

D5.1: Common optimisation protocol

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Document Information

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Deliverable status

Summary

The objective of this deliverable D5.1 is to enable catchment-scale modellers to perform a multi-objective optimisation of the allocation and combination of Natural/Small Water Retention Measures (NSWRMs) in their own case study (CS).

This report (i) introduces the Pareto optimal NSWRM implementation plans as one of the project's key products, (ii) describes OPTAIN's optimisation concept, (iii) outlines the requirements that a SWAT+ model setup must meet before it can be used for the optimisation, (iv) shows how to build a SWATmeasR project as a key tool for implementing NSWRMs in a SWAT+ model, and (v) provides a protocol on how to run the optimisation using the software CoMOLA.

The report should also be useful beyond the OPTAIN project for interested SWAT+ modellers who wish to use their model to optimise spatially explicit NSWRM or Best Management Practice (BMP) plans against multiple catchment-scale objectives.

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Abbreviation list

WP Work Package

OPTAIN

1. Introduction

1.1. **Objective**

The objective of Task 5.1 of the OPTAIN project is to explore where to implement which measure(s) within each of the case studies in order to best possible meet various environmental and socio-economic objectives. Based on previous Multi-Actor Reference Group (MARG) workshops, we have learned which water and nutrient retention problems are most relevant in the different case studies and have identified a number of promising Natural/Small Water Retention Measures (NSWRMs) that we can model at individual sites using the SWAT+ model in combination with the Contiguous Object COnnectivity Approach (COCOA) developed in this project (Schürz et al., 2022). We have run a set of model scenarios, where we evaluated the effectiveness of individual NSWRMs, considering all sites where the implementation appeared reasonable (Piniewski et al., 2024). The evaluation was based on a wide range of environmental performance indicators (EPIs) relating to water and nutrient retention and crop yield. We also developed and applied functions to calculate a number of socio-economic performance indicators - SPIs (such as agricultural gross margin and implementation costs), which are described in detail in deliverable D2.2 (Krzeminska & Monaco, 2022) and in the forthcoming deliverable D4.5 ('Attractiveness and socio-economic assessment of NSWRMs'). What is still missing, and will be addressed in WP 5, is an integrative assessment (a cost-benefit analysis) that takes into account both the most relevant EPIs and the most relevant SPIs, while at the same time allowing for any individual allocation and combination of the NSWRMs at hand. In a particular location within the catchment, some NSWRMs may be more efficient at retaining water and nutrients than others. At other sites, a combination of different NSWRMs may be most effective. Any NSWRM implementation is likely to involve trade-offs (e.g. with agricultural production and implementation costs), and local site-level effects may not translate linearly to the catchment level. Given the large number of possible implementation sites for the set of different NSWRMs (~3-6 depending on the case study), there could be millions of different combinations (hereafter referred to as 'NSWRM plans'), each with different environmental and socioeconomic performance.

In order to explore the best possible NSWRM plans for different, sometimes conflicting objectives, the SWAT+ models built in each case study need to be coupled with an efficient search algorithm. In OPTAIN, we use the Non-dominated Sorting Genetic Algorithm II (NSGA-II), a widely used Pareto-based optimisation algorithm (Deb et al., 2002), which is embedded in the Constrained Multi-objective Optimisation of Land use Allocation (CoMOLA) tool, developed at UFZ (Strauch et al., 2019). The result of such a search is not a single optimal NSWRM plan, but a large number of Pareto optimal solutions (Figure 1.1). Pareto optimal solutions are solutions for which no objective can be further improved without compromising at least one of the other objectives. From such a set of best alternatives (with minimal trade-offs between the objectives), decision makers can discuss and select appropriate solutions according to their preferences (Cord et al., 2017, Kaim et al.,

2020). In a later task of work package (WP) 5, each case study will conduct such a preference analysis with its MARG (Task 5.4). The ultimate goal of this Task 5.1 is to enable CS modellers to conduct their own multi-objective optimisation of NSWRM plans. As a result, each case study should be able to run its SWAT+ model within CoMOLA to identify NSWRM plans that are close to Pareto optimality. The following chapters were discussed at a 3-day workshop with the OPTAIN CS modellers and (partially) applied in practice.

Agricultural gross margin

Figure 1.1: Schematic illustration of Pareto optimal NSWRM implementation plans

2. OPTAIN's optimisation concept

Multi-objective optimisation of management options requires four components, regardless of its purpose:

(1) The definition of optimization objectives,

- (2) the definition of the decision space,
- (3) functions evaluating the objectives based on the decisions, and
- (4) a multi-objective optimisation algorithm.

In the following, we briefly describe how these components are defined in OPTAIN to optimise NSWRM plans.

Definition of optimisation objectives $2.1.$

First and foremost, we have to define the objectives (or goals) for which the NSWRM plans are to be optimised. OPTAIN's WP2 defined a number of different environmental and socio-economic performance indicators (EPIs/SPIs) that could be used as optimisation objectives. The multi-objective optimisation algorithm implemented in CoMOLA (NSGA-II) can handle up to four objectives. Ideally, NSWRM plans should solve or minimise the most relevant water and nutrient related problem(s) of the case study at minimum cost to farmers and society. Therefore, the set of optimisation objectives should consist of the following indicators:

- (1) One EPI that is addressing the most relevant water related problem (e.g. soil water content or a river discharge indicator for a specific period within the year),
- (2) Another EPI that is addressing the most relevant nutrient related problem (e.g. nitrogen, phosphorus or sediment concentrations or loads at the catchment outlet). If there is no nutrient related problem in the case study or if SWAT+ has not been calibrated for nutrients, a second water related EPI could be selected.
- (3) One SPI that is indicating the catchment's agricultural production (e.g. the sum of all crop-specific yields expressed as grain units or the agricultural gross margin).
- (4) Another SPI that is referring to implementation and maintenance costs of NSWRMs, with or without considering subsidies. Agricultural production, NSWRM costs and subsidies can also be combined in one indicator if appropriate. This would leave room for one more indicator to be considered as optimisation objective

It is also possible to integrate several indicators into one objective function. For example, if total nitrogen (TN) and total phosphorus (TP) loads were to be used together as one water quality objective, TN and TP loads could simply be summed. However, as TN loads are naturally higher than TP loads, TN loads would be overemphasised. Appropriate weighting may therefore be required. The definition of performance indicators as optimisation objectives is a crucial step for each case study. Section 4.3 shows in more detail how to define the objectives within the CoMOLA workflow.

Definition of the decision space $2.2.$

The decisions that can be made when optimising NSWRM plans are obvious: which individual NSWRM (type and location) are implemented simultaneously in the catchment. In collaboration with local stakeholders, each case study has defined its own types of NSWRMs and, for each type, all the potential sites where implementation could make sense. Each decision where to implement a certain type of NSWRMs may result in a different achievement of the objectives defined above. NSGA-II is a genetic optimisation algorithm. Catchment-wide NSWRM plans must therefore be encoded as a genome. A genome is a string of genes expressed as integers. In our optimisation, each gene refers to an individual NSWRM type at a particular location within the catchment, where the integer value of each gene can be either 1 (NSWRM is not implemented) or 2 (NSWRM is implemented). Figure 2.1 illustrates the encoding of NSWRM plans as genomes.

Figure 2.1: Encoding NSWRM plans as genomes

The SWAT+ model will be used to evaluate the performance (fitness) of individual NSWRM plans proposed by NSGA-II. It is therefore necessary that SWAT+ model input files, representing individual NSWRM at specific locations, can be written according to the genome in an automated workflow within the optimisation. This is achieved using the SWATmeasR R package (https://git.ufz.de/schuerz/swatmeasr), which is described in more detail in Section 3.2.

Functions evaluating the objectives $2.3.$

In OPTAIN, the environmental optimisation objectives will be calculated by running SWAT+ and postprocessing its output using a set of R scripts (see Section 4.3). The most challenging part of the optimisation is to provide a SWAT+ model that meets a number of requirements, which are explained in more detail in Section 3.1.

The calculation of the socio-economic optimisation objectives is based on selected input files (e.g. parameterization of the agricultural management) and output files (e.g. crop yield simulations) of the SWAT+ model, as well as a set of CS-specific economic parameters. In practice, the calculation of the SPIs is also done using an R script in the postprocessing of a SWAT+ run. Details on the calculation of the socio-economic optimisation objectives are described in OPTAIN's in deliverables D2.2 and D4.5.

$24.$ Multi-objective optimisation algorithm

Multi-objective optimisation algorithms are increasingly used for land use allocation problems in agricultural landscapes, as they are able to approximate Pareto optimal solutions from a large number of possible land use/land management configurations (Kaim et al., 2018; Memmah et al., 2015). OPTAIN uses NSGA-II (Deb et al., 2002), a popular multi-objective genetic algorithm included in the CoMOLA software. The step-by-step process of this algorithm is as follows (see also Figure 2.2):

Step 1: Initialization

Generate Initial Population: Create an initial population P_0 of N individuals (i.e. individual NSWRM plans) randomly within the feasible^{[1](#page-11-1)} solution space. Population size ^N is an important parameter to be adjusted by case studies (section 4.3).

