

COMBINE 2024 - Conference of the COmputational Modeling in Blology NEtwork

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Abstract

The COMBINE consortium is an initiative to coordinate the development of community standards and formats for computer models in biology and medicine. COMBINE members organize COM-BINE as an annual conference at varying places. This conference is aimed at scientists at all career levels interested in using and developing these standards. It also provides a platform for exchange, discussion, and interactive experimentation and learning.

From September 1 to 5, the COMBINE 2024 conference took place at the University of Stuttgart, Campus Stuttgart-Vaihingen. The event was held as an in-person workshop-style event with the opportunity for remote participation during break-out sessions to enable broad community participation. COMBINE 2024 has taken place as a satellite event of the Virtual Physiological Human (VPH) 2024 Conference, which also took place in Stuttgart from September 4-6, 2024. The event's co-location was advantageous and promoted exchange and collaboration between the two scientific networks.

This year's COMBINE was co-hosted by the Stuttgart Cluster of Excellence EXC2075 "Data-Integrated Simulation Science (SimTech)". SimTech is an engineering-driven cluster that develops and applies multi-scale computational models and simulation schemes in various fields. Developing standards and workflows to enable and facilitate seamless management, exchange, and reuse of models and data is an important topic in the cluster. Moreover, the development of research software and strategies for long-term maintenance for the community are intensively discussed in the cluster across different communities and application fields. We believe that the long-term maintenance of models and computational tools and, along with this, a broad usage requires community efforts. In this respect, we see the COMBINE consortium as a successful role model in the Systems Biology field. On the other hand, SimTech researchers have also successfully developed software tools, e.g., for model coupling (preCICE) or efficient simulations of engineering-driven models and multi-scale models (Dynamore) and in biocatalysis (EnzymeML). The idea was to enable an exchange between COMBINE and SimTech and lively discussions, which worked out nicely.

Keywords: Reproducibility; Standards; Formats; Systems Biology, COMBINE; Stuttgart

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Invited talks 1

1.1 The EnzymeML framework: improving efficiency and quality of biocatalytic science

Jürgen Pleiss, Jan Range, and Max Häußler, University of Stuttgart, Germany

Biocatalysis is entering a promising era as a data-driven science. High-throughput experimen-6 tation generates a rapidly increasing stream of biocatalytic data, which is the raw material for 7 mechanistic and data-driven modeling to design improved biocatalysts and bioprocesses. How-8 ever, data management has become a bottleneck to progress in biocatalysis. In order to take 9 full advantage of rapid progress in experimental and computational technologies, biocatalytic data 10 should be findable, accessible, interoperable, and reusable (FAIR). The EnzymeML framework 11 (https://github.com/EnzymeML) provides reusable and extensible tools and a standardized data 12 exchange format for FAIR and scalable data management in biocatalysis (Range et al., 2022). 13 To enable storage, retrieval, and exchange of enzymatic data, the XML-based markup language 14 EnzymeML has been developed (Pleiss, 2021). An EnzymeML document contains information 15 about reaction conditions and the measured time course of substrate or product concentrations. 16 Kinetic modelling is performed by uploading EnzymeML documents to the modelling platforms 17 COPASI or PySCeS or by using the JAX platform. The rate equation and the estimated kinetic 18 parameters are then added to the EnzymeML document. The EnzymeML document containing 19 the experimental and the modelling results is then uploaded to a Dataverse installation or to the 20 reaction kinetics database SABIO-RK. The workflow of a project is encoded as Jupyter Notebook, 21 which can be re-used, modified, or extended The feasibility and usefulness of the EnzymeML tool-22 box was demonstrated in six scenarios, where data and metadata of different enzymatic reactions 23 are collected, analysed, and uploaded to public data repositories for future re-use (Lauterbach 24 et al., 2023). FAIRification of data and software and the digitalization of biocatalysis improve the 25 efficiency of research by automation and guarantee the quality of biocatalytic science by repro-26 ducibility4. Most of all, they foster reasoning and creating hypotheses by enabling the reanalysis 27 of previously published data, and thus promote disruptive research and innovation. 28

Reproducible digital twins for personalized liver function assessment 1.2

Matthias König, Humboldt-University Berlin, Germany

Essential prerequisites for the practical application and translation of computational models 32 include: (i) reproducibility of results; (ii) model reusability and extensibility; (iii) data availability; 33 and (iv) strategies for model stratification and individualization. Here, we present a modeling workflow built around these foundational prerequisites, with a focus on liver function tests. Despite the 35 paramount significance of liver function assessment in hepatology, reliable quantification remains 36 a clinical challenge. Dynamic liver function tests offer a promising method for non-invasive in vivo 37 assessment of liver function and metabolic phenotyping. By leveraging whole-body physiologically-38 based pharmacokinetic (PBPK) models, we're simulating these tests and positioning PBPK models 39 as digital twins for metabolic phenotyping and liver function assessment. To develop and validate 40 our models, we established the open pharmacokinetics database, PK-DB, containing curated data 41 from 600+ clinical studies (Grzegorzewski et al., 2021b; Grzegorzewski et al., 2021a). Our models 42 are individualizable and stratifiable, enabling simulation of lifestyle factors and co-administration 43 effects on drug metabolism. Our models have been instrumental in clinical scenarios: from predicting individual outcomes post-hepatectomy (Köller et al., 2021b; Köller et al., 2021a) to discerning 45 the impact of CYP2D6 gene variants on liver function tests (Grzegorzewski et al., 2022). These 46 models are constructed hierarchically, describing metabolic and other biological processes in organs 47 like the liver and kidneys, seamlessly integrated with whole-body physiology. Notably, all models 48 and data are readily available and reproducible for reuse, encoded in the Systems Biology Markup 49 Language (SBML) (Keating et al., 2020). We will provide an overview of these PBPK models 50 and demonstrate how SBML and FAIR principles can facilitate model development, coupling, and 51 reuse. 52

preCICE – A General-Purpose Simulation Coupling Library 1.353 Benjamin Uekermann, University of Stuttgart, Germany 54 55

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preCICE (https://precice.org/) is an open-source coupling software for partitioned multi-56 physics and multi-scale simulations including PDE-PDE and PDE-ODE coupling. Thanks to 57 the software's library approach (the simulations call the coupling) and its high-level API, only 58 minimally-invasive changes are required to prepare an existing (legacy) simulation software for 59 coupling. Moreover, ready-to-use adapters for many popular simulation software packages are 60 available, e.g. for OpenFOAM, SU2, CalculiX, FEniCS, and deal.II. For the actual coupling, 61 preCICE offers methods for fixed-point acceleration (quasi-Newton acceleration), fully parallel 62 communication (MPI or TCP/IP), data mapping (radial-basis function interpolation), and time 63 interpolation (waveform relaxation). Today, although being an academic software project at heart, 64 preCICE is used by more than 100 research groups in both academia and industry. In this presen-65 tation, I introduce the basic concepts of preCICE and discuss existing and potential applications 66 in biology. 67

1.4 Improving Curation: Biomodels and Annotation

Lucian Smith, Herbert Sauro, John Gennari, David Nickerson, S. Malik-Sheriff Rahuman, V. N. Nguyen Tung, University of Washington, USA

The BioModels Database has over 1000 curated models from published papers. Curators at 72 the EBI ensure that the model can be used to reproduce at least one figure from the paper, and 73 extensively annotate the model as well. However, until the advent of SED-ML, it was impossible to 74 store what the curator did to reproduce the model in a standard format, and until more widespread 75 use of SED-ML, it was impossible to reliably validate any SED-ML that was produced. The Center 76 for Reproducible Biomedical Modeling has produced new SED-ML interpreters and validators that 77 have bridged this gap, and we have partnered with the EBI to 'retro-curate', as far as possible, the 78 curated branch of BioModels, to include validated SED-ML, which we have then tested using the 79 SED-ML interpreters on multiple simulation engines. In addition, we have extended the Antimony 80 modeling language, and present the Antimony Web Editor, with particular features useful for 81 adding curation of species, reactions, and parameters. 82

1.5 CompuTiX: A library for agent based modeling (not only) at a tissue-scale

Jiří Pešek, Jules Dichamp, Peter Kottman, Boulitrop Charles, Dirk Drasdo, INRIA, team SIMBIOTX, France

