

# Forest growth and dynamics (practice)

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# Outline

1. Forest growth inputs
2. Running forest growth
3. Evaluation of growth predictions
4. Forest dynamics
5. Forest dynamics including management

M.C. Escher - Up and down, 1947



# 1. Forest growth inputs

# Creating the forest growth input object

We assume we have an appropriate forest object:

```
1 data(exampleforest)
```

a species parameter data frame:

```
1 data(SpParamsMED)
```

a soil data frame:

```
1 examplesoil <- defaultSoilParams(4)
```

and simulation control list:

```
1 control <- defaultControl("Granier")
```

With these four elements we can build our input object for function `growth()`:

```
1 x <- forest2growthInput(exampleforest, examplesoil, SpParamsMED, control)
```

# Structure of the growth input object (1)

The growth input object is a `list` with several elements:

```
1 names(x)
[1] "control"           "soil"
[3] "snowpack"          "canopy"
[5] "herbLAI"           "herbLAImax"
[7] "cohorts"           "above"
[9] "below"             "belowLayers"
[11] "paramsPhenology"   "paramsAnatomy"
[13] "paramsInterception" "paramsTranspiration"
[15] "paramsWaterStorage" "paramsGrowth"
[17] "paramsMortalityRegeneration" "paramsAllometries"
[19] "internalPhenology"  "internalWater"
[21] "internalLAIDistribution" "internalCarbon"
[23] "internalAllocation" "internalMortality"
[25] "internalFCCS"
```

Element `above` contains the above-ground structure data that we already know, but with an additional column `SA` that describes the estimated initial amount of *sapwood area*:

	SP	N	DBH	Cover	H	CR	SA	LAI_live
T1_148	148	168.0000	37.55	NA	800	0.6605196	383.4520992	0.84874773
T2_168	168	384.0000	14.60	NA	660	0.6055642	47.0072886	0.70557382
S1_165	165	749.4923	NA	3.75	80	0.8032817	0.9753929	0.03062604
	LAI_expanded	LAI_dead	LAI_nocomp	Loading	ObsID			
T1_148	0.84874773	0	1.29720268	0.32447403	<NA>			
T2_168	0.70557382	0	1.01943205	0.20102636	<NA>			
S1_165	0.03062604	0	0.04412896	0.01407945	<NA>			

## Structure of the growth input object (2)

Elements starting with `params*` contain cohort-specific model parameters. An important set of parameters are in `paramsGrowth`:

	RERleaf	RERSapwood	RERfineroot	CCleaf	CCsapwood	CCfineroot
T1_148	0.01210607	5.15e-05	0.0009610199	1.5905	1.47	1.3
T2_168	0.01757808	5.15e-05	0.0072846640	1.4300	1.47	1.3
S1_165	0.02647746	5.15e-05	0.0072846640	1.5320	1.47	1.3

	RGRleafmax	RGRsapwoodmax	RGRcambiummax	RGRfinerootmax	SRsapwood
T1_148	0.09	NA	0.002628095	0.1	0.000135
T2_168	0.09	NA	0.002500000	0.1	0.000135
S1_165	0.09	0.002	NA	0.1	0.000135

	SRfineroot	RSSG	fHDmin	fHDmax	WoodC
T1_148	0.001897231	0.3725000	80	160	0.4979943
T2_168	0.001897231	0.9500000	40	100	0.4740096
S1_165	0.001897231	0.7804035	NA	NA	0.4749178

Elements starting with `internal*` contain state variables required to keep track of plant status. For example, the metabolic and storage carbon levels can be seen in `internalCarbon`:

	sugarLeaf	starchLeaf	sugarSapwood	starchSapwood
T1_148	0.4029239	0.00925123	0.5738487	3.201897
T2_168	0.3585751	0.00925123	1.0741383	3.100817
S1_165	0.7223526	0.00925123	0.2857655	2.654773