Evaluate Population: Calculate the fitness values of each individual based on the objective functions. In other words, run SWAT+ for each NSWRM plan of the initial population and calculate the value of the optimisation objectives based on the SWAT+ outputs and the socio-economic parameters (section 2.1).

Step 2: Non-dominated Sorting

Rank Individuals: Sort the population into different fronts using non-dominated^{[2](#page-11-2)} sorting. Front F_1 consists of non-dominated individuals, F_2 consists of individuals dominated only by those in F_1 , and so on.

Crowding Distance Assignment: Calculate the crowding distance³ for each individual in the population. This metric helps maintain diversity by favouring individuals in less crowded regions of the solution space.

¹ In OPTAIN, maximum implementation scenarios of single NSWRM have been co-created with local stakeholders. We therefore assume that all possible NSWRM plans are feasible, although some NSWRM combinations may in fact be unrealistic. Feasibility of NSWRM combinations will be addressed at a later stage when preferred solutions are identified with local stakeholders from the final set of Pareto optimal NSWRM plans. By assuming feasibility of all NSWRM plans during the optimisation, computationally intensive constraint handling is avoided.

 2 An individual A dominates another individual B if A is no worse than B in all objectives and better in at least one objective.

 3 For each individual, the crowding distance is calculated based on the average distance of the two neighbouring individuals on either side along each objective. This distance measures how close an individual is to its neighbours.

Step 3: Selection

Tournament Selection: Use binary tournament selection based on rank and crowding distance to select parent individuals for the next generation. An individual with a lower rank is preferred, and if ranks are equal, the one with a higher crowding distance is preferred.

Step 4: Create and Evaluate Offspring Population

The offspring population Q_t of size N is generated using crossover and mutation from the selected parents.

Crossover: Apply crossover (recombination) operators to selected parents to generate offspring. Common crossover methods include single-point, multi-point, and simulated binary crossover (SBX).

Mutation: Apply mutation operators to offspring to introduce variability. This can involve bit-flip mutations for binary representations or polynomial mutations for real-valued representations.

Evaluate Offspring Population: Calculate the fitness values of each individual based on the objective functions. In other words, run SWAT+ for each NSWRM plan of the offspring population and calculate the value of the optimisation objectives based on the SWAT+ outputs and the socio-economic parameters (section 2.1).

Step 5: Combine Populations

Combine the parent population P_t and offspring population Q_t to form a combined population R_t of size 2N.

Step 6: Non-dominated Sorting of Combined Population

Perform non-dominated sorting on the combined population R_t to identify the different fronts $F_1, F_2, ...$

Step 7: Selection of New Population

Start with an empty new population $P_{\text{t+1}}$. Add individuals from each front F_{t} to $P_{\text{t+1}}$ until the population size exceeds N.

If adding all individuals from front F_i causes the population size to exceed N, sort the individuals in F_i by their crowding distance in descending order and select the top individuals to fill the population up to size N.

Step 8: Loop Until Termination

Repeat the process from step 3 to step 7 for a predefined number of generations or until a convergence criterion is met. In OPTAIN, case studies need to define the number of generations (section 4.3). The core principles of this loop are shown in Figure 2.2.

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Figure 2.2 (source: Jiang et al., 2021): Core principles of the NSGA-II: Apply non-dominated sorting to the combined (parent and offspring) population. If necessary, use crowding distance to cut off the sorted combined population to population size N (forming the new parent population). Apply tournament selection, crossover and mutation to the new parents to form a new offspring population and calculate its fitness values. The procedure is repeated until a termination criterion is met.

3. SWAT+ model preparation

$3.1.$ Basic SWAT+ model configuration

All SWAT+ modelling tasks in OPTAIN strongly build on the preceding steps in the modelling workflow. This is particularly true for the NSWRM implementation in the SWAT+ model setups and eventually for the optimization of NSWRM combinations. Based on the decisions made for a SWAT+ model setup in OPTAIN the R package SWATmeasR has been developed to enable an automatable approach for the implementation of NSWRMs in SWAT+ model setups. The conceptual approach on how specific NSWRMs are represented in SWAT+ model setups has been defined by Marval et al. (2022). To enable the implementation of structural NSWRMs in particular, the SWAT+ model setups are required to represent a great detail of connection of water and nutrient fluxes in the landscape. In OPTAIN we therefore developed the contiguous object connectivity approach (COCOA) and implemented this concept in the newly developed SWAT+ model builder SWATbuildR (Schürz, 2024a) which must be used by every OPTAIN case study to set up their SWAT+ models. Farm management operation schedules were implemented in every case study model setup with the use of the R package SWATfarmR (Schürz, 2024b). In order to analyse the effects of management-related NSWRMs, different management plans were set up for each field unit in a model

that represents the status quo and a possible scenario. For an analysis, the individual management schedules must be interchangeable to switch between them.

Only SWAT+ model setups that follow the OPTAIN model setup guidelines can use the full functionality for the implementation of NSWRMs with SWATmeasR which is documented below. A detailed guideline on the required model setup procedure is documented in Schürz et al. (2022). Nevertheless, SWAT+ model setups created with different approaches (e.g. using QSWAT+) can use a limited set of NSWRM types to be implemented with SWATmeasR. The implementation of land use change type NSWRMs (such as any greening measure or afforestation) does not necessarily require any specific configuration of the model's land object connectivity. Also, the addition of wetland water storages to HRUs can be done with SWATmeasR independent of the land object connectivity (if water routing of a land object is not changed). The implementation of structural measures, such as ponds, utilises the COCOA approach when implemented in a model setup and therefore requires the SWAT+ model to be built with SWATbuildR. The implementation of farm management related NSWRMs rely on SWATfarmR projects in the current form of SWATmeasR. However, this can be generalised to the use of management.sch input files in a later version of SWATmeasR.

Building a SWATmeasR project $3.2.$

SWATmeasR is an R package (https://git.ufz.de/schuerz/swatmeasr) for a fast and easy implementation of NSWRMs in SWAT+ model setups. SWATmeasR enables the implementation of selected NSWRMs, such as land use change, changes in the farm management, or structural measures such as ponds and wetlands. With SWATmeasR, NSWRMs are defined by specific parameters and their "locations" in a model setup. All defined NSWRMs can be easily combined and implemented in the corresponding SWAT+ model setup.

3.2.1. General SWATmeasR workflow

The SWATmeasR workflow can be separated into two major parts, the setup of a SWATmeasR project and the implementation of NSWRMs with an existing SWATmeasR project. Figure 3.1 provides a general overview of the workflows in the setup and the implementation phase.

The setup of a SWATmeasR project involves three steps: (i) the initialization of a new SWATmeasR project with the function new_measr(), (ii) loading the definitions (parameterizations) of all NSWRMs that should be implemented in the model setup with measr project\$load nswrm definition() and (iii) the definition of the locations of all NSWRMs that should be implemented with measr_project\$load_nswrm_location().

The NSWRM implementation is exemplary illustrated in Figure 3.1 for the implementation of measures in the CoMOLA workflow. However, the NSWRM implementation procedure is the same in any modelling workflow in which the effects of NSWRMs are to be simulated. NSWRMs are implemented in the SWAT+

model with the function measr_project\$implement_nswrm(), which implements the selected NSWRMs in the SWAT+ input tables in R. To write the changes in the SWAT+ text input files into the SWAT+ project folder, the function measr_project\$write_swat_inputs() must be called. Once the NSWRMs have been implemented, model simulations are carried out and the results are analysed to examine the effects of the implemented NSWRMs. The initial state of the SWAT+ text input files in the project folder and in the corresponding tables in R is restored by calling the function measr_project\$reset().

Figure 3.1: Overview of the SWATmeasR workflow.

3.2.2. Initialising a new SWATmeasR project

SWATmeasR is based on the R6 object class (Chang, 2022), which is an implementation of object-oriented programming in R. In simple words, R6 objects can contain data and functions which, when executed, can modify the data stored in the object. The following code generates a new SWATmeasR project that creates such an object in R, which stores data and provides functions to apply to the stored data.

```
new_measr(project_name = 'measr_project', 
           project_path = 'Path:/to/your/SWAT/txt/folder')
```
The function new_measr() initialises a new SWATmeasR project (in the following simply called measr project) with the name that was provided by the input argument project_name. The input argument project_path defines the path of the SWAT+ project folder on the hard drive for which the new SWATmeasR project should be generated. The initialization of the new SWATmeasR project reads all relevant input text files from the SWAT+ project folder in project_path and stores them in an object with the project_name in the R environment. Additionally to the SWAT+ input tables, the new SWATmeasR provides a set of functions that will be used later to e.g. load additional data or implement NSWRMs. At the same time when the R object is generated, a *.measr file with the name "project_name.measr" is saved in the SWAT+ project folder. It contains the same data as the object in the R environment and is essentially a backup of the project in R, which can be loaded into R with the function $load_measure()$. Each time the SWATmeasR project is modified in R, also the *.measr file on the hard drive is written and updated.