In recent years, many studies have shown that the tissue microarchitecture along with the me-87 chanical environment has a crucial yet poorly understood impact on the biological processes inside 88 living tissues This have a significant impact on progression of any potential disease or treatment. 89 The limitations of in-vivo imaging techniques together with the small scale and isolated nature of 90 many in-vitro experiments, makes these systems a suitable candidate for in-silico approach, where 91 initial in-vitro experiments can be used to formulate and tune the underlying models and in-vivo 92 imaging is then used to generate a patient specific setup. In particular, an agent based models, 93 where the global effect is achieved by interaction between many, relatively simple, entities, are 94 suitable to capture the spatial and behavioral heterogeneity and complexity of living tissues. In 95 this talk we will present a new open-source computational library, CompuTiX, suitable for agent 96 based simulations of tissues, organoids and more. We will split the talk into two parts. In the 97 first part we will briefly introduce basic bio-physical models starting from simple center based 98 models to more complex models like deformable cell model. In the second, more technical, part 99 we will discuss the architecture of the library, design choices, trade-offs and challenges in our goal 100 to provide a versatile and extensible platform for agent based simulations. 101

1.6 MeDaX – two years towards bioMedical Data eXploration

Judith AH Wodke, University Medicine Greifswald

Research based on clinical care data is gaining attention across the world. However, the quality of clinical care data is generally not maximised for research purposes. Instead, according to economic principles, medical staff and time costs are commonly minimised, rendering the enrichment with sufficient metadata for easy data reuse at least challenging. In addition, a heterogeneous landscape of laws concerning medical data reuse on national, state, and county levels make (international) interoperability an ambitious aim. The MeDaX project was initiated about

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two years ago and its underlying idea presented at COMBINE 2022: connect and semantically 111 enrich highly diverse clinical and other biomedical data in knowledge graphs (KG) to design, im-112 plement, and use graph technologies for innovative data exploration. The MeDaX-KG prototype 113 has been designed and implemented building on the BioCypher framework to harmonise biomedi-114 cal knowledge graphs and using synthetic patient data. The proof of concept pipeline consists of 115 i) a FHIR input adapter, including an optimisation module for the generically generated graph 116 structure, ii) a semi-automatic data schema generation based on the BioLink ontology, and iii) 117 the visualisation of the resulting MeDaX-KG using Neo4j. Currently, the pipeline is improved, a 118 user interface is implemented, and the first pilot in a german university clinic's data integration 119 center is set up while the first stable release is prepared. To particularly tackle the gaps between 120 the different scientific domains in medical informatics, we got involved in or coordinate several 121 community projects. The international MIRAPIE community project aims to propose a prove-122 nance standard for biomedicine by defining a MInimal Requirements for Automated Provenance 123 Information Enrichment guideline (https://codeberg.org/MIRAPIE/MIRAPIE). Participating in 124 the BioCypher project, we adopted and are currently adapting the Biomedical Resource Ontol-125 ogy (BRO) (https://github.com/biocypher/biomedical-resource-ontology) to FAIRify our 126 own software but also to allow the better classification of biomedical data. Within the Medical 127 Informatics Initiative (MII) Germany we are coordinating the FAIRification of the MII core data 128 set (https://github.com/medizininformatik-initiative). In summary, we are aligning sev-129 eral interdisciplinary efforts towards exploration of high quality clinical care data for biomedical 130 research. 131

1.7 Reproducible tools for dealing with highly variable data

Nicole Radde, University of Stuttgart, Germany

In the biomedical context, data is often sparse, and replicates show a high variability. This 135 is because complex procedures, costs, and ethical aspects constrain measurements. Sparsity and 136 high variability pose a challenge for modeling, especially when building models aiming to capture 137 quantitatively dynamic responses. Here, we present two complementary approaches we developed 138 in our group to deal with sparse and variable data. Bayesian Modeling of Time Series Data (Bay-139 ModTS) uses a Bayesian approach and a simulation model to process sparse and highly variable 140 serial data (Höpfl et al., 2024). BayModTS can be used to quantify uncertainty in the observed 141 process or as a noise filtering approach, as we will demonstrate with selected examples. Second, 142 Eulerian Parameter Inference (EPI) formulates the parameter estimation problem for a simulation 143 model from experimental data as a stochastic inverse problem and infers a parameter distribution 144 that can reproduce the variability of the input data (Wagner et al., 2024). Both approaches are 145 implemented as documented software packages that use standards such as SBML (Keating et al., 146 2020) or PEtab (Schmiester et al., 2021). In my talk, I will briefly explain our methods and discuss 147 the current challenges regarding reproducibility and FAIR principles from a modeler's perspective. 148

1.8 The past, present and possible futures

Herbert Sauro, University of Washington, USA

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It has almost been 25 years since Hiroaki Kitano initiated the development of SBML as part 152 of the ERATO project. Together with Bolouri, Doyle, Finney, Hucka, myself and a number of 153 key stakeholders (who continue to meet at COMBINE), we published the first draft and software 154 support libraries for the SBML specification. Around the same time we also saw the publication of 155 the specification for CellML that was a more mathematically oriented proposal. What resulted was 156 most unexpected, the emergence of a new vibrant ecosystem which stimulated further development, 157 created a host of new ancillary standards as well as the indispensable BioModels repository. That 158 ecosystem still exists today. In this talk I will review what I feel remains to be done or is incomplete, 159 what new modeling challenges we face, and describe what the center of model reproducibility in the 160 US is doing in terms of software provision. In particular I will describe a number of new client-based 161 web tools and desktop apps. The client-based tools are unusual in that they can be hosted from 162 any free basic server such as a GitHub, Neocities or Cloudflare page. This makes such apps very 163 low maintenance and tend to persist long after funding stops. Examples from our center include a 164 model annotation (AWE) platform, a simple model checking app (ratesb), a high speed BioModels 165 cache, a reproducibility portal, a model verification service, a new SBML/Antimony web utility, 166 a Biosimulators/Biosimulations repository, a new SBML compliant desktop app, a number of new 167

Python packages for network visualization, a new desktop network editor (Alcuin), new extensions to Antimony (See talk by Lucian Smith), a standard protocol for multi-scale modeling (See talk by Eran Agmon), and the first model credibility hackathon held this summer.

1.9 Computational design of biological receivers using multi-scale models and data standards

Goksel Misirli, Keele University, UK

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Engineering genetic regulatory circuits that sense external molecules and respond is essential for 175 developing diverse biological applications. As the complexity of designs increases, a model-driven 176 design process becomes desirable to explore large design spaces that involve different biological 177 parts and parameters. Moreover, the amount of these molecules reaching a receiver is usually 178 assumed to be constant, and the diffusion dynamics and the interference caused by late-arriving 179 molecules and the cellular dynamics are often not integrated. Additionally, each molecule type 180 may represent a single biological signal and be unsuitable for encoding and decoding multiple 181 data bits. Here, we present the virtual parts repository, a computational framework that provides 182 modular, reusable and composable models. The framework facilitates automating the design of 183 predictable applications via simulations. It builds on the Systems Biology Markup Language 184 to model cellular behaviour and the Synthetic Biology Open Language to capture the details of 185 genetic circuits. We then extend this automation approach to design the end-to-end transmission 186 of signalling molecules from a transmitter to cellular receivers for multi-bit data communications. 187 The resulting framework can be used to understand the cellular response for a sequence of custom 188 data bits, each representing a group of molecules released from a transmitter and diffusing over a 189 molecular channel. The framework validates and verifies various communication parameters and 190 identifies the best communication scenarios. We also present a novel algorithm to minimise signal 191 interference by employing equalisation techniques from communication theory. Our data standards-192 enabled and multi-scale modelling workflow combines engineering genetic circuits and molecular 193 diffusion dynamics to encode and decode data bits, design efficient cellular signals, minimise noise, 194 and develop biologically plausible applications. 195

1.10 Networks, simple models and model diversity in the description of biological 196 systems

Marc Hütt, Constructor University Bremen, Germany

My talk will address three distinct, but interrelated, topics: (1) networks as structural models to interpret high-throughput data; (2) the distinction between mathematical models and their computer implementations; (3) simple models vs. complicated, parameter-rich models.