## 2. Forest growth

# Forest growth run

The call to function `growth()` needs the growth input object, the weather data frame, latitude and elevation:

```
1 G <- growth(x, examplometeo, latitude = 41.82592, elevation = 100)
```

```
Initial plant cohort biomass (g/m2): 5068.34
Initial plant water content (mm): 4.73001
Initial soil water content (mm): 290.875
Initial snowpack content (mm): 0
Performing daily simulations

Year 2001:.....

Final plant biomass (g/m2): 5256.53
Change in plant biomass (g/m2): 188.193
Plant biomass balance result (g/m2): 188.193
Plant biomass balance components:
  Structural balance (g/m2) 104 Labile balance (g/m2) 92
  Plant individual balance (g/m2) 196 Mortality loss (g/m2) 8
Final plant water content (mm): 4.73794
Final soil water content (mm): 274.787
Final snowpack content (mm): 0
Change in plant water content (mm): 0.00793033
Plant water balance result (mm): -0.00116903
Change in soil water content (mm): -16.0878
Soil water balance result (mm): -16.0878
Change in snowpack water content (mm): 0
Snowpack water balance result (mm): 7.10543e-15
Water balance components:
  Precipitation (mm) 513 Rain (mm) 462 Snow (mm) 51
  Interception (mm) 00 Net rainfall (mm) 270
```



# Growth output object

Function `growth()` returns an object of class with the same name, actually a list:

```
1 class(G)
[1] "growth" "list"
```

... whose elements are:

```
1 names(G)
[1] "latitude"      "topography"      "weather"
[4] "growthInput"   "growthOutput"    "WaterBalance"
[7] "CarbonBalance" "BiomassBalance"  "Soil"
[10] "Snow"          "Stand"           "Plants"
[13] "LabileCarbonBalance" "PlantBiomassBalance" "PlantStructure"
[16] "GrowthMortality"
```

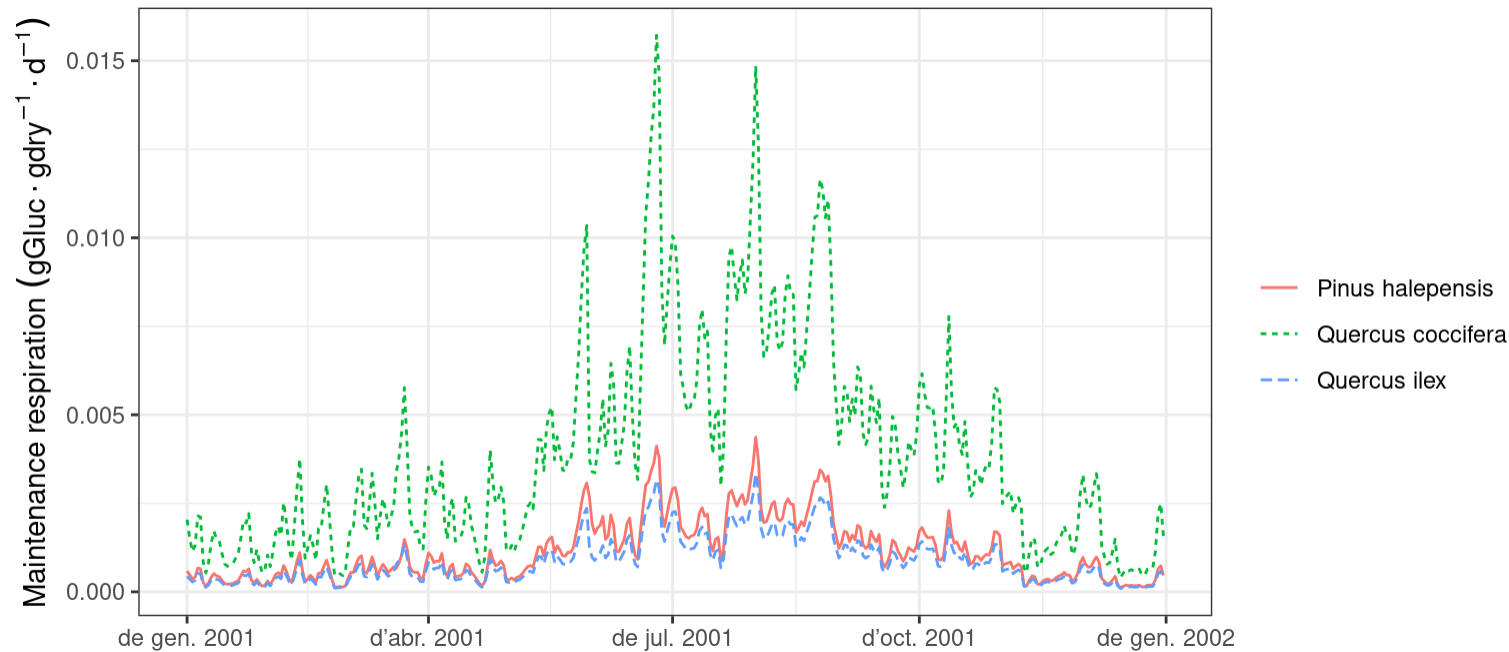
Elements	Information
<code>latitude</code> , <code>topography</code> , <code>weather</code> , <code>growthInput</code>	Copies of the information used in the call to <code>growth()</code>
<code>growthOutput</code>	State variables at the end of the simulation (can be used as input to a subsequent one)
<code>WaterBalance</code> , <code>Soil</code> , <code>Snow</code> , <code>Stand</code> , <code>Plants</code>	[same as <code>spwb</code> ...]
<code>CarbonBalance</code>	Stand-level carbon blance
<code>LabileCarbonBalance</code>	Components of the individual-level labile carbon balance
<code>PlantBiomassBalance</code>	Components of individual- and cohort-level biomass balance
<code>PlantStructure</code>	Structural variables (DBH, height, sapwood area...)
<code>GrowthMortality</code>	Growth and mortality rates

