

Biotech Beer Brewing

Project status report

Dominik Schmidt
Jakob Wittmann

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1 Project status

1.1 Milestones

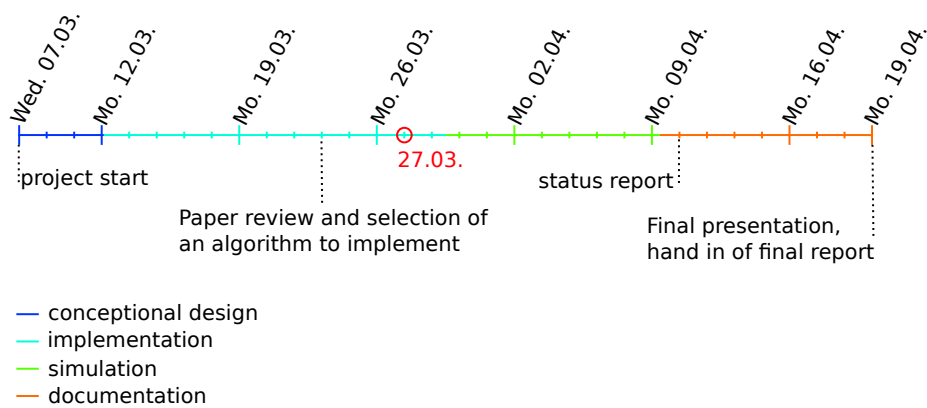


Figure 1: Project timeline

1.2 Progress

The progress of the main project processes in table 1 is illustrated in figure 2. The dashed colored lines show a simple estimated future progress of the processes. If the current trend holds on, the project will be finished approximately 12 working days after the final due date.

Table 1: Project processes

Process	Effort		start date	due date	progress
	Days	Percent			
conceptional design	3.1 d	10 %	07.03.	12.03.	100 %
implementation	12.4 d	40 %	12.03.	28.03.	50 %
research	4.65 d	15 %	12.03.	16.03.	90 %
concept	1.55 d	5 %	16.03.	20.03.	50 %
coding	6.2 d	20 %	20.03.	28.03.	10 %
simulation	7.75 d	25 %	28.03.	09.04.	17 %
research	4.65 d	15 %	28.03.	30.03.	50 %
setup	1.55 d	5 %	30.03.	02.04.	0 %
simulate setup	1.55 d	5 %	02.04.	09.04.	0 %
documentation	7.75 d	25 %	09.04.	17.04.	0 %
analyze results	1.55 d	5 %	09.04.	10.04.	0 %
prepare presentation	3.1 d	10 %	10.04.	16.04.	0 %
prepare report	3.1 d	10 %	16.04.	17.04.	0 %

2 Presentation of intermediate results

2.1 Considered DFBA approaches

As already discussed in the grant application, we can define the following basic requirements to the implementation of the simulator:

1. simulation of co-cultures
2. in a batch process
3. with following bacteria cultures: *Saccharomyces cerevisiae*, *Lactobacillus plantarum*
4. starting conditions are parameterized
5. simulation results include: metabolite densities and biomass of bacteria cultures after a defined time interval

Additional secondary goals are:

- generic or automated integration of bacteria models
- optimized or even graphical user interface
- enhanced model of the fermentation process for more accurate results

Zomorri et al. summarizes in [7] models to predict the behavior of bacteria cultures and introduces different categories. Three of them are especially interesting to be used in this project: *steady-state models*, *spatio-temporal models* and *dynamic models*. *Steady-state models* like compartmentalized community-level metabolic modeling can not be used since a common objective can not be generally assumed, as it would be in a purely competitive co-cultures. *Spatio-temporal models* have a very high computational effort as they take spacial