Systems biology and systems medicine frequently use network-based strategies for data interpretation and data contextualization. These methods, at times, lack standardization and comparability. Here I briefly discuss, how such methods work, and which implicit hypotheses are associated with them.

The formal representation of a mathematical model is often incomplete, compared to the details required for an implementation of the model to run numerical simulations. Implementation differences can in principle lead to drastically different results. For the case of models of excitable dynamics, I illustrate this point, showing that even the simplest models can display such implementation differences.

Lastly, residing on the topic of simple models, I briefly draw the attention to the co-existence of parameter-rich and simple models of biological systems, outlining a few pros and cons and caveats. ²¹²

2 Lightning Talks

2.1 Computational Model Development Using SBML: sbmlutils, sbm4humans, cy3sbm4s

Matthias König, Humboldt-University Berlin, Germany

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The Systems Biology Markup Language (SBML) (Keating et al., 2020) is recognized as the standard framework for representing and exchanging complex mathematical models in biological 219

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systems research. SBML facilitates the depiction of a diverse array of biological phenomena, encompassing metabolic networks, signaling pathways, and regulatory networks. It is versatile enough to handle models ranging from simple individual processes to intricate multi-scale representations.

One of the primary challenges faced by newcomers in computational biology is the encoding 223 and development of ordinary differential equation (ODE) models within the SBML framework. 224 Addressing this hurdle, we introduce two innovative Python tools: sbmlutils (https://github. 225 com/matthiaskoenig/sbmlutils), sbml4humans (https://sbml4humans.de), and the Cytoscape 226 application cy3sbml (https://github.com/matthiaskoenig/cy3sbml). These tools collectively 227 streamline the process of SBML model creation, enhancing both the programmatic aspect and 228 the user experience. Specifically, something facilitates the programmatic construction of SBML 229 models, while sbml4humans generates user-friendly reports for model interpretation. Furthermore, 230 cy3sbml integrates with Cytoscape to offer advanced visualization capabilities, thereby augmenting 231 the comprehension and analysis of SBML-encoded models. 232

These advancements significantly contribute to the ease of SBML model development and ²³³ interpretation, fostering greater accessibility and understanding for those entering the field of ²³⁴ computational systems biology. ²³⁵

2.2 Utilizing Nix for rapid BayModTS development

Simon Hauser and Fritz Otlinghaus, Helsinki Systems, Germany

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BayModTS, a Python project for FAIR Bayesian Modelling of Time Series workflows, has a 239 complex setup that requires users of that software to install and compile multiple Python packages 240 that have native C dependencies (Höpfl et al., 2024). This is a complex endeavor and currently 241 it is not possible to fully resolve these issues using poetry install. We present a solution that 242 utilizes a general purpose package manager called Nix, that guarantees that a package and all 243 its dependencies can be built reproducibly. This package manager can be used to build all kinds 244 of software packages, including C libraries and Python packages, which we need to realize our 245 solution. We utilize prepackaged system and Python dependencies already made available by the 246 Nix community to build new, complex packages, like libsbml, libroadrunner, tellurium, and others, 247 to realize a portable and reproducible local development environment. In the end, we automate 248 our solution by using a well established GitHub continuous integration solution that builds all 249 packages and makes them available via HTTP using a binary cache. This can be used so that 250 packages no longer have to be build locally and can be downloaded from CI instead, improving the 251 onboarding process for new team members and simplifying collaborations for external researchers. 252

2.3 Biological and Biophysics Simulation in Tissue Forge

T.J. Sego, University of Florida, USA

Tissue Forge is open-source simulation software for interactive particle-based physics, chemistry 256 and biology modeling and simulation. Tissue Forge allows users to create, simulate and explore 257 models and virtual experiments based on soft condensed matter physics at multiple scales, from the 258 molecular to the multicellular, using a simple interface. While Tissue Forge is designed to simplify 259 solving problems in complex subcellular, cellular and tissue biophysics, it supports applications 260 ranging from classic molecular dynamics to agent-based multicellular systems with dynamic pop-261 ulations. Tissue Forge users can build and interact with models and simulations in real-time and 262 change simulation details during execution, or execute simulations off-screen and/or remotely in 263 high-performance computing environments. Tissue Forge provides a growing library of built-in 264 model components along with support for user-specified models during the development and appli-265 cation of custom, agent-based models. Tissue Forge includes an extensive Python API for model 266 and simulation specification via Python scripts, an IPython console and a Jupyter Notebook, as 267 well as C and C++ APIs for integrated applications with other software tools. Tissue Forge sup-268 ports installations on Windows, Linux and MacOS systems and is available for local installation 269 via conda. The talk complements a tutorial at COMBINE 2024 that intends to introduce the basic 270 concepts, modeling and simulation features, and some relevant modeling applications of Tissue 271 Forge through guided simulation scripting. 272

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Sheriff Rahuman, EMBL-EBI, UK, and T.J. Sego, University of Florida, USA

Multi-approach Multi-scale (MAMS) modelling represents a cutting-edge method for modelling 276 and analysis of biological systems, leveraging an integrated suite of diverse modelling frameworks. 277 This multi-approach modelling will encompass a combination of diverse modelling formalisms, such 278 as ordinary differential equations (ODE), partial differential equations (PDE), logical, constraint-279 based, and agent-based models across multiple scales. These models are intricately tied together 280 to facilitate complex simulations. During the dedicated breakout sessions at Harmony 2021, COM-281 BINE 2021, and HARMONY 2024, we delved into the existing state-of-the-art technologies and 282 standards, including SBML and SED-ML, and their support for multi-approach modelling. These 283 discussions also illuminated the current challenges and gaps within the field. For COMBINE 2024, 284 our objective is to further this conversation by identifying published MAMS models and bringing 285 together the community to enable the creation of novel standards or the enhancement of existing 286 COMBINE standards to support MAMS. This effort aims to foster interoperability and support 287 the rapidly evolving paradigm of MAMS modelling. The talk complements a breakout session at 288 COMBINE 2024 that intends to continue the described work and invite new collaborators to join 289 continuing efforts.

2.5 Continuing Work Towards Reproducible Stochastic Biological Simulation

T.J. Sego, University of Florida, USA, and Sheriff Rahuman, EMBL-EBI, UK

Stochastic simulations are commonly used to quantitatively or semi-quantitatively describe the 294 dynamics of biological systems. At various scales and in multiple applications, stochastic simula-295 tion better reflects observed biological processes and robustness. Various methods are widely used 296 to incorporate stochasticity into biological simulation, such as the Gillespie stochastic simulation 297 algorithm for systems biology modeling, stochastic Boolean networks for network modeling, and 298 the Cellular Potts model methodology for multicellular modeling. Proving reproducibility of sim-299 ulation results is critical to establishing the credibility of a model. To this end, BioModels, the 300 largest repository of curated mathematical models, tests and reports the reproducibility of simula-301 tion results for all submitted models when possible. A recent study showed that about 50% of the 302 deterministic ordinary differential equation models on BioModels could not be reproduced when 303 applying criteria for reproducibility to the information provided in their associated publication, 304 reflecting a current crisis of reproducibility. Furthermore, there are no well-accepted metrics or 305 standards for reproducing stochastic simulation results, thus perpetuating the crisis of reproducibil-306 ity for a broad class of biological models. This lightning talk survey recent progress to establish an 307 accepted framework for testing the reproducibility of stochastic simulations in biological modeling. 308 The talk will provide a brief overview of recent progress towards defining quantitative measures to 309 determine whether stochastic simulation results can be reproduced, and when results have been 310 reproduced. The talk complements a breakout session at COMBINE 2024 that intends to continue 311 the described work and invite new collaborators to join continuing efforts. 312

2.6 Morpheus model repository: Experiences with reproducible multi-cellular models

Lutz Brusch, Jörn Starruß, Diego Jahn, and Robert Müller, Technische Universität Dresden, Germany 316