# Post-processing

Users can inspect the output of `growth()` simulations using functions `extract()`, `summary()` and `plot()` on the simulation output.

Several new plots are available in addition to those available for `spwb()` simulations (see `?plot.growth`). For example:

```
1 plot(G, "MaintenanceRespiration", bySpecies = TRUE)
```



... but instead of typing all plots, we can call the interactive plot function `shinyplot()`.

### 3. Evaluation of growth predictions

# Observed data frame

Evaluation of growth simulations will normally imply the comparison of predicted vs observed **basal area increment** (BAI) or **diameter increment** (DI) at a given temporal resolution.

Here, we illustrate the evaluation functions included in the package using a fake data set at *daily* resolution, consisting on the predicted values and some added error.

```
1 data(exampleobs)
2 head(exampleobs)
```

	dates	SWC	ETR	E_T1_148	E_T2_168	FMC_T1_148	FMC_T2_168
1	2001-01-01	0.3007733	2.2436218	0.09187857	0.14142950	125.9071	93.07915
2	2001-01-02	0.3091627	2.3236565	0.26480973	0.19095008	125.9137	93.07863
3	2001-01-03	0.2996498	0.7409083	0.15345643	0.17546363	125.8760	93.10512
4	2001-01-04	0.3042764	1.7173522	0.23470647	0.04643454	125.8643	93.07022
5	2001-01-05	0.3054886	2.0002562	0.37687792	0.10623552	125.8493	93.08487
6	2001-01-06	0.3062005	2.0722706	0.16342360	0.05550329	125.9367	93.07343

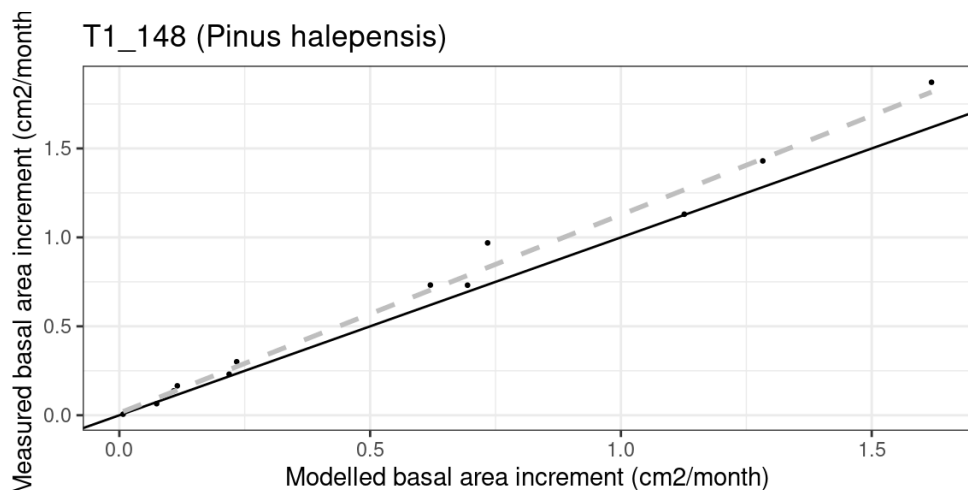
	BAI_T1_148	BAI_T2_168	DI_T1_148	DI_T2_168
1	6.222625e-06	0	9.948881e-08	0
2	3.091274e-10	0	1.071090e-11	0
3	1.298482e-13	0	0.000000e+00	0
4	2.886195e-11	0	5.552753e-13	0
5	1.287020e-03	0	1.367289e-05	0
6	1.471202e-03	0	1.000411e-05	0