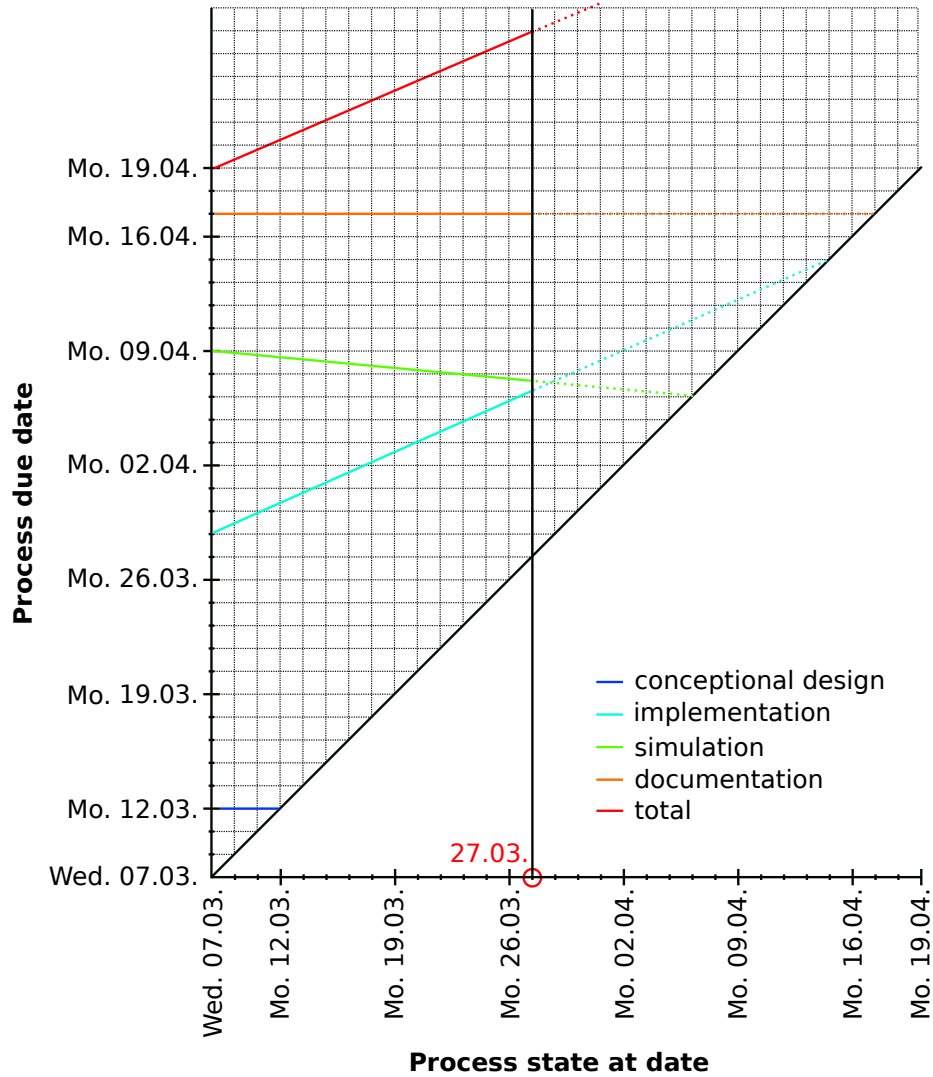


Figure 2: Project progress

and temporal varying bacteria densities into account. As the spacial aspect is not necessarily required in this project a more optimal approach shall be preferred. The remaining category of *dynamic models* is a well established method to simulate microbial co-cultures in batch processes and summarizes different extensions to dynamic flux balance analyses methods (DFBA)[7]. They use genome- scale models (GEM) to simulate the behavior of the bacteria cultures and add differential equations to model the external system dynamics.

Mehadevan et al. introduces two basic categories of DFBA approaches: *dynamic optimization approach* (DOA) and *static optimization approach* (SOA)[5]. In DOA a the linear programming problem (LP) which predicts the bacteria behavior is reformulized to a non-linear programming problem (NLP). This approach has a very high computational effort [2] compared to SOA and has

Table 2: Rating of considered DFBA methods

Method	comp. effort	impl. complexity	flexibility
dynamic optimization approach (DOA)	high	medium-high	?
static optimization approach (SOA)	low	low	low
direct approach (DA)	medium	low	medium
reformulation to a differential-glgebraic equation system	low-medium	high	?

only been used for relatively small GEMs with up to 13 modeled fluxes and 8 metabolites [4] [3].

Mehadevan et al. introduces SOA in [5] as follows: The simulation interval is divided into several intervals and the LP is solved for each of these time intervals dependent on the metabolite densities. The solution of the LP defines the bacteria growth and metabolite production at a certain point of time in the simulation time interval. These values are then used to solve the differential equations which models the external system dynamics. To solve the LP for the next time interval the new calculated, changed metabolite densities are used. This procedure is repeated until the end of the simulation time interval is reached. This approach makes use of the assumption that the cell internal dynamics are much faster than the external dynamics. In SOA the behavior of the bacteria is assumed to be constant during one time interval what leads to a linear approximation approach when solving the system of ordinary differential equations (ODE), similar to Euler-Cauchy methods.

Höffner et al. adds in [2] a further group, the *direct approach* (DA) which basically describes methods similar to SOA which uses an ODE solver instead of the Euler-Cauchy method. Due to the used ODE solver different numerical approximation methods can be used, not only the linear approximation. A good documented example for this group is the *Dynamic Multispecies Metabolic Modeling* framework by Zhuang et al. [6].

Henson et al. mentions a third group, *reformulation to a differential-glgebraic equation system* [1]. It shows also many similarities to SOA with the difference that the LP is reformulated but still solved as a LP embedded within the external ODE. The reformulated equation system makes it possible to enhance the efficiency of algorithm compared to SOA and DA [2].

2.2 Approach selection

The described DFBA methods in section 2.1 were rated based on the given information in the above mentioned papers, see table 2.

DOA can not be used due to its high computational effort and medium-high implementation complexity. The approach which uses *reformulation to a differential-glgebraic equation system* is currently available in matlab code and must be implemented in python in this project. Due to the high implementation complexity this approach will also be excluded. The remaining methods, SOA and DA, have similar ratings but as DA is more flexible as different ODE solvers can be used this approach seems more sustainable. Besides its flexibility the DA implementation DMMM by Zhuang et al. [6] can be publicly accessed and they

provide a good documentation which will facilitate the implementation in this project.

3 Outlook

Due to unexpected high effort in the research of simulation algorithms and technical problems the project progress is not as fast as expected (see figure 2). It is assumed that the technical problems were a single event and will not occur again, so this will not lead to further delay. As the high effort part in research for simulation algorithms is done, it can also be assumed that this will not cause further delay, too. Under these assumptions (no further delay in all processes) the project delay can be reduced from 12 to 6 days.

To catch up the remaining delay, a higher focus was set to choose a simulation algorithm with less implementation complexity. Furthermore, if this is not enough to catch up the delay, the complexity of the later simulations can be further reduced.

The subsequent steps include:

- research on modeling uptake limits dependent on metabolite densities in the substrate using Michaelis–Menten kinetics
- mathematical formulation of the model
- implementation of the simulation algorithm

References

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