Collaborative modeling and simulation become increasingly important for studying self-oganization₃₁₈ patterning, morphogenesis and disease processes from the intracellular to the tissue and organ 319 scales. To support collaborations, we have developed the Morpheus model repository (https: 320 //morpheus.gitlab.io/models). This model repository is an open access and citable platform 321 for publishing, sharing and archiving multi-scale and multi-cellular models that are encoded in the 322 model description language MorpheusML (https://doi.org/10.25504/FAIRsharing.78b6a6). 323 We will explore statistics and examples of the usage of the Morpheus model repository. Differ-324 ent simulators like Artistoo (https://artistoo.net/converter.html) and Morpheus (https: 325 //morpheus.gitlab.io) can process MorpheusML models from the repository. Among them, the 326 model editor and simulator Morpheus is open source and allows to develop multi-scale models in 327 a modular manner and manage the entire workflow through a user-friendly GUI. Moreover, Mor-328 pheus is SBML-compliant, supports simulations based on experimental data, e.g. segmented cell 329 configurations, and is integrated with the FitMultiCell toolbox for robust and efficient parameter es-330 timation of stochastic models (https://gitlab.com/fitmulticell/fit, https://doi.org/10. 331 1093/bioinformatics/btad674). Importantly, the strict separation of the model description in 332 MorpheusML from any solver code allows to readily reproduce model results locally in different (fu-333 ture) versions of the simulation software. Beyond reproduction of published results, MorpheusML 334 models can easily be extended (copy-paste parts between models) and merged among each other, 335 thus vitalizing model reuse and exchange. 336

2.7 Modeling and simulation using industrial standards Modelica, FMI and web components.

Tomas Kulhanek, Charles University, Prague, Czech Republic

We use industrial standard Modelica to express complex models of human physiology (Mateják 341 et al., 2014; Ježek et al., 2017). Recently we have published enabling technology that allows to ex-342 port complex models in standard functional mockup interface API (FMI) as a web component to be 343 integrated with other web standards and technologies to create modern web application (Kulhanek 344 et al., 2023) (https://bodylight.physiome.cz). Thanks to it the models does not necessarry 345 need to be implemented in Modelica language, but and standard FMI needs to be implemented by 346 other standards to compute model derivatives and do simulation step using a prefered numerical 347 method. 348

In the exemplar case report of metabolic disorder we will demonstrate the process of creating 349 a component models, export them as web component and integrate with other web components or 350 web standards to create interactive application. Model implementation of cardioavscular system, 351 respiratory and blood gas exchange in Modelica will be used. Co-simulation and enriched with 352 chart and numbers presented in virtual monitor of vital signs will control application flow. The 353 foundation of technologies are published with open source license and thanks to chain of scientific 354 and/or industrial standards and tools. The resulting simulator can be executed on any device with-355 out the need to install special software. Platform needs only modern web browser and supported 356 are Windows/Linux/iOS computers, mobile phones, tablets, virtual reality headset, etc. Exemplar 357 application with accompanied learning material to learn pathophysiology of metabolic disorder is 358 available online at https://egolem.online/dka/. 350

2.8 A Standardized Protocol for Integrative, Multiscale Modeling

Ion Moraru, and Eran Agmon, University of Connecticut, USA

We are developing a standardized protocol for multi-algorithmic model composition, based on 363 standardized schemas for process interfaces, composition patterns, and orchestration patterns. This 364 will provide the foundation for robust infrastructure for systems biology models. The BioSimu-365 lators project aims to establish this protocol, ensuring reproducibility, tool compatibility, and 366 "plug-and-play" integration of new processes and data. Software tools built around these schemas 367 can include databases, applications, graphical user interfaces, and simulation tools, supported by 368 both local and remote operations, such as containerized and web-based services. By aligning 369 with existing standard formats like SBML and CellML, standard formats for spatial models, and 370 multi-cellular models, the protocol can foster a unified approach that connects these efforts. This 371 approach addresses many challenges by advancing the FAIR (Findable, Accessible, Interoperable, 372 Reusable) principles, allowing researchers to more reliably find simulation modules, understand 373 those models, and connect them reliably into hybrid, multiscale models. We initiated the project 374 with a Verification API, that brings together COPASI, Tellurium, and AMICI-each of them fit 375 with a standardized process interface for uniform timescourses-load them with the same SBML 376 model and simulation instructions, runs them in parallel, and compares results. 377

2.9 openTECR: community curation of Thermodynamics of Enzyme-Catalyzed Reactions 379

Robert Giessmann, Institute for Globally Distributed Open Research and Education (IGDORE), Berlin, Germany

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openTECR (""Open database on Thermodynamics of Enzyme-Catalyzed Reactions"") is a database and a community.

We create a data collection of apparent equilibrium constants of enzyme-catalyzed reactions, being reliable, open and machine-actionable, with a clear change process to integrate new data and correct errors. We believe that Open Science principles, and specifically Open Data and Open Source are key to achieving our vision.

The openTECR database serves computational and experimental scientists in the fields of metabolic engineering, genome-scale metabolic modelling, biocatalysis and related fields by providing curated information. It is used by eQuilibrator as the data basis for making predictions about any possible reaction.

Recently, we organized an open community curation effort (https://opentecr.github.io/ invitation-to-curate). We prepared a curation workflow to analyze 278 pages of pages densely packed with tables and textual information. We invited volunteer contributions, and are immensely grateful about 17 volunteers investing almost 100 working hours.

At the COMBINE 2024 meeting, I would like to present our initiative and share our lessons learned about organizing successful community curation. I believe that our example can serve as a blueprint for other databases / project ideas which require a large amount of working hours. 399

We discovered that key to receiving contributions is to offer very small packages of work and a detailed curation manual. Our smallest task was 3 minutes long and well received.

Our small community (40 members) shares a mailing list (https://w3id.org/opentecr) and a GitHub organization where we store our data and code under open licenses (https://github. com/opentecr/).

2.10 OpenVT – Developing Framework Description Standards for MultiCellular Agent-Based Virtual Tissue Models 406

James Glazier, Indiana University, USA

Many simulation frameworks implement multicellular agent-based models using a variety of 409 methodologies (center model, vertex model, Cellular Potts model, finite-element mechanics, ...) 410 and support a variety of biological and mathematical processes it can be often confusing and 411 time consuming to for a researcher to know which simulation framework can fulfill their modeling 412 needs. In our breakout session, we will discuss an approach to defining and categorizing simulation 413 framework capabilities. The session will start with an overview of the various methods employed in 414 multicellular simulations, highlighting their unique features and common challenges. It will present 415 our approach to describing framework descriptions in a standardized way followed by discussion 416 on this approach. 417

2.11 OpenVT: MultiCellular Agent-Based Virtual Tissue Models: Defining Topics and Priorities for Working Groups and Virtual Workshops

James Glazier, Indiana University, USA

Virtual Tissues (VT), agent based multicellular modeling has become indispensable in under-422 standing complex biological phenomena, from tissue development to disease progression. But the 423 diversity in simulation methods poses challenges in reproducibility, modularity, reusability, and 424 integration for multiscale models, leading to a fragmented ecosystem and hindering growth. The 425 OpenVT Community is trying to address these challenges by bringing siloed research groups to-426 gether to improve the sharing of VT knowledge. The OpenVT Community supports the expansion 427 of and broader adoption of multicellular modeling beyond academic research labs into greater in-428 dustry practice. Development of best practices and better reproducibility will ultimately lead to 429 models that more closely follow FAIR (Findable, Accessible, Interoperable, and Reusable) princi-430 ples, leading to wider use in the approaches, toxicology, drug discovery and personalization 431 of testing and treatment. 432

2.12 OpenVT – Developing Reference Models for Multicellular Agent-Based Virtual Tissue Models 434

James Glazier, and James Osborne (University of Melbourne, Australia), Indiana 435 University, USA 436

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An increasing number of packages implement multicellular agent-based models using a variety 438 of methodologies (center model, vertex model, Cellular Potts model, finite-element mechanics, 439 ...). In principle a set of underlying biological and physical processes should yield the same result 440 independent of the package in which they are implemented. However, at the moment, comparison 441 between methodologies or even between different packages implementing the same methodology, 442 is quite challenging. As a first step to building a shared understanding of modeling capabilities 443 and to improve rigor and reproducibility, we define a minimal set of standard reference models 444 which should be implemented in each framework to illustrate their capabilities and reveal hidden 445 discrepancies of approach. This talk will discuss these efforts and introduce our planned breakout 446 session. 447