To specify observed growth data at *monthly* or *annual scale*, you should specify the first day of each month/year (e.g. [2001-01-01](#), [2002-01-01](#), etc for years) as row names in your observed data frame.

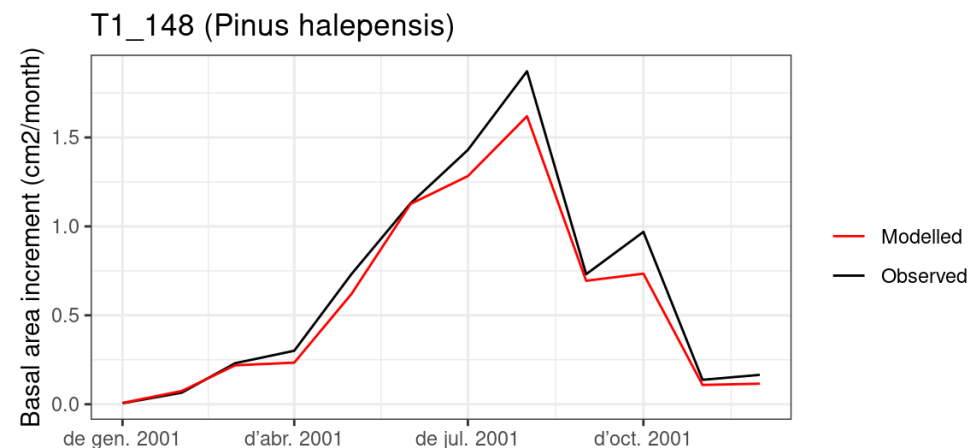
# Evaluation plot

Assuming we want to evaluate the predictive capacity of the model in terms of monthly basal area increment for the *pine cohort* (i.e. T1\_148), we can plot the relationship between observed and predicted values using `evaluation_plot()`:

```
1 evaluation_plot(G, exampleobs, "BAI",
2               cohort = "T1_148",
3               temporalResolution = "month",
4               plotType = "scatter")
```



```
1 evaluation_plot(G, exampleobs, "BAI",
2               cohort = "T1_148",
3               temporalResolution = "month",
4               plotType = "dynamics")
```



Using `temporalResolution = "month"` we indicate that simulated and observed data should be temporally aggregated to conduct the comparison.

The following code would help us quantifying the *strength* of the relationship:

```
1 evaluation_stats(G, exampleobs, "BAI", cohort = "T1_148", temporalResolution = "month")
```

n	Bias	Bias.rel	MAE	MAE.rel	r
12.000000000	-0.07781339	-12.01831387	0.07962106	12.29751007	0.99395791
NSE	NSE.abs				
0.95935190	0.83966426				

## 4. Forest dynamics

# Forest dynamics run

## Weather preparation

In this vignette we will fake a three-year weather input by repeating the example weather data frame four times:

```
1 meteo <- rbind(examplemeteo, examplemeteo, examplemeteo, examplemeteo)
```

we need to update the dates in row names so that they span four consecutive years:

```
1 meteo$dates <- seq(as.Date("2001-01-01"),
2                   as.Date("2004-12-30"), by="day")
```