In the example in Figure 3.2, a SWATmeasR project was initialised with the project_name = "schoeps_240502". When calling the function new_measr() it prints that the SWAT+ input files were read from the SWAT+ project folder defined by project_path (Figure 3.2a). Figure 3.2 a) also shows the list of SWAT+ input tables, which were read and saved in the new R object schoeps_240502. The new SWATmeasR project is present in the R environment (Figure 3.2b) and a file with the name "schoeps_240502.measr" was written into the SWAT+ project folder (Figure 3.2c).

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Figure 3.2: The SWATmeasR project with the project_name = 'schoeps_240502', which was generated with the function new_measr()(a). The new measr_project is available as an object in the R environment (b), and as a *.measr file in the SWAT+ project folder (c).

3.2.3. General structure of a SWATmeasR project

The initialised SWATmeasR project is an object in R that can be handled similarly to lists in R. In simple terms, it is an object with several elements that can be accessed with the \$ operator, just like lists in R. The major difference to lists is that the SWATmeasR project also contains functions that affect the internal data elements (functional programming). In the following, the general structure of a SWATmeasR project is outlined. Most of the elements may not be directly relevant for all SWATmeasR users. Nevertheless, it can be helpful when identifying problems or customising the content of a SWATmeasR project if you know where to find what.

Functions and .data

The example in Figure 3.3 shows the initialised project schoeps_240502. The \$ operator can be used to access the elements of schoeps_240502 and all the functions available in the SWATmeasR project become visible:

- .\$reset resets the SWATmeasR project and the corresponding SWAT+ project folder to its initial condition if NSWRMs were implemented before with .\$implement nswrm and .\$write swat inputs.

- .\$save is usually triggered internally by the other functions and will save the current state of the R object to the *.measr file in the SWAT+ project folder. This can be useful if, for example, elements in the SWATmeasR project were adjusted manually.
- \$write_swat_inputs writes the modified SWAT+ input files into the SWAT+ project folder after NSWRMs were implemented with .\$implement_nswrm.
- .\$reload_swat_inputs reloads all SWAT+ input files. This may be necessary if e.g. entries in the SWAT+ input files were missing but are required for the implementation of specific NSWRMs.
- . Simplement nswrm implements NSWRMs in the SWAT+ input files in the SWATmeasR object in R (not in the SWAT+ project folder!).
- .\$load_nswrm_location is used to load the NSWRM locations input file into the SWATmeasR project in R.
- .\$load_nswrm_definition is used to load an NSWRM definition input file into the SWATmeasR project in R.
- . \$initialize is the synonymous function to new_measr() and should not be called from the SWATmeasR project.
- .\$. enclos_env is some R6 specific argument and is not relevant for the SWATmeasR user.
- .\$.data stores all data of a SWATmeasR project including all SWAT+input tables and the NSWRM definitions.

Meta information:

A SWATmeasR project stores some meta information about the project. The meta information can be accessed as follows:

```
> schoeps_240502$.data$meta
#> $project_name
#> [1] "schoeps_240502"
#>
```


```
#> $project_path
#> [1] "C:/Users/schuerz/Documents/optain/swat/240322_txt"
#> 
#> $measr_version
#> [1] "0.8.0"
```
The element .\$.data\$meta returns the project_name, so the name of the SWATmeasR project in the R environment and on the hard drive, the project_path, which is the path from where the SWATmeasR project was loaded (if the project folder was moved before loading the project, this path will also be different), and the version of the SWATmeasR project, which is the R package version of SWATmeasR that was used to build the SWATmeasR project. It is always recommended that the SWATmeasR version of the project and the current version of the R package match.

Model input tables

The SWAT+ model input files, which were read and saved in the SWATmeasR R object, can be found in .\$data\$model_setup. Two copies of the relevant SWAT+ input files are saved in a SWATmeasR project, the unmodified input files stored in .\$data\$model_setup\$original_inputs and the input files that will be modified when implementing NSWRMs stored in .\$data\$model_setup\$modified_inputs. After the initialisation, the original_inputs and the modified_inputs are identical. Table 3.1 summarises the SWAT+ input tables that are read and stored in a SWATmeasR project.

Not all of the listed SWAT+ input tables will be modified when NSWRMs are implemented. Some of the input tables are read as lookup tables and to check whether specific entries and parameterisations are defined in a model setup.

In addition to the SWAT+ inputs, .\$data\$model_setup\$original_inputs and .\$data\$model_setup\$modified_inputs contain elements that track the implementation of NSWRMs in the input tables. The element file_updated returns a boolean vector which shows the input files that were updated through the implementation of NSWRMs with TRUE.

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The element files_written returns whether or not the modified input files were written to the SWAT+ project folder. In the example below for the initial setup of schoeps_200502, no input files were written yet.

```
> schoeps_240502$.data$model_setup$modified_inputs$files_written
#>
#> [1] FALSE
```
If NSWRMs were implemented, the element implemented_nswrms becomes available in .\$data\$model_setup\$modified_inputs. implemented_nswrms is a table which tracks the spatial objects in the SWAT+ model setup where NSWRMs were implemented. In the example below from the SWATmeasR project schoeps_200502, the NSWRM types grassslope, wetland and pond were implemented in some HRUs. The table implemented_nswrms can be accessed as follows:

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The column nswrm indicates the type of NSWRM which was implemented. obj_typ and obj id define the location where an NSWRM was implemented. NSWRMs are currently only implemented in HRUs, but there may be options to implement NSWRMs in other object types (e.g. channels) in the future. If the implementation of an NSWRM changes the object type of a spatial object, this change is documented in the columns obj_type_new and obj_id_new. In the case of a pond implementation, the HRU land objects are replaced by reservoir objects. The new object type in this case is therefore res. The object ids in the respective SWAT+ input files are specified by obj_id_new.

NSWRM definition

If NSWRMs have been defined for a SWATmeasR project, the element .data\$nswrm_definition becomes available. It collects all relevant input to define the NSWRM parameters and locations of a SWATmeasR project. Most of the elements .data\$nswrm_definition are explained in more detail in the following sections, as the availability of specific elements depends either on whether certain NSWRMs are defined or whether a definition step has already been carried out or not (e.g. the definition of the NSWRM locations). The only element, which is always available is the nswrm_lookup. This provides a general overview of the NSWRMs that were defined for the SWATmeasR project.

```
> schoeps_240502$.data$nswrm_definition$nswrm_lookup
#>
# * A tibble: 8 \times 2
#> type nswrm
#> <chr> <chr>
#> 1 land_use buffer 
#> 2 land_use hedge
#> 3 land_use grassslope
#> 4 land use contr drn
#> 5 management lowtillcc
#> 6 management status_quo
#> 7 wetland
#> 8 pond pond
```
nswrm_lookup links the general NSWRM types with the actual names of the defined NSWRMs. This is particularly relevant for the NSWRM types land_use and management, as the example above shows.

3.2.4 NSWRM definition

All NSWRMs that can potentially be implemented in a SWAT+ model setup must be defined. The definition of NSWRMs is done with the function .\$load_nswrm_definition(). With the current SWATmeasR version 0.8.0, five NSWRM types can be implemented. Table 3.2 summarises the NSWRM types:

NSWRM type	Definition
management	Farm related measures that cause changes in the management schedules, e.g. conservation farming, reduced tillage, cover crops.
land use	Any land use change type measure, e.g. implementation of buffer strips, grassed waterways, or afforestation.
wetland	Transformation of a land object to a wetland by adding water storage to the land object and optionally changing land use parameters.
constr wetland	Implementation of an in stream constructed wetland. In its function this measure is identical to ponds and only different in its naming.
pond	Replacement of a land object by a reservoir object with additional changes in the object routing, e.g. receiving water from channels, routing water directly to channels.

Table 3.2: Overview of the NSWRM types that can be implemented with SWATmeasR.

The general way of NSWRM definition is outlined in the small code example below. The required input file, which provides all necessary parametric inputs, is provided with the input argument file_path. To indicate which type of NSWRM is defined the correct type must be passed with the argument type. The optional argument overwrite gives the option to overwrite already existing NSWRM definitions of the same type. By default overwrite = FALSE. Therefore, if for example a definition of land use type NSWRMs already exists, but a new one should be loaded, overwrite has to be set to overwrite = TRUE. The five possible NSWRM types require different definition input files, which will be explained in the following.

```
> land_def_path <- './nswrm_definition/settings_land_use.csv'
> schoeps_240502$load_nswrm_definition(
     file_path = land_def_path,
    type = 'land\_use')
```
Definition of NSWRM type = 'management'

The definition of management type NSWRMs requires that the management of the status quo and all potential management scenarios are defined by SWATfarmR projects, which are all located in the SWAT+ project folder. Figure 3.4 shows an example of two .farm files of the status quo (farmR_sq_15yr.farm) and one scenario (lowtillcc.farm). The example shows the .farm files only. For retrieving the scheduled operations, also the .mgts files for the SWATfarmR projects are required to be

present in the project folder. It is very important at this stage, that the SWATfarmR projects include the current version of the SWAT+ project and no modifications were made to input files which are relevant for management operation scheduling (e.g. hru-data.hru, landuse.lum, plant.ini, or management.sch). If any of the files in the SWAT+ project have been updated after the generation of the SWATfarmR projects, those changes will not be present in the SWATfarmR projects and will eventually be missing in the SWATmeasR project.