2.13 A functional tissue unit approach to understanding lung function in health and disease 449

Ruobing Li, Alys Clark, Merryn Tawhai, David Nickerson, Kelly Burrowes, Auckland Bioengineering Institute, University of Auckland, New Zealand

The primary functional tissue unit of the lungs is the acinus. An acinar unit brings together 453 diverse functions, including airflow, blood flow, gas exchange, mechanical deformation and the effect 454 of surfactant on this, and fluid transport from the blood to the lymphatic vessels. Surfactant is 455 important in reducing alveolar surface tension, ensuring stability and preventing alveolar collapse. 456 Ventilation is driven by the dynamic processes of lung tissue expansion and recoil during breathing. 457 Regional tissue recoil pressures also influence pulmonary perfusion, impacting the distribution of 458 blood flow within the lung. The lymphatic system, integral for maintaining fluid balance and 459 optimal immune function, is affected by these mechanical forces too. The interdependence of 460 these factors is vital for maintaining optimal pulmonary function under physiological conditions. 461 Existing models of varying geometric complexity have been developed to simulate lung mechanical 462 behaviours and various fluid transport, currently as separate systems. The ventilation model of 463 Swan et al. [J Theor Biol. 2012; 300:222-31] combines lung airway structure and tissue mechanics 464 with airflow dynamics. A perfusion model by Clark et al. (2010) simulates pulmonary blood flow 465 within the vasculature. A lung lymphatic model developed by Ashworth et al. (2023) estimates 466 the transfer of fluid from the capillary blood vessels into the interstitial space and the lymphatic 467 vessels. The surfactant model based on Otis et al. (1994) work simulates the dynamic adsorption-468 desorption process at the air-liquid interface and estimates the impact of surfactant on tissue 469 compliance. These different models reflect the diverse functions occurring within each acinus that 470 work together to determine emergent lung function. However, there is no comprehensive model 471 that integrates these aspects to form a complete functional tissue unit (FTU) of the lung. This 472 study addresses this gap by developing a respiratory FTU that integrates these different models 473 to simulate acinar function and link this to represent whole lung function. Model implemented by 474 CellML, Fortran, and Python. By integrating these individual models, we aim to provide a better 475 understanding of the interactions and dependencies within the lungs, essential for simulating lung 476 function in health and disease. 477

3 Breakouts

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3.1 Workshop: refineGEMs and SPECIMEN for automated model reconstruction and annotations

Gwendolyn O. Döbel, Martin Luther University Halle-Wittenberg, Germany

"Metabolic model reconstruction usually relies on several cumbersome steps. Different tools exist, which are only partially automated and need to be connected manually. Our aim is to simplify and reduce the manual workload. Thus, we developed the toolbox refineGEMs and the workflow collection SPECIMEN. A stable release of refineGEMs was already used in practice (Bäuerle et al., 2023). Both tools are currently under active development (enhancement and extension).

This workshop aims to give the attendees a brief introduction to automatic metabolic modelling with the tools refineGEMs (https://github.com/draeger-lab/refinegems) and SPECIMEN (https://github.com/draeger-lab/SPECIMEN). As part of the workshop, an open discussion will be held about issues arising from automatic energy-generating cycle (EGC) dissolution and gap filling.

Simon Hauser and Fritz Otlinghaus, Helsinki Systems, Germany

Introduction into Nix for scientific software

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Working on software in a team brings all kinds of challenges, especially because everyone has 496 a slightly different development environment. These challenges usually start with onboarding new 497 team members, include complications of moving your local environment to a high performance 498 cluster and end in unreproducible bugs that boil down to ""works on my machine"". Some of these 499 issues can be resolved by providing dependency pinning using poetry or other package managers, 500 but these solutions do not cover the operating system and require additional install documentation 501 that usually contains apt usage. Nix is a general purpose package manager that emerged in the last 502 couple of years that solves these issues, by not just pinning the version of dependencies but also 503 system libraries and tools, like the glibc library, python and also python packages. This session 504 will cover the fundamentals of Nix, including installation, command usage and writing your own 505 custom development environment for a specific software. Participants will learn how to leverage Nix 506 to create reproducible scientific workflows, manage dependencies, and ensure consistent software 507 environments across different systems. Through practical demonstrations and hands-on activities, 508 attendees will gain the skills necessary to integrate Nix into their scientific projects, enhancing 509 both the reliability and portability of their software. Join us to discover how Nix can streamline 510 your scientific software development and deployment processes, fostering greater collaboration and 511 innovation in your research endeavors. 512

Biological and Biophysics Simulation in Tissue Forge: Introduction and Guided 3.3513 Simulation Building

T.J. Sego, University of Florida, USA

Tissue Forge is open-source simulation software for interactive particle-based physics, chemistry 517 and biology modeling and simulation. Tissue Forge allows users to create, simulate and explore 518 models and virtual experiments based on soft condensed matter physics at multiple scales, from the 519 molecular to the multicellular, using a simple interface. While Tissue Forge is designed to simplify 520 solving problems in complex subcellular, cellular and tissue biophysics, it supports applications ranging from classic molecular dynamics to agent-based multicellular systems with dynamic pop-522 ulations. Tissue Forge users can build and interact with models and simulations in real-time and 523 change simulation details during execution, or execute simulations off-screen and/or remotely in 524 high-performance computing environments. Tissue Forge provides a growing library of built-in 525 model components along with support for user-specified models during the development and appli-526 cation of custom, agent-based models. Tissue Forge includes an extensive Python API for model 527 and simulation specification via Python scripts, an IPython console and a Jupyter Notebook, as 528 well as C and C++ APIs for integrated applications with other software tools. Tissue Forge sup-529 ports installations on Windows, Linux and MacOS systems and is available for local installation 530 via conda. This tutorial introduces the basic concepts, modeling and simulation features, and 531 some relevant modeling applications of Tissue Forge through guided simulation scripting. Tutorial 532 concepts will introduce basic Tissue Forge modeling concepts and simulation features through the 533 development of interactive simulations in Python. Attendees are encouraged, but not required, 534 to code along as the tutorial interactively develops and tests simulations in multicellular and bio-535 physics modeling applications. 536

A COMBINE Standard for Multi-Approach Multi-Scale (MAMS) Modelling 3.4

Sheriff Rahuman, EMBL-EBI, UK

Multi-approach Multi-scale (MAMS) modelling represents a cutting-edge method for modelling 540 and analysis of biological systems, leveraging an integrated suite of diverse modelling frameworks. 541 This multi-approach modelling will encompass a combination of diverse modelling formalisms, such 542 as ordinary differential equations (ODE), partial differential equations (PDE), logical, constraint-543 based, and agent-based models across multiple scales. These models are intricately tied together 544 to facilitate complex simulations. During the dedicated breakout sessions at Harmony 2021, COM-545 BINE 2021, and HARMONY 2024, we delved into the existing state-of-the-art technologies and 546 standards, including SBML and SED-ML, and their support for multi-approach modelling. These 547 discussions also illuminated the current challenges and gaps within the field. For COMBINE 2024, 548

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our objective is to further this conversation by identifying published MAMS models and bringing together the community to enable the creation of novel standards or the enhancement of existing COMBINE standards to support MAMS. This effort aims to foster interoperability and support the rapidly evolving paradigm of MAMS modelling.