## Simulation

The call to `fordyn()` has the following form:

```
1 fd<-fordyn(exampleforest, examplesoil, SpParamsMED, meteo, control,
2           latitude = 41.82592, elevation = 100)
```

```
Simulating year 2001 (1/4): (a) Growth/mortality, (b) Regeneration nT = 2 nS = 1
Simulating year 2002 (2/4): (a) Growth/mortality, (b) Regeneration nT = 2 nS = 1
Simulating year 2003 (3/4): (a) Growth/mortality, (b) Regeneration nT = 2 nS = 1
Simulating year 2004 (4/4): (a) Growth/mortality, (b) Regeneration nT = 2 nS = 1
```

### Important

- `fordyn()` operates on `forest` objects directly, instead of using an intermediary object (such as `spwbInput` and `growthInput`).
- `fordyn()` calls function `growth()` internally for each simulated year, but all console output from `growth()` is hidden.

# Forest dynamics output (1)

As with other models, the output of `fordyn()` is a list, which has the following elements:

Elements	Information
<code>StandSummary</code> , <code>SpeciesSummary</code> , <code>CohortSummary</code>	<i>Annual</i> summary statistics at different levels
<code>TreeTable</code> , <code>ShrubTable</code>	Structural variables of <b>living</b> cohorts at each annual time step.
<code>DeadTreeTable</code> , <code>DeadShrubTable</code>	Structural variables of <b>dead</b> cohorts at each annual time step
<code>CutTreeTable</code> , <code>CutShrubTable</code>	Structural variables of <b>cut</b> cohorts at each annual time step
<code>ForestStructures</code>	Vector of <code>forest</code> objects at each time step.
<code>GrowthResults</code>	Result of internally calling <code>growth()</code> at each time step.
<code>ManagementArgs</code>	Management arguments for a subsequent call to <code>fordyn()</code> .
<code>NextInputObject</code> , <code>NextForestObject</code>	Objects <code>growthInput</code> and <code>forest</code> to be used in a subsequent call to <code>fordyn()</code> .



## Forest dynamics output (2)

For example, we can compare the initial `forest` object with the final one:

1	exampleforest	1	fd\$NextForestObject
	\$treeData		\$treeData
	Species N DBH Height Z50 Z95		Species DBH Height N Z50 Z95
	1 Pinus halepensis 168 37.55 800 100 600		1 Pinus halepensis 38.01410 824.7217 166.7796 100 600
	2 Quercus ilex 384 14.60 660 300 1000		2 Quercus ilex 15.04679 673.8471 382.6513 300 1000
	\$shrubData		\$shrubData
	Species Cover Height Z50 Z95		Species Height Cover Z50 Z95
	1 Quercus coccifera 3.75 80 200 1000		1 Quercus coccifera 75.4552 3.237287 200 1000
	\$herbCover		\$herbCover
	[1] 10		[1] 10
	\$herbHeight		\$herbHeight
	[1] 20		[1] 20
	\$seedBank		\$seedBank
	[1] Species Percent		Species Percent
	<0 files> (o «row.names» de longitud 0)		1 Pinus halepensis 100
			2 Quercus ilex 100
			3 Quercus coccifera 100
	attr(,"class")		attr(,"class")
	[1] "forest" "list"		[1] "forest" "list"

## Forest dynamics output (3)

The output includes **summary statistics** that describe the structural and compositional state of the forest corresponding to *each annual time step*.

For example, we can access *stand-level* statistics using:

```
1 fd$StandSummary
```

	Step	NumTreeSpecies	NumTreeCohorts	NumShrubSpecies	NumShrubCohorts
1	0	2	2	1	1
2	1	2	2	1	1
3	2	2	2	1	1
4	3	2	2	1	1
5	4	2	2	1	1

	TreeDensityLive	TreeBasalAreaLive	DominantTreeHeight	DominantTreeDiameter
1	552.0000	25.03330	800.0000	37.55000
2	551.3663	25.20814	806.2122	37.66571
3	550.7269	25.38313	812.4144	37.78185
4	550.0818	25.55825	818.5877	37.89804
5	549.4309	25.73303	824.7217	38.01410