Name	Date modified	Type	Size		
rout_unit.ele	2023-02-24 6:06 PM	ELE File	445 KB		
farmR_sq_15yr.farm	2024-02-14 2:27 PM	FARM File	11.895 KB		
lowtillcc.farm	2023-10-25 8:21 PM	FARM File	11,905 KB		
field.fld	2023-02-24 6:07 PM	FLD File	321 KB		
fertilizer.frt	2022-07-26 7:55 PM	FRT File	8 KB		

Figure 3.4: SWATfarmR projects located in the SWAT+ project folder.

The input file required to define management type NSWRMs is prepared with the function prepare_management_scenario_inputs(), which uses the available SWATfarmR projects. The function performs multiple checks on the provided SWATfarmR projects, such as if all cover the same time period in scheduled operations, or if all schedule names given in the status quo are also available in all scenarios. When processing the management schedules from the SWATfarmR projects, prepare_management_scenario_inputs()also compares whether scheduled operations in the status quo are considered to be the same operation in the corresponding schedule in a management. If this is the case, but only the dates between status quo and scenario differ within a time frame of +/- 21 days, the date of the operation in the scenario is set to the date of the corresponding operation in the status quo case. This step ensures that all matching operations are identical and only those operations that are present in a scenario but not in the status quo are different. An example for the implementation of prepare_management_scenario_inputs()is given below.

```
> # Path to the SWAT+ project folder
> proj_path <- './txtinout'
> # Name of the status quo SWATfarmR project
> statquo_name <- 'farmR_sq_15yr'
> # If op_data3 labels in status_quo and scenarios are synonymous
> syns <- data.frame(status_quo = c('cultiv25', 'cultiv20'),
                    scenario = c('fldcul12', 'fldcul12'))> # Path to write the management definition file
> mgt_wrt_path <- './nswrm_definitions'
>
> prepare_management_scenario_inputs(project_path = proj_path,
                                      status_quo = statquo_name,
                                     synonyms = syns,
                                     write\_path = mgt\_wrt\_path)
```


The input argument project path specifies the path to the SWAT+ project folder. To identify the SWATfarmR project, which has to be considered as the status quo project, the argument status_quo must provide the name of the SWATfarmR project. synonyms is an optional input argument to define operations that the date correction routine should consider as identical operations, even though the op_data3 labels of the operation in the status quo and in a scenario are different. This may be the case, for example, if the tillage operation in the status quo and a conservation tillage scenario should be on the same day, but the tillage type has changed in the conservation tillage scenario. The write_path defines where the management type definition file should be written to.

Once all checks on the SWATfarmR projects and the date correction of the scenario operation schedules have been successful, the management type definition file is written to the defined path. The returned file is an .rds file, which always has the timestamp of its generation as a prefix, followed by the name 'mgt_scenarios'. Figure 3.5 shows an example of a successfully written management input file which can then be used to load the management type NSWRM definition.

ㅅ Name	Date modified	Type	Size
20231128_1751_mgt_scenarios	2023-11-30 2:14 PM	File folder	
R 20231128_1751_mgt_scenarios.rds	2023-11-28 5:51 PM	RDS File	1.079 KB
Xa measure_location.csv	2023-11-02 4:53 PM	Microsoft E	12 KB
xa settings_land_use.csv	2023-11-30 1:52 PM	Microsoft E	1 KB
xm settings_pond.csv	2024-05-02 1:29 PM	Microsoft E	1 KB

Figure 3.5: Example of a management input file, which was generated with the function prepare_management_scenario_inputs().

The generated management input file can then be loaded into the SWATmeasR project with the function .\$load_nswrm_definition(). As shown in the example below the file_path must be the one of the generated management .rds file and type must be 'management'.

```
> mgt_def_path <-
     './nswrm_definition/20231128_1751_mgt_scenario.rds'
> schoeps_240502$load_nswrm_definition(
     file_path = mgt_def_path,
     type = 'management')
```
After loading the management definition file, the management NSWRMs are added to the lookup table .\$.data\$nswrm_definition\$nswrm_lookup, and the element management is added to .\$.data\$nswrm_definition which stores the management inputs for the status quo and for all scenario operation schedules. As for the implementation of management operations not only the management schedules in management.sch are relevant, but also the corresponding plant communities and land use definitions, all management scenarios and the status

quo provide the input files hru_data.hru, landuse.lum, management.sch, and plant.ini.

Definition of NSWRM type = $'$ land use'

The definition of land use type NSWRMs requires the definition of their general land use information into which the land use of a land object will be converted when the NSWRM is implemented. The land use definitions are prepared in a *.csv file and loaded into the SWATmeasR project with the function .\$load_nswrm_definition(). An example land use input table is shown in Table 3.3.

Table 3.3: Example land use input table for loading land use type NSWRMs into a SWATmeasR project.

The example in Table 3.3 defines the four land use type NSWRMs buffer, hedge, grassslope and contr_drn (controlled drainage). The labels defined in the column nswrm will be the names of the measures in the SWATmeasR project. The columns plnt_com to tile are the entries in the *landuse.lum* SWAT+ input file which are written when a defined land use type measure is implemented in the SWAT+ model setup. The labes in the columns are references to other SWAT+ input files, whereby, for example, the label in the plnt_com column refers to a plant community in the plant.ini input file or mgt refers to a management schedule in the management.sch file. The last column lum_dtl points to entries in the lum.dtl decision table. Each of the entries must already be defined in the respective input files when the land use type NSWRM definition is loaded. .\$load_nswrm_definition() checks that all of the entries are already defined for the SWAT+ model setup and returns an error if some of the land use definitions are missing.

The example in Figure 3.6 shows the error that is returned if entries are missing in the SWAT+ input files. The error message shows that the entry 'wrong_com' is not defined in plant.ini and 'missing_mgt' is not defined in management.sch. The error message provides guidance on how to proceed. To resolve the problem, the user must add the missing entries in the respective SWAT+ input files. As mentioned above, SWATmeasR takes a 'snapshot' of the SWAT+ input files when a new SWATmeasR project is generated. To update the input tables in the SWATmeasR project, all input files from the project folder must therefore be reloaded into the SWATmeasR project using the function .\$reload_swat_inputs(). The user can then proceed loading the NSWRM definition file for land use.

Figure 3.6: Error message when entries in SWAT+ input files are missing which were defined in the land use type NSWRM definition file.

The example in Table 3.3 shows empty elements, for example all *landuse.lum* entries for contr_drn. An empty element means that the initial value of this element remains unchanged when implementing a land use NSWRM in an HRU. In the case of contr_drn this would mean that all *landuse.lum* entries remain unchanged and only a controlled drainage is added to the HRU. If a *landuse.lum* entry should be specifically deactivated with the implementation of a land use type NSWRM, this can be done with the label 'null'. This is shown in Table 3.3 for the definition of 'buffer' where tile should be set to 'null' in all cases where buffer strips are implemented.

Once the land use type NSWRM input csv file has been generated, it can be loaded into the SWATmeasR project as shown below for the example SWATmeasR project schoeps_240502:

```
> land_def_path <- './nswrm_definition/settings_land_use.csv'
> schoeps_240502$load_nswrm_definition(
     file_path = land_def_path,
    type = 'land use')
```
Definition of NSWRM type = 'wetland'

The definition of wetlands is prepared in a *.csv input file. Table 3.4 provides an example for a wetland input table. The definition of a wetland at least requires to define the ID of the HRU (hru_id) to which a wetland water storage should be added. Further, a lu_mgt entry can be defined if the wetland should have a different land use than the initial one when the wetland is implemented. By default, the connectivity of the land object where a wetland is implemented remains unchanged, and water and nutrient fluxes are routed to the neighbouring objects as initially defined. If instead water and nutrient fluxes should be routed directly to a channel after the implementation of a wetland, the channel to which fluxes should be routed can be defined with cha_to_id. The columns hru_ps to hru_frac correspond to the parameter entries in the SWAT+ input file hydrology.wet, while the columns rel to nut correspond to the columns with the same names in the file wetland.wet, which hold pointer labels to entries in other SWAT+ input files.

Table 3.4: Example wetland input table for loading wetland type NSWRMs into a SWATmeasR project.

All columns except the hru id are optional inputs and default values are implemented if an entry is kept empty or if the column is not provided at all. Table 3.5 provides an overview of the optional inputs together with the default values to which the parameters are set if they were not provided in the wetland definition input file.