3.5 Continuing Work Towards Reproducible Stochastic Biological Simulation

T.J. Sego, University of Florida, USA

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Stochastic simulations are commonly used to quantitatively or semi-quantitatively describe the 556 dynamics of biological systems. At various scales and in multiple applications, stochastic simu-557 lation better reflects observed biological processes and robustness. Various methods are widely 558 used to incorporate stochasticity into biological simulation, such as the Gillespie stochastic simu-550 lation algorithm for systems biology modeling, stochastic Boolean networks for network modeling, 560 and the Cellular Potts model methodology for multicellular modeling. Proving reproducibility of 561 simulation results is critical to establishing the credibility of a model. To this end, BioModels, 562 the largest repository of curated mathematical models, tests and reports the reproducibility of 563 simulation results for all submitted models when possible. A recent study showed that about 564 50% of the deterministic ordinary differential equation models on BioModels could not be repro-565 duced when applying criteria for reproducibility to the information provided in their associated 566 publication, reflecting a current crisis of reproducibility. Furthermore, there are no well-accepted 567 metrics or standards for reproducing stochastic simulation results, thus perpetuating the crisis of 568 reproducibility for a broad class of biological models. This breakout session will continue work to-569 wards establishing an accepted framework for testing the reproducibility of stochastic simulations 570 in biological modeling. The session will provide a brief overview of recent progress towards defin-571 ing quantitative measures to determine whether stochastic simulation results can be reproduced, 572 and when results have been reproduced. Attendees will discuss current issues to address towards 573 consensus and broad adoption in relevant modeling communities, as well as future work towards 574 reproducibility of stochastic simulation results using multiscale and complex models. 575

3.6 Morpheus: A user-friendly simulation framework for multi-cellular systems 576 biology 577

Lutz Brusch and Jörn Starruß, Technische Universität Dresden, Germany

Multi-cellular modeling and simulation become increasingly important to study tissue morpho-580 genesis and disease processes. This tutorial introduces Morpheus (https://morpheus.gitlab.io) 581 in an overview presentation with live demos and hands-on exercises runnable in sync on the pre-582 senter's and your own laptop. The focus lies on importing SBML models into Morpheus, extending 583 them in space as reaction-diffusion processes and automatically ""cloning"" them into many indi-584 vidual cells that can dynamically interact. Also, own modeling ideas can be explored with the help 585 of a tutor. Morpheus offers modeling and simulation of multi-cellular dynamics in a Graphical 586 User Interface (GUI) without the need to program code. It uses the domain-specific language 587 MorpheusML to define and simulate multicellular models in 3D space including the most common 588 cell behaviors and tissue mechanics. Morpheus is open-source software and provides readily instal-589 lable packages for macOS, Windows, Linux (https://morpheus.gitlab.io/download/latest/). 590 Please download before the tutorial and have a look around the homepage incl. ;90 example 591 models. 592

3.7 Combine spatial multi-cellular modelling with SBML

Jörn Starruß, TU Dresden, Germany

Modularity is key to creating complex multi-cellular models while preserving the accessibility of meaningful submodels. Naturally, composition also encourages reusability and the likes. We want to discuss and establish a common practice how to overlay the spatial dynamics of multi-cellular models with reaction dynamics defined in the SBML standard. 599

Most obvious features to be represented separately from the spatial cell dynamics are intracellular regulatory systems, inter-cellular communication and spatial reaction-diffusion processes using SBML-spatial (https://sbml.org/documents/specifications/level-3/version-1/spatiad/). Further issues arise when inter-connecting identical submodels residing in individual cells and the definition of instantaneous assignments upon entity operations (e.g. cell birth and death).

As an introductory motivation we will present our latest Morpheus (https://morpheus.gitlab. 605 io) results in embedding spatial reaction-diffusion submodels within moving cells. Using that experience we will sketch a way how to exploit the HMC package (https://sbml.org/documents/ specifications/level-3/version-1/comp/) to compose SBML models and attach them in a second step to the individual scopes of our spatial model. We hope for a lively discussion on best practice approaches interconnecting spatial multi-cellular modeling and the SBML standard. 610

3.8 Training Models using PEtab

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Fabian Fröhlich, The Francis Crick Institute, UK

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PEtab is a standardized file format for specifying parameter estimation problems (Schmiester et al., 2021). The interoperable format is currently supported by 11 different tools (https://github. com/PEtab-dev/petab#petab-support-in-systems-biology-tools), enabling users to benefit from standardized parameter estimation across frameworks based in Python, Julia, R, MATLAB, C++, or GUIs.

Although PEtab was initially developed for parameter estimation, recent efforts have extended the format to improve standardization of various adjacent tasks, including: model selection, multiscale modeling, PKPD and NLME modeling, optimal control, and visualization.

In this breakout session, based on audience interests, we will present introductions to PEtab and its extensions, then discuss current efforts to improve PEtab. People unfamiliar with PEtab are welcome to attend, and might first like to check out the tutorial (https://petab.readthedocs. io/en/latest/tutorial.html).

3.9 SBGN PD: current and future development

Adrien Rougny, Luxembourg Centre for Systems Biomedicine, University of Luxembourg, Luxembourg

Visualization of biological processes plays an essential role in life science research. Over time, 630 diverse forms of diagrammatic representations, akin to circuit diagrams, have evolved without 631 well-defined semantics potentially leading to ambiguous network interpretations and difficult pro-632 grammatic processing. The Systems Biology Graphical Notation (SBGN) standard aims to reduce 633 ambiguity in the visual representation of biomolecular networks. It provides specific sets of well-634 defined symbols for various types of biological concepts. SBGN comprises three complementary 635 languages: Process Description (PD), Entity Relationship (ER), and Activity Flow (AF). The 636 XML-based SBGN Markup Language (SBGN-ML) facilitates convenient storage and exchange of 637 SBGN maps. The SBGN languages as well as SBGN-ML are described in detail in specifications 638 (see sbgn.org). This breakout session will focus on the development of SBGN PD. We invite all 639 participants interested in SBGN to join this session, where we will discuss specific issues related 640 to the next version of the PD specification, as well as more open issues related to a future level of 641 SBGN PD. 642

3.10 Developing a proof of concept for a heap of git-versioned json files as a FAIR alternative to relational databases

Robert Giessmann, Institute for Globally Distributed Open Research and Education (IGDORE), Berlin, Germany 646

I propose, if there is interest of fellow participants, to think about, and just try out, a heap of json files, versioned in git, as an alternative form to store structured data.

Relational databases are great, but hard for "non-computational people" (read as: the typical experimental, wet-lab person) to create and to change. A heap of json files on GitHub seems still far fledged for some of those persons, but might be the minimal necessary technical barrier they have to cross.

Of course, data inside those json files must, for instance, be "quality controlled", i.e. checked for correctness and sticking to schemata. But that could be implemented as CI/CD actions. Git itself might put practical limits on the general feasibility of that idea – which is to be tested out.

If anyone else is up for it, let's try it out!

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3.11 OpenVT – Developing Framework Description Standards for MultiCellular Agent-Based Virtual Tissue Models

James Glazier, Indiana University, USA

Many simulation frameworks implement multicellular agent-based models using a variety of 662 methodologies (center model, vertex model, Cellular Potts model, finite-element mechanics, ...) 663 and support a variety of biological and mathematical processes it can be often confusing and time 664 consuming to for a researcher to know which simulation framework can fulfill their modeling needs. 665 This session will discuss an approach to defining and categorizing simulation framework capabil-666 ities. The session will start with an overview of the various methods employed in multicellular 667 simulations, highlighting their unique features and common challenges. It will present our ap-668 proach to describing framework descriptions in a standardized way followed by discussion on this 669 approach. 670

3.12 OpenVT: MultiCellular Agent-Based Virtual Tissue Models: Defining Topics and Priorities for Working Groups and Virtual Workshops

James Glazier, Indiana University, USA

Virtual Tissues (VT), agent based multicellular modeling has become indispensable in under-675 standing complex biological phenomena, from tissue development to disease progression. But the 676 diversity in simulation methods poses challenges in reproducibility, modularity, reusability, and 677 integration for multiscale models, leading to a fragmented ecosystem and hindering growth. The 678 OpenVT Community is trying to address these challenges by bringing siloed research groups to-679 gether to improve the sharing of VT knowledge. The OpenVT Community supports the expansion 680 of and broader adoption of multicellular modeling beyond academic research labs into greater in-681 dustry practice. Development of best practices and better reproducibility will ultimately lead to 682 models that more closely follow FAIR (Findable, Accessible, Interoperable, and Reusable) princi-683 ples, leading to wider use in the approaches, toxicology, drug discovery and personalization 684 of testing and treatment. This session aims to discuss current progress undertaken by the OpenVT 685 community towards a shared ecosystem and look to attendees for insight into what they believe 686 will encourage broader adoption of community guidelines. 687

3.13 OpenVT – Developing Reference Models for Multicellular Agent-Based Virtual Tissue Models

James Glazier, Indiana University, USA

An increasing number of packages implement multicellular agent-based models using a variety 692 of methodologies (center model, vertex model, Cellular Potts model, finite-element mechanics, ...). 693 In principle a set of underlying biological and physical processes should yield the same result in-694 dependent of the package in which they are implemented. However, at the moment, comparison 695 between methodologies or even between different packages implementing the same methodology, 696 is quite challenging. As a first step to building a shared understanding of modeling capabilities 697 and to improve rigor and reproducibility, we define a minimal set of standard reference models 698 which should be implemented in each framework to illustrate their capabilities and reveal hid-699 den discrepancies of approach. This session will discuss these efforts and look for feedback from 700 attendees. 701