	QuadraticMeanTreeDiameter	HartBeckingIndex	ShrubCoverLive	BasalAreaDead
1	24.02949	53.20353	3.750000	0.00000000
2	24.12711	52.82391	3.092051	0.03917375
3	24.22476	52.45105	3.139858	0.03983747
4	24.32243	52.08602	3.188231	0.04050957
5	24.41991	51.72923	3.237287	0.04118879

	ShrubCoverDead	BasalAreaCut	ShrubCoverCut
1	0.000000000	0	0
2	0.005308898	0	0
3	0.004784468	0	0
4	0.004858342	0	0
5	0.004933117	0	0

## Forest dynamics output (4)

Another useful output of `fordyn()` are tables in long format with cohort structural information (i.e. DBH, height, density, etc) for each time step:

1 fd\$TreeTable											
	Step	Year	Cohort	Species	DBH	Height	N	Z50	Z95	ObsID	
1	0	NA	T1_148	Pinus halepensis	37.55000	800.0000	168.0000	100	600	<NA>	
2	0	NA	T2_168	Quercus ilex	14.60000	660.0000	384.0000	300	1000	<NA>	
3	1	2001	T1_148	Pinus halepensis	37.66571	806.2122	167.6992	100	600	<NA>	
4	1	2001	T2_168	Quercus ilex	14.71218	663.4915	383.6671	300	1000	<NA>	
5	2	2002	T1_148	Pinus halepensis	37.78185	812.4144	167.3956	100	600	<NA>	
6	2	2002	T2_168	Quercus ilex	14.82389	666.9588	383.3314	300	1000	<NA>	
7	3	2003	T1_148	Pinus halepensis	37.89804	818.5877	167.0890	100	600	<NA>	
8	3	2003	T2_168	Quercus ilex	14.93552	670.4135	382.9928	300	1000	<NA>	
9	4	2004	T1_148	Pinus halepensis	38.01410	824.7217	166.7796	100	600	<NA>	
10	4	2004	T2_168	Quercus ilex	15.04679	673.8471	382.6513	300	1000	<NA>	

### Note

The `NA` values in `Year` correspond to the initial state.

## Forest dynamics output (5)

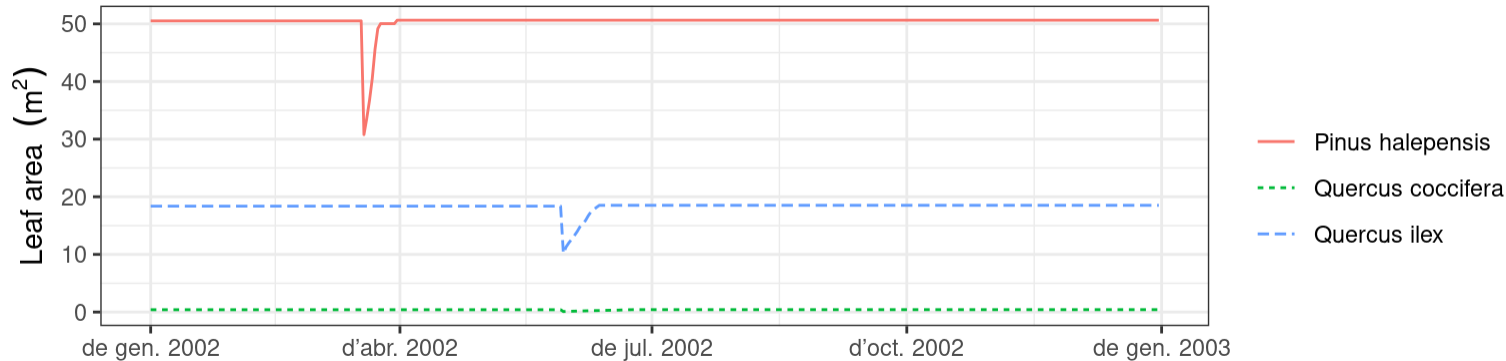
The same information can be shown for trees that are predicted to die during each simulated year:

1 fd\$DeadTreeTable										
	Step	Year	Cohort		Species	DBH	Height	N	N_starvation	
1	1	2001	T1_148		Pinus halepensis	37.66571	806.2122	0.3007828		0
2	1	2001	T2_168		Quercus ilex	14.71218	663.4915	0.3328893		0
3	2	2002	T1_148		Pinus halepensis	37.78185	812.4144	0.3036481		0
4	2	2002	T2_168		Quercus ilex	14.82389	666.9588	0.3357419		0
5	3	2003	T1_148		Pinus halepensis	37.89804	818.5877	0.3065253		0
6	3	2003	T2_168		Quercus ilex	14.93552	670.4135	0.3386082		0
7	4	2004	T1_148		Pinus halepensis	38.01410	824.7217	0.3094089		0
8	4	2004	T2_168		Quercus ilex	15.04679	673.8471	0.3414828		0
			N_dessication	N_burnt	Z50	Z95	ObsID			
1			0		0 100	600	<NA>			
2			0		0 300	1000	<NA>			
3			0		0 100	600	<NA>			
4			0		0 300	1000	<NA>			
5			0		0 100	600	<NA>			
6			0		0 300	1000	<NA>			
7			0		0 100	600	<NA>			
8			0		0 300	1000	<NA>			

# Post-processing

Accessing elements of `GrowthResults`, we can extract, summarize or plot simulation results for a particular year:

```
1 plot(fd$GrowthResults[[2]], "LeafArea", bySpecies = TRUE)
```



It is also possible to plot the whole series of results by passing a `fordyn` object to the `plot()` function:

```
1 plot(fd, "LeafArea", bySpecies = TRUE)
```



Finally, we can create interactive plots using function `shinyplot()`, in the same way as with other simulations.

## 5. Forest dynamics including management

# Running simulations with management

`fordyn()` allows the user to supply an *arbitrary* function implementing a desired management strategy for the stand whose dynamics are to be simulated.

The package includes an in-built default function called `defaultManagementFunction()` along management parameter defaults provided by function `defaultManagementArguments()`.

To run simulations with management we need to define (and modify) management arguments...

```
1 # Default arguments
2 args <- defaultManagementArguments()
3 # Here one can modify defaults before calling fordyn()
4 #
```

... and call `fordyn()` specifying the management function and its arguments:

```
1 fd<-fordyn(exampleforest, examplesoil, SpParamsMED, meteo, control,
2           latitude = 41.82592, elevation = 100,
3           management_function = defaultManagementFunction,
4           management_args = args)
```

When management is included, tables `CutTreeTable` and `CutShrubTable` will contain extraction data:

```
1 fd$CutTreeTable
2 fd$CutShrubTable
```

# Forest management parameters (1)

Function `defaultManagementArguments()` returns a list with default values for *management parameters* to be used in conjunction with `defaultManagementFunction()`:

Element	Description
<code>type</code>	Management model, either ‘regular’ or ‘irregular’
<code>targetTreeSpecies</code>	Either "all" for unspecific cuttings or a numeric vector of target tree species to be selected for cutting operations
<code>thinning</code>	Kind of thinning to be applied in irregular models or in regular models before the final cuts. Options are "below", "above", "systematic", "below-systematic", "above-systematic" or a string with the proportion of cuts to be applied to different diameter sizes
<code>thinningMetric</code>	The stand-level metric used to decide whether thinning is applied, either "BA" (basal area), "N" (density) or "HB" (Hart-Becking index)
<code>thinningThreshold</code>	The threshold value of the stand-level metric causing the thinning decision
<code>thinningPerc</code>	Percentage of stand’s basal area to be removed in thinning operations
<code>minThinningInterval</code>	Minimum number of years between thinning operations
<code>finalMeanDBH</code>	Mean DBH threshold to start final cuts
<code>finalPerc</code>	String with percentages of basal area to be removed in final cuts, separated by ‘-’ (e.g. “40-60-100”)
<code>finalYearsBetweenCuts</code>	Number of years separating final cuts



## Forest management parameters (2)

The same list includes *state variables* for management (these are modified during the simulation):

Element	Description
<code>yearsSinceThinning</code>	State variable to count the years since the last thinning occurred
<code>finalPreviousStage</code>	Integer state variable to store the stage of final cuts ('0' before starting final cuts)
<code>finalYearsToCut</code>	State variable to count the years to be passed before new final cut is applied.

### Tip

Instead of using the in-built management function, you could code your own management function and specify its own set of parameters!

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