Table 3.5: Default values and definitions of wetland parameters that can be defined in the wetland NSWRM input file.

* keep the initial landuse.lum entry for an HRU, ** keep the initial routing of the HRU, *** set those default values if provided in the respective input files, otherwise set 'null'.

Once the wetland input csv file is generated, it can be loaded into the SWATmeasR project as shown below for the example SWATmeasR project schoeps_240502:

> wetl_def_path <- './nswrm_definition/settings_wetland.csv'

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```
> schoeps_240502$load_nswrm_definition(
    file_path = wetl_def_path,
   type = 'wetland')
```
Definition of NSWRM type = 'pond'and type = 'constr_wetland'

The definitions of ponds and constructed wetlands (constr_wetland) are prepared in *.csv input files. Although the implementation of these two NSWRM types in a SWAT+ model setup is identical, their definition is done in separate input files that have the same structure. Table 3.6 provides an example for such an input table. The definition of a pond or constr_wetland requires at least the definition of the columns hru_id and cha_to_id. In contrast to wetlands, hru_id can include multiple HRU IDs which are merged into one reservoir object when the NSWRM is implemented. cha_to_id must always be a single channel ID to which the fluxes of the new reservoir object are routed. In addition, a cha_from_id can be defined if the new reservoir object should also receive fluxes from channel objects. In this case also the routing of the channel objects which were defined with cha_from_id is changed to route into the new reservoir object. This can for instance be useful if a pond or constructed wetland is integrated into the existing channel network. The columns area_ps to shp_co2 correspond to the parameter entries in the SWAT+ input file *hydrology.res*, while the columns rel to nut correspond to the columns with the same names in the file reservoir.res, which hold pointer labels to entries in other SWAT+ input files.

Table 3.6: Example pond or constr_wetland input table for loading pond or constr_wetland type NSWRMs into a SWATmeasR project.

The cha_from_id, the *hydrology.res*, and the reservoir.res parameters are optional inputs and default values are implemented if an entry is kept empty or if a column is not provided at all. Table 3.7 provides an overview of the optional inputs together with the default values to which the parameters are set if they were not provided in the wetland definition input file. If the user provides values for the areas and volumes of the new reservoir object, the function \$load_nswrm_definition() performs checks for the provided values. The provided areas area_ps, and area_es must be less than the sum of the areas of the replaced HRUs, as the new water surface cannot be larger than the replaced land surfaces. Further, area_ps must be less then or equal to the area_es and the vol_ps must be less then or equal to the vol_es, as the values at the emergency spillway water level cannot be less than they would be at principal spillway water levels.

* keep the initial routing of the channel objects, ** set those default values if provided in the respective input files, otherwise set 'null'. *** ha-m (hectare * metres) is equivalent to 10,000 m^3 .

Once the pond or constr_wetland input csv file has been generated, it can be loaded into the SWATmeasR project as shown below for the example SWATmeasR project schoeps_240502:

```
> # In case of ponds
> pond_def_path <- './nswrm_definition/settings_pond.csv'
> schoeps_240502$load_nswrm_definition(
    file_path = pond_def_path,<br>type = 'pond')
              = 'pond')
> #
> # In case of constructed wetlands
> cwtl_def_path <- './nswrm_definition/settings_constr_wetl.csv'
> schoeps_240502$load_nswrm_definition(
     file_path = cwtl_def_path,
    type = 'constr_wetland')
```


3.2.5 Definition of NSWRM locations

After all NSWRMs that can be implemented in a SWAT+ model setup were defined and loaded into the SWATmeasR project, the user must define the potential locations of the NSWRMs. The definition of the NSWRM locations is done in a *.csv locations input file and loaded into the SWATmeasR project with the function .\$load_nswrm_location(). Table 3.8 shows an example of an NSWRM locations input table. The input table requires the four columns id, name, nswrm, and obj_id. id defines the ID of an NSWRM which is later used as an identifier for the implementation of NSWRMs. name is the unique name of a measure location and can be defined freely by the user. This column is not used at the moment. nswrm refers to the entries in the definition files for the respective NSWRM types. The labels of the specific NSWRMs must be used here to identify the respective measures, for example buffer, hedge, or grassslope in the example of the SWATmeasR project schoeps_250502 when referring to the different land use type NSWRMs. obj_id refers to the IDs of the spatial objects (HRU IDs) in which an NSWRM is implemented. For management and land use type NSWRMs, multiple HRU IDs can be grouped together in one obj_id entry to define a single NSWRM location. When such an NSWRM is implemented, the change in management or land use is implemented in all of the defined HRUs at the same time. For wetland, pond and constr_wetland NSWRMs the obj_id must match the respective entry in the column hru_id in the wetland or pond (constr_wetland) definition file.

Table 3.8: Example for a NSWRM location definition table for loading measure locations into a SWATmeasR project.

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Once the NSWRM location csv file is generated, it can be loaded into the SWATmeasR project as shown below for the example SWATmeasR project schoeps_240502. Similar to loading NSWRM definitions, the optional argument overwrite is required in .\$load_nswrm_location() to overwrite an existing locations table if necessary.

> location_path <- './nswrm_definition/nswrm_location.csv' > schoeps_240502\$load_nswrm_location(file_path = location_path)

326 NSWRM implementation

Although the implementation of NSWRMs is not part of the generation of a SWATmeasR project, it is briefly outlined here. The NSWRM implementation is part of the SWAT+ workflow in the optimization routine, and the function calls to implement NSWRMs and to write modified SWAT+ input tables into the SWAT+ project folder will be executed in the R script SWAT.R (see also section 4.3.2). It is, however, good practice to perform tests with the NSWRM implementation, run SWAT simulations and analyse the impacts of the implemented NSWRMs for plausibility.

Once all NSWRMs and their potential locations were defined, they can be implemented in the corresponding SWAT+ model setup. The NSWRM implementation is always a two-stage procedure. First, the measures are implemented in the SWAT+ input tables stored in the SWATmeasR project in .data\$model_setup\$modified_inputs using the function .\$implement_nswrm(). In a second step, the changed SWAT+ input files which were modified in the SWATmeasR project are written into the SWAT+ project folder with the function .\$write_swat_inputs() and overwrite the initial input files.

The NSWRM locations are implemented by providing their id value with the argument nswrm_id in the function .\$implement_nswrm(). nswrm_id can be a single value if only one NSWRM location should be implemented, but can also be a vector of IDs if multiple locations should be implemented. In the code example below, the NSWRM locations of buffer_1 to buffer_4, grassslope_2, hedge_3, and pond_1 should be implemented, which were defined in the example in Table 3.8. From the table we can read the IDs of those NSWRM locations, which are nswrm id $= c(1:4, 8, 12, 19).$

 $>$ schoeps_240502\$implement_nswrm(swrm_id = c(1:4, 8, 12, 19))

The implementation of NSWRMs can be verified by looking into the table implemented_nswrms, which was generated when the measures were implemented. The table can be accessed as shown below:

The table shows all HRUs where NWRMs have been implemented. A closer look shows that the implementations of grassslope_2 (in HRU 122) and buffer_3 (in HRU 468) are missing. The reason for that is an overlap with the implementation of other NSWRMs in the same HRUs. HRU 122 was also affected by the implementation of hedge_3 and HRU 468 eventually was replaced by a pond, which is represented by the reservoir 156. As the example shows, there is a specific order in which NSWRMs are implemented between the different NSWRM types, but also within the same type. Figure 3.7 shows the order in which the NSWRMs are implemented. The NSWRM types are processed in the sequence management > land_use > wetland > constr_wetland > pond. Within each NSWRM type, the measures are implemented in the order in which they were provided in the NSWRM location table.

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Figure 3.7: Order of implementation of NSWRMs.

In the illustrated example, hedge_3 is implemented in HRU 122 after the implementation of grassslope_2 in the same HRU and therefore overrules the first implementation. pond_1 is implemented as the last measure in the shown sequence. It replaces the HRU 468 with a new reservoir and therefore overrules the implementation of buffer_3 in the same HRU.

Once the NSWRM locations were implemented, the changes in the SWAT+ model input files can be written into the SWAT+ project folder. This is done as follows:

```
> schoeps_240502$write_swat_inputs()
```
Once implemented, SWAT simulations can be executed and the simulated outputs can be analysed to assess the impacts of the implemented NSWRMs. After finishing the simulation, the original SWAT+ model setup can be restored with the function .\$reset(). With the reset all modified input files are overwritten by the original input files stored in .\$data\$model_setup\$original_inputs. It is strongly advised to always reset the SWAT+ project setup after the simulation and analysis of results is completed. Otherwise, there may be the risk that NSWRMs will remain in the model input files and the generation of a new SWATmeasR project for example would read the modified input files and would therefore start with a wrong initial model setup.