4 Poster

4.1	SPECIMEN: Collection of Workflows for Automated and Standardised Recon-	703
	struction of Genome-Scale Metabolic Models	704

Carolin Brune, Gwendolyn O. Döbel, Famke Bäuerle (QBiC), Natia Leonidou (IBMI, DZIF, QBiC), Reihaneh Mostolizadeh (Justus Liebig University Gießen), and Andreas Dräger, Martin Luther University Halle-Wittenberg, Germany

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SPECIMEN (https://github.com/draeger-lab/SPECIMEN) is an open-source collection of 709 different workflows designed for the automated and standardised curation of genome-scale models. 710 It is Python-based and integrates a variety of tools, including MCC, SBOannotator, refineGEMs, 711 and BOFdat, and more, to concatenate modelling steps like gap filling, annotation, duplicate 712 removal, and biomass normalisation into a single pipeline. SPECIMEN offers different workflows 713 tailored to various modelling approaches and types of input data, facilitating efficient and consistent 714 genome-scale model reconstruction. 715

4.2 What is an appropriate standard for modeling microbial communities?

Beatrice Ruth, Peter Dittrich (FSU Jena), and Bashar Ibrahim, Friedrich Schiller Universität Jena, Germany

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Microbial communities exist almost everywhere on earth and play a major role in environmen-720 tal processes as well as health and aging. A microbial community consists of a complex network 721 of taxa and their interactions with one another. My task at the moment is to find a way to 722 use measurements to understand the hierarchical structure and to predict interactions within the 723 community. In this context, I noticed that there are no comprehensive standards for modeling 724 microbial communities. Therefore, I would like to ask the question, what would be a suitable 725 standard for modeling microbial communities? To represent the hierarchical structure, the inter-726 sections and unions of self-maintaining taxa combinations (organizations) are ordered in a lattice. 727 While the intersections highlight the core microbiome and the general similarities between different 728 organizations, additional unions reveal possible negative interactions. When working with mea-729 surements, in addition to the individual taxa composition, other environmental parameters such 730 as at least time and location are important. Based on chemical organization theory, the individual 731 organizations of a given reaction network and thus also their intersection and union sets can be 732 calculated. The reverse path is now examined for interaction prediction. Here I am currently using 733 resource-consumer models including toxins to also take negative interactions into account. Many 734 different models can create the same hierarchical structure. Therefore, the focus is on a model 735 for representation with as few resources as possible, assuming as few interactions as possible. I'm 736 curious how this could be accounted for in a standard, as it would increase clarity and simplify 737 exchange. Does it need more than, for example, SBML or a simple extension? 738

4.3 Integrate modelling standards with Energy-based System Analysis

Weiwei Ai, Peter Hunter, and David Nickerson, Auckland Bioengineering Institute, The University Of Auckland, New Zealand

Energy is a fundamental concept in physical processes. Physiological systems often involve various physical processes, with energy serving as a universal language across different domains. Energy-based modelling frameworks, such as bond graphs and port-Hamiltonian formulation, have been recently introduced into the computational biology community. The energy-based frameworks adopt a hierarchical and modular approach, which captures traceable energy storage, dissipation, and transduction, offering comprehensive insights into the systems under investigation. 746

The COMBINE community has established standards for computational biological models, such as CellML and SED-ML, which are used for encoding mathematical models and simulation experiments of physiological processes. This talk will explore the potential integration between these standards and energy-based approaches. We will demonstrate how we leverage these modelling standards to extract information from models for system analysis and invite input and suggestions from the community.

4.4 The preCICE v3 coupling library and the emerging preCICE ecosystem

Gerasimos Chourdakis, Jun Chen; Ishaan Desai, Carme Homs-Pons, Benjamin Rodenberg (Technical University of Munich), David Schneider, Miriam Schulte, Frédéric Simonis, and Benjamin Uekermann, University of Stuttgart, Germany

The coupling library preCICE (precice.org) allows coupling simulation codes at runtime, enabling flexible and efficient partitioned multi-physics simulations, exchanging data over point locations during a time loop. preCICE v1, released in 2016, introduced a high-level API, offering massively parallel communication and mapping methods, as well as advanced IQN coupling al-763 gorithms. preCICE v2, released in 2020, included several integrations to established open-source 764 codes (such as OpenFOAM, SU2, deal.II, FEniCS, and more), allowing users to execute simulations 765 often without having to write any code. This increasing number of components and examples is 766 now a citable preCICE Distribution. The latest preCICE Distribution (v2404) includes 23 com-767 ponents and 31 different example scenarios coupling multiple possible combinations of codes, fea-768 turing preCICE v3, which introduces several new features (such as time interpolation, much faster 769 Partition-of-Unity RBF mapping methods, experimental Geometric Multiscale mapping), and a 770 much simplified API. Common applications now extend far beyond fluid-structure interaction, now 771 including applications in biomechanics (via codes such as OpenDiHu or FEBio), porous media (e.g., 772 via DuMux), ice-sheet modelling, and more. This poster will summarize important updates in the 773 preCICE project, and discuss current plans for integrating the community towards a community-774 driven ecosystem of FAIR (Findable, Accessible, Interoperable, and Reusable) components and 775 simulation cases, opening up to new applications and scientific communities. 776

4.5 ModelPolisher: Enhancing the Quality and Completeness of Genome-Scale Metabolic Models (GEMs) 778

Dario Eltzner, Bahaa Ziadah, Thomas J. Zajac, Matthias König, Kaustubh Trivedi and Andreas Dräger, Computational Systems Biology of Infections and Antimicrobial-Resistant Pathogens, Institute for Biomedical Informatics (IBMI), Tübingen, Germany 782

Background:

Genome-scale metabolic models (GEMs) play a central role in systems biology and enable 785 detailed simulations and predictions of metabolic processes. However, the quality and completeness 786 of these models can vary considerably, which compromises their utility. GEMs often suffer from 787 inconsistencies, incomplete annotations, and structural inaccuracies that can limit their utility. A 788 measure of model quality, the MEMOTE score, often highlights these shortcomings, indicating 789 areas such as missing gene associations, metabolite inconsistencies, and incorrect mass balances. 790 To address this problem we developed ModelPolisher, a tool for standardizing, annotating and 791 refining SBML models. 792

The results:

ModelPolisher has drastically improved the quality of the annotation of GEMs. By aligning model components with BiGG IDs, the tool enriches models with consistent and detailed metadata, facilitating model sharing and reproducibility. In addition, ModelPolisher's built-in checks for structural correctness, such as mass balance and metabolite connectivity, have proven effective in identifying and correcting errors. Our application of ModelPolisher to a number of models from the BiGG and BioModels databases has resulted in a major improvement in model quality and metadata completeness.

Conclusion:

ModelPolisher is a must-have tool for systems biology that addresses the critical need for highquality, well-annotated GEMs. Its power to automatically enhance model metadata and ensure structural integrity not only improves model utility, but also promotes collaboration and data sharing. The tool's impact is particularly evident in large modeling projects where consistency and accuracy are paramount.

Availability:

ModelPolisher is open-source on GitHub at https://github.com/draeger-lab/ModelPolisher. It can be used from the command line or integrated into larger workflows and offers a flexible solution for researchers. Extensive documentation and examples make it easy to use, and the community is encouraged to contribute to the ongoing development and improvement of the tool.

