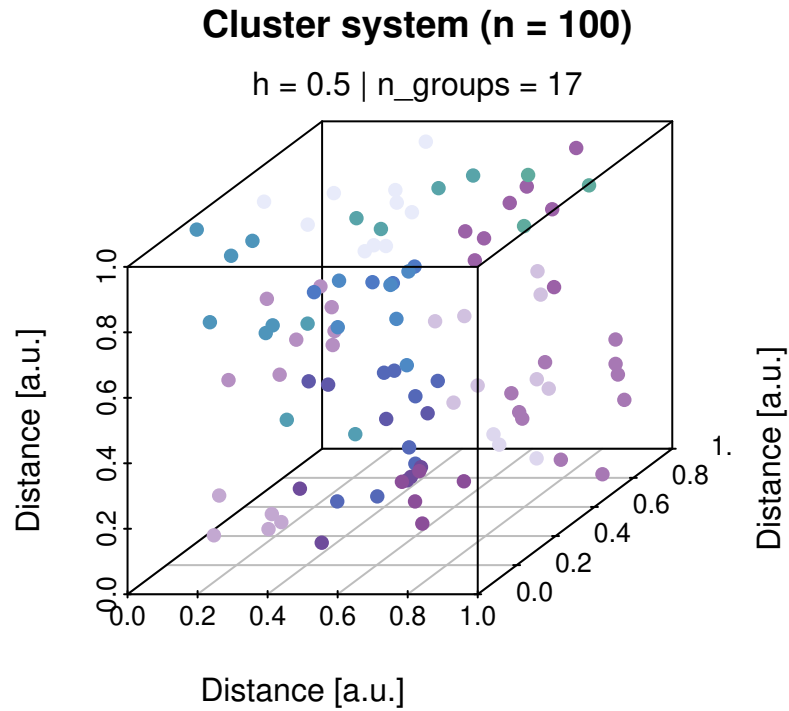
```
## add legend manually
legend(
  "topleft",
  bty = "n",
  legend = paste0("r_c: ", r_c),
  lty = 1,
  col = khroma::colour("bright")(length(r_c))
)
```



#### 4.4 Example 4: Dosimetric cluster systems

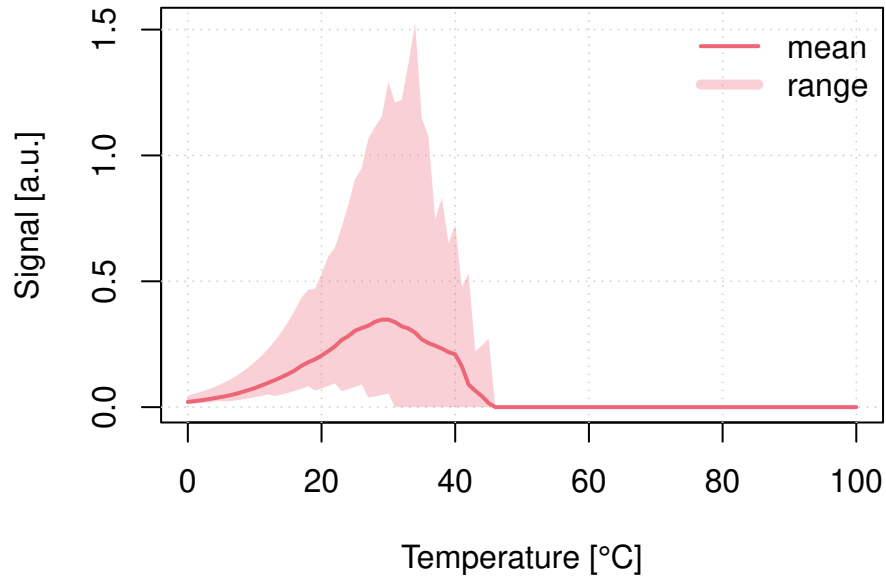
'RLumCarlo' supports the simulation of a cheap dosimetric cluster system with spatial correlation. Such a dosimetric cluster system can be created with the function `create_ClusterSystem()`:

```
clusters <- create_ClusterSystem(n = 100, plot = TRUE)
```



The result is an arbitrary dosimetric system with randomly distributed clusters. The Euclidean distance is used to group the clusters (colour code). To use the system in the simulation, instead of providing a scalar as input to `clusters`, the output of `create_ClusterSystem()` can be injected in every `run_MC` function.

```
run_MC_TL_LOC(
  s = 1e14,
  E = 0.9,
  times = 0:100,
  b = 1,
  n_filled = 1000,
  method = "seq",
  clusters = clusters,
  r = 1) %>%
plot_RLumCarlo()
```



Please note: For the simulation of a dosimetric cluster system, the meaning of `n_filled` changes. Instead of defining the number of electrons per cluster, it becomes the total number of electrons in the system. Electrons are distributed according to the grouping of the single clusters (the colours in the three-dimensional scatter plot). Within one group, electrons are distributed evenly.

## References

- Bulur, Enver. 1996. “An Alternative Technique for Optically Stimulated Luminescence (OSL) Experiment.” *Radiation Measurements* 26 (5): 701–9. [https://doi.org/10.1016/S1350-4487\(97\)82884-3](https://doi.org/10.1016/S1350-4487(97)82884-3).
- Chen, R, and Vasilis Pagonis. 2011. *Thermally and Optically Stimulated Luminescence - A Simulation Approach*. Thermally and Optically Stimulated Luminescence a Simulation Approach. John Wiley & Sons, Ltd.
- Jain, Mayank, Benny Guralnik, and Martin Thalbitzer Andersen. 2012. “Stimulated luminescence emission from localized recombination in randomly distributed defects.” *Journal of Physics: Condensed Matter* 24 (38): 385402. <https://doi.org/10.1088/0953-8984/24/38/385402>.
- Pagonis, Vasilis, Johannes Friedrich, Michael Discher, Anna Müller-Kirschbaum, Veronika Schlosser, Sebastian Kreutzer, Reuven Chen, and Christoph Schmidt. 2019. *Journal of Luminescence* 207: 266–72. <https://doi.org/10.1016/j.jlumin.2018.11.024>.