```
> schoeps_240502$reset()
#> 
#> Resetting input files in project folder... Done!
#> Resetting tables measR project ... Done!
```
4. Running the optimisation

python installation $4.1.$

CoMOLA is written in the script language python. Therefore, python must be installed on the computer in order to run CoMOLA, specifically python version 3.11.x. There are different approaches to install python, and it is very likely that a version (or several versions) of python are already installed on the computer on which CoMOLA is to be run. To minimise the risk of any CoMOLA/python related issues, a clean new installation of python and required python packages is recommended. The installation of python and required packages will be done using miniconda. miniconda is a lightweight python package manager (a minimum version of the well-known package manager anaconda) that allows users to easily install different versions of python and required python packages.

4.1.1 miniconda installation

Two approaches to installing miniconda are shown below, one that runs the installation automatically via Windows powershell and the second that uses the Windows installer. If it is OK to install the software in the computer's default programs folder, simply the command line approach can be used.

Installation from the command line

To run the installation from the command line, the Windows powershell must be started. It is started by clicking on the Windows Start button, typing "cmd" and pressing Enter (see Figure 4.1).

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Figure 4.1: Starting the Windows command prompt (powershell) from the Windows start menu.

A command prompt opens after pressing enter. The miniconda installation is executed in the command prompt by running the lines of code from the code box below. Just copy the few lines of code into the command line. The commands will execute the latest version of the miniconda installer. It downloads the executable file, installs it in the default program path and deletes the installer executable file after the installation.

```
curl https://repo.anaconda.com/miniconda/Miniconda3-latest-Windows-
x86_64.exe -o miniconda.exe
start /wait "" miniconda.exe /S
del miniconda.exe
```
Installation with the installer

Alternatively, the installer file can also be downloaded manually from the anaconda website. The latest Windows installer executable file can be retrieved from <https://docs.anaconda.com/free/miniconda/>. After downloading the installer, miniconda can be installed to the desired location on the computer's hard drive.

4.1.2. miniconda setup

The miniconda installation includes the installation of an own prompt, which is different to the Windows command prompt (cmd). To avoid registering the paths of conda and python in the Windows PATHs (which may cause troubles with other software) it is safer to work directly in the base environment of miniconda. The miniconda prompt can be started in the same way as the Windows command prompt, by clicking on the Windows start button but searching for the "Anaconda Prompt (miniconda3)" (see Figure 4.2).

Figure 4.2: Starting the Anaconda Prompt (miniconda3) from the Windows start menu.

python installation and installation of required python packages

By starting the Anaconda prompt, a new command window opens which should look similar as below (Figure 4.3). In brackets the cursor position shows (base) before the home path, which indicates that the user is currently working in the miniconda base environment.

Figure 4.3: The Anaconda Prompt (miniconda3) after starting.

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Due to the requirements of CoMOLA, the newest version of python 3.11x needs to be installed, which at the time of writing this document was version 3.11.9. To install specifically this python version the command below is executed in the Anaconda prompt:

conda install python=3.11.9

Executing the install command starts the python 3.11.9 installation, which should look similar to the screenshot below (Figure 4.4).

Anaconda Prompt (miniconda3) - conda install python=3.11.9					O	×
(base) C:\Users\schuerz>conda install python=3.11.9 Channels: - defaults						
Platform: win-64						
Collecting package metadata (repodata.json): done						
Solving environment: done						
## Package Plan ##						
environment location: C:\Users\schuerz\AppData\Local\miniconda3						
added / updated specs: $-$ python=3.11.9						
The following packages will be downloaded:						
package	build					
anaconda-anon-usage-0.4.4	py311hfc23b7f 100		30 KB			
$boltons-23.0.0$	py311haa95532 0	478 KB				
brotli-python-1.0.9	py311hd77b12b 7	310 KB				

Figure 4.4: The python installation process in the Anaconda Prompt (miniconda3).

During the installation process, you will be prompted to downgrade all previously installed python packages to the version 3.11 (which is not the latest python version). Pressing enter will continue the installation process.

CoMOLA depends on the python package numpy. numpy is not installed by default with the miniconda installation. python packages can be installed with the command conda install. The numpy installation is done by executing the code in the code box below in the Anaconda prompt. Again, the installation process will ask to proceed. Pressing enter continues the installation process.

```
conda install numpy
```
Checking the python and numpy installation

To start CoMOLA, the path to the correct python executable file (with the version 3.11.x) must be specified in the input file config.ini (see section 4.2). The path to the current python executable can be found with the command where in the Anaconda prompt (see example in Figure 4.5).

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```
where python
```


Figure 4.5: The python executable path is returned with the command where.

The python version can be checked as follows. If the installation was successful the python version 3.11.9 is shown (example in Figure 4.6).

Figure 4.6: Checking the version of the python executable.

Checking if numpy is installed can be done with the command conda list. The checking for numpy is performed as follows. A list of numpy related packages should be returned (see example in Figure 4.7).

CoMOLA file structure 4.2

After successfully installing python, the CoMOLA optimisation software is to be downloaded from GitHub (https://github.com/michstrauch/CoMOLA_SWATplus) or the UFZ GitLab: [https://git.ufz.de/optain/wp5-optimisation/comola.](https://git.ufz.de/optain/wp5-optimisation/comola) This section briefly describes the different folders and files of CoMOLA. Section 4.3 provides a detailed manual on how to set up and run the optimisation for a user-specific case study. The most important rule when using CoMOLA is not to change its file structure, including all file names (Figure 4.8).

Figure 4.8: CoMOLA file structure. Files in red fonts need to be added, replaced or adjusted by case studies.

The folder *input* contains two files that are automatically written when the optimisation is started. Users should not edit these files. *file_HRU.csv* specifies a vector of NSWRM sites to be included in the optimisation and the state of implementation at the start of the optimisation process. The vector represents the genome of the starting individual (i.e. individual 1 in the initial population) and by default contains the value of 1 for all genes (i.e. no NSWRM are implemented). The file worst_fitness_values.txt defines the fitness values to be used for infeasible individuals (individuals that violate certain constraints). Since we do not define any optimisation constraints in OPTAIN, this file is actually obsolete. However, the current version of CoMOLA requires such a file.

The *inspyred* folder contains modules from the inspyred python package (Garret, 2012), including optimisation algorithms such as NSGA-II (Section 2.4). Users should not deal with this folder.

The folder *models* must contain the SWAT+ model files and, if required, any other model that needs to be included for calculating the (fitness) values of the optimisation objectives. In OPTAIN, it is sufficient to include the SWAT+ model only. Users need to add their own SWAT+ project (txt) in a folder called $SWAT$ and make sure that a number of model requirements is met (further details in section 4.3). The folder *economic_model* contains the R scripts and input tables required to calculate the socio-economic performance indicators. At the time of writing this document, the files were still under preparation (Deliverable D4.5). They will be added as soon as possible. In addition, the *models* folder contains two R files (by default). The *calc_opt_indis.R* file defines all available functions for calculating the environmental performance indicators. If additional functions are required to calculate the optimisation objectives, the user must define them here or send a request to the authors of this report. The SWAT.R file must be adapted to specify the objectives used for the optimisation (see section 4.3).

The *output* folder is the directory where the log file (* *optimisation_log.txt*) and fitness values for each individual NSWRM plan tested in the optimisation (*_individuals_file.csv) are printed. Genomes and fitness values of the final Pareto optimal solutions are printed in the log file. The *output* file names contain a timestamp (dd_MM_yyyy_hh_mm_ss) which indicate the time at which the respective optimisation run was started. Users should access the results by running the CoMOLA_postprocessing.R file (section 4.4), which is provided in the output_analysis folder along with a file containing the postprocessing functions (functions_postprocessing.R).

In addition to the folders already presented, the CoMOLA folder contains the *init.R* file, which automatically writes the required files to the *input* folder, as well as a number of python files needed to run the optimisation (users should not bother with these). The only file in the main folder that is relevant for the user is the CoMOLA master file (*config.ini*). This file must be adapted by defining the paths to the R and python executables, as well as important optimisation parameters such as population size and maximum number of generations (see Section 4.3).

$4.3.$ Setting up and starting an optimisation run

At this point in the protocol we assume that the user has installed python, downloaded CoMOLA [\(https://git.ufz.de/optain/wp5-optimisation/comola\)](https://git.ufz.de/optain/wp5-optimisation/comola) and familiarised him or herself with the file structure of the optimisation tool (section 4.1 and 4.2). To set up an optimization run, three main steps must be taken:

- (1) Add your own txt folder
- (2) Adjust the SWAT.R script for your optimisation objectives
- (3) Configure the master file (config.ini)

4.3.1. Add your own txt folder

Create a copy of your SWAT+ project (txt folder) and add it to the models/SWAT directory. Note that the SWAT+ project folder must be named 'txt'. In the txt folder, make sure:

- 1. That the simulation period is correctly defined in the time.sim file. It is recommended to use the same period as used for the model calibration (but not more than 10-15 years to avoid exhausting model runtimes). Also make sure that the number of years to skip (parameter nyskip in the file print.prt) is set correctly for printing the simulation output (typically nyskip = 3).
- 2. That the model is sufficiently parameterised (e.g. landuse.lum parameters, initial soil nutrient values, correct operation schedules in the management.sch file corresponding to the defined simulation period).
- 3. That the model is sufficiently calibrated (i.e. a calibration.cal file with calibrated parameter values must be provided).
- 4. That the model prints only the relevant outputs required to calculate the optimisation objectives (irrelevant model outputs would consume unnecessary storage capacity and run time). If an optimisation objective requires channel outputs in daily resolution (e.g. low or high flow indicators), you must define the channel of interest (usually the outlet channel) in the object.prt file.
- 5. That you have removed unnecessary files and folders (SWAT+ output files can be extremely large).
- 6. That the SWAT+ master file (file.cio) lists all required input files, including those mentioned above (time.sim, cal_parms.cal, calibration.cal, object.prt).
- 7. That one SWATmeasR project file (<project_name>.measr) is included. The SWATmeasR project must contain appropriate settings for each of the NSWRMs considered. It is strongly recommended to check the plausibility of each individual NSWRM scenario prior to the optimisation.

The authors of this report thought several times that they had made all the settings in their own model, but then they were proven wrong. As a full optimisation run with SWAT+ models usually takes several weeks of time, double-checking is strongly recommended. A demo txt version from CS1 is available in the OPTAIN cloud (WPs & Tasks/WP5/SWAT+ model setups/CS1).

4.3.2. Adjust the SWAT.R script

The SWAT.R file is the script that is responsible for running SWAT+ within CoMOLA. When called, it implements the NSWRMs at individual sites within the study area according to the CoMOLA genome, runs the SWAT+ model and calculates the optimisation objectives. The optimisation targets are calculated based on a set of predefined indicator functions (stored in the *calc_opt_indis.R* file). Ensure that the correct indicator functions are used to calculate the objectives relevant to your case study. Table 4.1 lists all available indicators and their functions at the time of writing this report.

Table 4.1: Indicator functions available for the optimisation. Note that some functions require the variable x to be specified as an integer in square brackets; other functions require the indicator name (ind); and one function requires both. Functions using the HRU output files as input (e.g. ind_hru_wb_aa()) have the option of specifying the area considered in the calculation. Parameter *area* can be either 'basin' (considering all HRUs in the basin) or 'agr' (considering only cropland HRUs). The parameter *period* can be specified in function ind hru mon wb()to include only the months of interest (as integer values, e.g. c(5:9) for the months May to September) in the calculation.

Below are some examples of how to use these functions in the SWAT.R file. CoMOLA aims to maximise fitness values for all objectives. Therefore, if an objective is to be minimised (such as the average nutrient load or the frequency of high nutrient concentrations at the catchment outlet), the corresponding indicator function must be multiplied by -1. Please note that the parameter *path* must always be set to txt_path.

Example 1:

Optimisation objective 1 = Pload

- choose function ind_cha_aa()
- in file *print.prt*, indicate to print the average annual *channel_sd* output file
- define your channel of interest
- specify variable x in squared brackets (here $x = 3$)
- multiply with -1 as loads should be minimised

```
fit1 <- ind_cha_aa(path = txt_path,
                    channel = \text{chao926'} [3] * -1
```


Example 2:

Optimisation objective 2 = frequency of days with streamflow > lowflow threshold

- choose function ind cha day()
- define your channel of interest
- in file *object.prt*, indicate to print this channel's output to file *cha_day.out*
- define parameter *ind* (here: 'Q_low_days')
- define a meaningful lowflow threshold value (threshold_lowQ), such as the observed average minimum daily discharge of each year of the simulation period)
- specify variable x in squared brackets (here $x = 12$)

```
fit2 <- ind_cha_day(path = txt_path,
                     channel = 'cha0926',
                    ind = 'Q\_low\_days',
                    threshold_lowQ = 0.0344)[12]
```
Example 3:

Optimisation objective 3 = soil water (top 30cm) for period May to June in cropland

- choose function ind_hru_mon_wb()
- in file *print.prt*, indicate to print the monthly *hru_wb* output file
- define indicator of interest (here 'sw300')
- define period of interest (provide the numbers for the months of the year)
- define area of interest (can be either 'basin' for whole basin area or 'agr' for cropland HRUs only)
- as the period consists of multiple months (May and June), calculate the mean value of this period using mean(as.numeric())

```
fit3 <- mean(as.numeric(ind_hru_mon_wb(path = txt_path,
                                        ind = 'sw300',
                                       period = c(5:6),
                                       area = 'agr'))
```
Example 4:

Optimisation objective 4 = grain unit sum for the whole basin

- choose function ind bsn aa $crp()$
- define the crop types to be considered (here all crop types are used that are listed in R object *grain_units*)
- define the indicator of interest (here 'grain_units')
- define crop-specific grain units (here the R object grain_units is defined for the same-named function parameter); the R object *grain_units* with your crops of interest must be specified before defining the optimisation objective

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```
grain_units <- data.frame('wbar' = 1.163,
                         'csil' = 1.071,'wwht' = 1.209,
                         'wira' = 1.429,
                         'barl' = 1.163,
                         'akgs' = 0.682,'wiry' = 1.174,
                         'sgbt' = 1)fit4 <- ind_bsn_aa_crp(path = txt_path,
                        crop_sel = names(grain_units),
                       ind = 'grain_units',
                       grain_units = grain_units)
```
$4.3.3.$ Configure the master file (config.ini)

The config.ini file lists all parameters which are relevant to run an optimisation with CoMOLA. To set up the CoMOLA runs, the parameter sections [config_model] and [config_optimization_algorithm] have to be adjusted. In the section [config_model] the paths to the R and python executable files must be set $(\textit{file_path_R}$ and file_path_Python, respectively). An easy way to find the correct path to the R executable file (which is also used by RStudio) is to open RStudio and execute the command R.home('bin') in the console. This returns the path to the R executable, which must be assigned to the parameter file_path_R in the config.ini.

> R.home('bin') #> [1]"C:/Users/schuerz/AppData/Local/Programs/R/R-4.2.2/bin/x64"

The identification of the path to the correct python executable file was shown in the miniconda setup in section 4.1.2. The execution of the command where python in the Anaconda prompt (see Figure 4.6) returns the path to the python version 3.11.9 which was installed before. The path can be copied and assigned to the parameter file_path_Python in the config.ini.

Of the optimisation parameters, only two need to be adjusted, pop_size and max_generations.

pop_size defines the size of the population (i.e. the number (n) of individual NSWRM plans within a population). The population size should be large enough to ensure sufficient diversity between individuals, to accurately approximate the true Pareto front, and to prevent premature convergence to local optima by providing a broader genetic pool and more opportunities for crossover and mutation. Common practice suggests population sizes in the range of 50 to 500 for many optimisation problems (e.g. Coello Coello et al., 2007; Deb et al., 2002; Zitzler et al., 2001), but this can vary widely. Defining the population size requires consideration of problem complexity, computational resources, and empirical testing to find a balance that

ensures efficient and effective optimisation. As the SWAT+ models in OPTAIN require long computation times to evaluate the fitness of an individual NSWRM plan (~10 to 30 minutes for a 10-year simulation period), we do not have the time and computational resources for in-depth empirical testing in any of the case studies. Based on our experience from previous land use optimisation studies (e.g. Kamamia et al., 2022; Verhagen et al., 2018), we suggest setting *pop_size* to a value of 100. At the start of the optimisation, CoMOLA would then create 100 copies of the models folder in which the SWAT+ model can be run in parallel. Make sure that you have enough disc space in your CoMOLA environment. For the first test runs, pop_size should not be greater than 5.

max_generations defines the maximum number of generations (i.e. the number of iterations in which a new offspring population of NSWRM plans is generated and evaluated during the optimisation). While *pop_size* must be large enough to ensure good coverage of the search space, *max_generations* must be large enough to allow the process of refining and improving upon the existing solutions to enhance their fitness. Due to the limited resources mentioned above, we suggest limiting the number of generations to 150. This would result in a CoMOLA routine with 15,100 SWAT+ model runs, which could take 7 to 30 days to compute for an average-sized SWAT+ model, depending on the resources available (in particular the type of processor and the number of cores to be used for parallel processing). For initial testing, it is recommended to set *max_generations* to 2.

In summary, there are four settings to be made in the *config.ini* file:

- (1) $file_path_R =$ path to your R executable
- (2) $file_path_Python =$ path to your python executable (version 3.11.9)
- (3) $pop_size = 100$ (for initial tests $pop_size = 5$)
- (4) $max_generations = 150$ (for initial tests $pop_size = 2$)

$4.3.4$ Starting the optimisation (run_comola.bat)

The CoMOLA routine starts with a double click on the batch file run comola.bat. This opens the Windows command prompt and performs some internal checks on important optimisation requirements (e.g. if the correct version of the SWATmeasR R package has been used to build the SWATmeasR project). If any of the requirements are not met, an error message is printed to the file error_log.txt and the routine is aborted (Figure 4.9). All problems must be solved before the actual CoMOLA routine is started.

C:\windows\system32\cmd.exe

```
Initializing CoMOLA run...
Initialization resulted in errors! See 'error_log.txt' and fix reported issues before re-running.
Press any key to quit...
```
Figure 4.9: Prompt indicating errors at the start of CoMOLA.

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If there are no errors, the command prompt will ask for the number of cores to be used to run CoMOLA (Figure 4.10). It is strongly recommended to use a High Performance Computing (HPC) cluster for Windows (the current SWAT+ executable used in OPTAIN has not been compiled for Linux). An HPC would allow CoMOLA (actually the SWAT+ model) to run on more than 20 nodes. The optimal number of cores is equal to pop_size. Then all individuals of a generation would run in parallel. If no HPC is available, a single machine can be used. However, this machine should be powerful and not be used for other tasks while the optimisation is running.

C:\windows\system32\cmd.exe

Initializing CoMOLA $run...$ Start of CoMOLA routine Define the number of cores for running CoMOLA:

Figure 4.10: Prompt indicating a successful start of the CoMOLA routine.

Running SWAT+ models can take a while (be patient with a log text such as shown in Figure 4.11). Errors will be printed in the command prompt and in the *_optimisation_log.txt file. Post the error message on the UFZ GitLab or GitHub if you cannot fix the problem yourself. A successful run of CoMOLA will automatically close the command prompt.

BUT C:\windows\system32\cmd exe 30-05-2024 17:55:26 | Run external models ... 30-05-2024 17:55:26 | Start model C:\+PAPER_WORK+\Opti-Tool\CoMOLA_CS1_240503\models\SWAT\SWAT.R 30-05-2024 17:55:26 | Start model C:\+PAPER WORK+\Opti-Tool\CoMOLA CS1 240503\models 3\SWAT\SWAT.R 30-05-2024 17:55:26 | Start model C:\+PAPER WORK+\Opti-Tool\CoMOLA CS1 240503\models 2\SWAT\SWAT.R 30-05-2024 17:55:26 | Start model C:\+PAPER WORK+\Opti-Tool\CoMOLA CS1 240503\models 1\SWAT\SWAT.R 30-05-2024 17:55:26 | Start model C:\+PAPER_WORK+\Opti-Tool\CoMOLA_CS1_240503\models_4\SWAT\SWAT.R

Figure 4.11: Prompt indicating SWAT+ models running in parallel.

$4.4.$ Analysis of results

For each successful optimisation run, CoMOLA creates five output files in the output folder, of which only two files are relevant for the analysis of results: *_*optimisation_log.txt* (also called 'log file', * indicating the date and time of the start of the optimisation run) and *_individuals_file.csv (also called 'ind file').

The log file contains all the information printed to the command prompt during the optimisation and can be used to check for any errors that may have occurred. To do this, simply open the file in a text editor and search for the term 'Error'. At the end of this file, the user can find the final set of Pareto optimal solutions (fitness

values and genomes). The genomes and fitness values of all NSWRM plans tested in the optimisation can be found in the ind file.

As the output format in these files may not be convenient for users, it is recommended to access the optimisation results using the R script CoMOLA_postprocessing.R provided in the folder output_analysis. Running this file will extract and save the genomes and the fitness values of the Pareto optimal NSWRM plans in the files *pareto_genomes.txt* and *pareto_fitness.txt*, respectively. In addition, the hypervolume metric proposed by Zitzler and Thiele (1999) is computed to evaluate the different solution sets for each generation. Hypervolume is a widely accepted multi-objective performance metric that measures both convergence and diversity on a single scale without requiring knowledge of the true Pareto front for comparison (Jiang et al., 2014). It represents the volume in the objective space dominated by the set of solutions for a given reference point (here we use the origin of the coordinates). Thus, higher values of the hypervolume indicate that the solutions are closer to the true Pareto front and, at the same time, that they are more evenly distributed in the objective space (Jiang et al., 2014). The file CoMOLA_postprocessing.R can be used to plot the evolution of hypervolumes over generations. A converging curve (see an example in Figure 4.12) indicates sufficient exploration of the solution space. If the curve does not converge, this indicates that it is worth repeating the optimisation with a higher number of generations (parameter max_generations in the file *config.ini*).

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Eventually, the user can create a simple Pareto plot for the set of best solutions (see for example Figure 4.13). To do this, meaningful names should be defined for the different objectives at the beginning of the R script (corresponding to the indicator functions used for fit1, fit2, etc. in file SWAT.R, section 4.3).

Figure 4.13: Scatterplot of Pareto optimal solutions created with the CoMOLA_postprocessing.R file (example plot).

The R code to post-process (e.g. visualise, filter) Pareto optimal solutions will be extended in the next phase of the project and published in Deliverable D5.2 'Report & Code for post-processing of optimisation results'.

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5. References

- Chang, W., 2022. R6: Encapsulated Classes with Reference Semantics. https://r6.rlib.org, https://github.com/r-lib/R6/.
- Coello Coello, C. A., Lamont, G. B., Van Veldhuizen, D. A., 2007. Evolutionary Algorithms for Solving Multi-Objective Problems. Springer.
- Cord, A.F. et al., 2017. Towards systematic analyses of ecosystem service trade-offs and synergies: Main concepts, methods and the road ahead. Ecosystem Services, 28: 264-272. DOI:10.1016/j.ecoser.2017.07.012
- Deb, K. et al., 2002. A fast and elitist multiobjective genetic algorithm: NSGA-II. IEEE Transactions on Evolutionary Computation, 6(2): 182-197. DOI:10.1109/4235.996017
- Garrett, A., 2012. inspyred (Version 1.0.1) [software]. Inspired Intelligence. Retrieved fro[m](https://github.com/aarongarrett/inspyred) <https://github.com/aarongarrett/inspyred> [accessed 27.05.2014]
- Jiang, R. et al., 2021. A Hybrid Multi-Objective Optimization Method Based on NSGA-II Algorithm and Entropy Weighted TOPSIS for Lightweight Design of Dump Truck Carriage. Machines, 9(8): 156. DOI:10.3390/machines9080156
- Jiang, S. et al., 2014. Consistencies and contradictions of performance metrics in multiobjective optimization. IEEE Trans. Cybern. 44 (12), 2391–2404. DOI:10.1109/TCYB.2014.2307319
- Kaim, A., Cord, A.F., Volk, M., 2018. A review of multi-criteria optimization techniques for agricultural land use allocation. Environmental Modelling & Software, 105: 79-93. DOI:10.1016/j.envsoft.2018.03.031
- Kaim, A., Strauch, M., Volk, M., 2020. Using Stakeholder Preferences to Identify Optimal Land Use Configurations. Frontiers in Water, 2. DOI:10.3389/frwa.2020.579087
- Krzeminska, D. & Monaco, F., 2022. Tailored environmental and socio-economic performance indicators for selected measures. Deliverable D2.2 of the EU Horizon 2020 project OPTAIN. DOI:10.5281/zenodo.7050653
- Marval, Š., et al., 2022. SWAT+ and SWAP retention measure implementation handbook. Deliverable D2.3 of the EU Horizon 2020 project OPTAIN. DOI:10.5281/zenodo.11232719
- Memmah, M.-M. et al., 2015. Metaheuristics for agricultural land use optimization. A review. Agronomy for Sustainable Development, 35(3): 975-998. DOI:10.1007/s13593-015-0303-4
- Piniewski, M. et al., 2024: Assessment of NSWRM effectiveness under current and future climate at the catchment scale. Deliverable D4.4 of the EU Horizon 2020 project OPTAIN. DOI:10.5281/zenodo.11233621
- Schürz, C., 2024a: SWATbuildR: An object connectivity based SWAT+ model builder, R package version 0.1.17. URL:<https://git.ufz.de/schuerz/swatbuildr>

- Schürz, C., 2024b SWATfarmR: Simple rule based scheduling of management operations in SWAT, R package version 4.0.3. URL: https://github.com/chrisschuerz/SWATfarmR
- Schürz, C. et al., 2022. SWAT+ modeling protocol for the assessment of water and nutrient retention measures in small agricultural catchments. Deliverable D4.2 of the EU Horizon 2020 project OPTAIN. DOI:10.5281/zenodo.7463395
- Strauch, M. et al., 2019. Constraints in multi-objective optimization of land use allocation – Repair or penalize? Environmental Modelling & Software, 118: 241- 251. DOI:10.1016/j.envsoft.2019.05.003
- Zitzler, E., Laumanns, M., & Thiele, L., 2001. SPEA2: Improving the Strength Pareto Evolutionary Algorithm. TIK-Report, 103.
- Zitzler, E. & Thiele, L., 1999. Multiobjective evolutionary algorithms: a comparative case study and the strength Pareto approach. IEEE Trans. Evol. Comput. 3 (4), 257–271. DOI:10.1109/4235.797969