4.6 Partitioned simulations using the neuromuscular simulation framework OpenDiHu₁₂

Carme Homs-Pons, and Miriam Schulte, University of Stuttgart, Germany

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"OpenDiHu is a high-performance computing framework for skeletal muscle simulations. Created and developed at the University of Stuttgart, it is an open-source project written in C++. It uses python scripting and provides CellML support. OpenDiHu offers ready-to-use physics-specific solvers that can be combined by the user to create a tailored muscle solver. The available models include 3D finite element hyperelastic models, subcellular models and a motor neuron pool model among others. The user does not have to worry about the mapping between models or time 820 sub-cycling, as this is automatically done by the internal coupling tool from OpenDiHu. Besides, 821 OpenDiHu has a preCICE adapter. preCICE is a coupling library for partitioned multi-physics 822 simulations. Using preCICE, we can couple OpenDiHu to other software, e.g., FEBio and deal.ii. 823 Finally, the poster will include use-cases to better showcase what we can do using OpenDiHu and 824 preCICE. In particular, we will present our latest results for a human biceps simulation and our 825 work-in-progress towards a model of the agonist-antagonist myoneural interface." 826

4.7 SBSCL: A Library of Efficient Java Solvers and Numerical Methods to Analyze 827 Computational Models in Systems Biology 828

Arthur Neumann (Eberhard Karl University of Tübingen, Germany), Max Hatfield 829 (German Center for Infection Research (DZIF), Quantitative Biology Center (QBiC), 830 Eberhard Karl University of Tübingen), Taichi Araki (Graduate School of Science 831 and Technology, Keio University, Japan), Akira Funahashi (Graduate School of Sci-832 ence and Technology, Keio University, Japan), and Andreas Dräger (Data Analytics 833 and Bioinformatics, Institute of Computer Science Martin Luther University Halle-834 Wittenberg, German Center for Infection Research (DZIF), Quantitative Biology 835 Center (QBiC), Eberhard Karl University of Tübingen, Germany) 836

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Numerical calculations are at the heart of systems biology. Such computations require inter-838 preting models in a specific framework and performing several preprocessing steps to pass the 839 model to a specialized solver. The Systems Biology Simulation Core Library (SBSCL) is a soft-840 ware library that simulates and analyzes diverse systems biology models, including flux balance 841 constraints, stochastic simulation, and ordinary differential equation systems. It parses SBML 842 models (the Systems Biology Markup Language), a common language used to describe biological 843 processes, and SED-ML files to conduct more involved analyses. The library supports various 844 algorithms for deterministic and stochastic simulations, allowing precise and efficient simulation 845 of even complex biological and biochemical processes. Furthermore, the library supports integra-846 tion with other tools and frameworks. SBSCL has been well-tested and benchmarked against the 847 entire SBML Test Suite. It supports several extension packages and provides a highly efficient 848 solver package for SBML models that can be incorporated into any program that runs on the 849 Java Virtual Machine (JVM). SBSCL is available free of charge, even for commercial purposes, at 850 https://github.com/draeger-lab/SBSCL. 851

BayModTS: A Bayesian workflow to process variable and sparse time series 4.8 852 data. 853

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Biomedical data generation is limited due to cost and ethical aspects. This leads to sparse time series with only a few replicates available. The analysis of this data is further complicated by the inter-individual variability of organisms and the variability within one organism over time. In this context, analyses that consider only the means and ignore the data variability fail due to low signal-to-noise ratios.

Bayesian Modeling of Time-Series Data (BayModTS) processes the data and takes the un-861 certainty from highly variable time-series data into account. It employs the retarded transient 862 functions of C. Kreutz as a universal simulation model and can be easily adapted to user-specified 863 SBML models. Using an appropriate noise model, a parameter posterior distribution is inferred 864 via Markow-Chain-Monte-Carlo sampling. Posterior predictive distributions transfer parameter 865 samples from the posterior to model predictions, providing continuous predictions with filtered 866 noise. We demonstrate BayModTS' feasibility on rats' in vivo liver perfusion after 60% Portal 867 Vein Ligation. BayModTS acts as a noise filter and transforms MRI perfusion measurements into 868 time-continuous predictions about the perfusion of individual liver lobes equipped with credibility 869 tubes. These can be used as input for liver function models. 870

In summary, BayModTS is a Findable, Accessible, Interoperable, and Reusable (FAIR) Bayesian 871 workflow to analyse variable and sparse time series data. A user-friendly toolbox can be found on 872 GitHub. 873

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The European EDITH (Ecosystem Digital Twins in Healthcare) project (https://www.edith-879 csa.eu) funded by the European Commission is paving the road for building a European infras-880 tructure for Virtual Human Twins (VHTs) in healthcare. A Virtual Human Twin (VHT) is data-881 and/or knowledge-driven multiorgan and multiscale representation of the quantitative human phys-882 iology of a single individual or a group of individuals and can be used for complex personalized 883 predictive computational simulations that are applicable for personal health forecasting, for disease 884 and treatment prognosis prediction, for personalized clinical decision support systems to simulate 885 medical treatment options, for the development of personalized medical products, and for the use 886 in biomedical research (e.g. for the data-driven generation of hypotheses in the development of 887 mechanistic models), as well as for many other possible applications in the health domain. Building 888 such an infrastructure for Virtual Human Twins requires interoperability of the manifold data and 889 computational models constructed based on those data, and thus, a high degree of standardization 890 of data and models, as well as applied workflows, modelling approaches and provenance informa-891 tion for traceability. Such standards are defined by initiatives of the scientific community, such 892 as COMBINE, GA4GH, ASME and others, as well as by formal Standard Defining Organizations 893 (SDOs), such as ISO with their technical committees. EDITH develops a proof of concept for a 894 data and model repository and a simulation platform and comprises also ethical, legal, social im-895 plications (ELSI) and regulatory compliance aspects, so that in the long run, EDITH will establish 896 a marketplace for digital twins in healthcare. 897

4.10 Recommendations and requirements for implementing computational models ⁸⁹⁸ in clinical integrated decision support systems ⁸⁹⁹

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Years of progress in biomedical technology have generated a vast number of omics, medical 911 imaging, and health data in multiple formats and described by corresponding metadata in het-912 erogenous ways. Despite its significant promise for clinical use, this big data remains underutilized. 913 The EU-funded EU-STANDS4PM project has established a pan-European expert forum for eval-914 uating existing standards and develop new guidelines for in silico methodologies in personalized 915 medicine. In this context an ad-hoc working group has been created to discuss the practical recom-916 mendations and requirements that should be considered for implementing computational models 917 in clinical integrated decision support systems. The outcome of these discussions has resulted in 918 the standard draft ISO/TS 9491-2 "Guidelines for implementing computational models in clinical 919 integrated decision support systems" submitted to and accepted by the ISO committee ISO/TC 920 276 Biotechnology. Its publication by ISO is anticipated. 921

This standard draft delivers fundamental requirements for: 1) clinically-driven projects standardization, 2) data handling, 3) assessment of data availability and quality in clinically-driven projects, 4) data modeling and interpretability, 5) validation of existing and development of new models for different populations, 6) uncovering patient-specific and population-related patterns that can improve care, 7) reinforcing a multidisciplinary decision-making process, 8) creating a virtuous cycle of learning, 9) patient involvement and 10) risk management.

We here introduce a guideline for setting up, detailing, annotating, as well as ensuring the interoperability and integration of health data and resulting models, along with their accessibility and origin, in a way that is both understandable and grounded in evidence. It outlines the integration of these guidelines with the conduct of clinical trials through standard operating pro-931 cedures. Additionally, it deals with the criteria and advice for the data needed to build or validate these models. These recommendations aim to contribute to the standardization of a framework to regulate the use of data-driven systems for clinical research.

4.11 Shining Light on Single-Cell Dynamics and Heterogeneity: Design and analysis of a hybrid population model for an epigenetic memory system.

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Heterogeneity in biological systems can be quantified efficiently by single-cells measurement 939 techniques like flow cytometry. However, many modelling approaches currently cannot capture 940 this behavior, as often only the average cell is covered in commonly used ODE models. Single-cell 941 data can be reduced to summary statistics for use with these models, but this leads to a loss of 942 information in the data and, more importantly, only provides a complete description if the data 943 is close to a normal distribution. Especially in the case of bimodal distributions, which can for 944 example occur in bistable systems, averages are poor descriptions of the data and lack the ability 945 to reproduce important features of the system. The synthetic epigenetic memory system from Graf 946 et al. (2022) is such a particular system. It is characterized by the ability to switch from an OFF-947 to an ON-state through a transient metabolic trigger. This ON-state is sustained via positive 948 feedback based on DNA methylation. A large part of the cells can remember this state for many 949 days, but eventually, more and more cells switch back to the OFF-state. In the experimental data, 950 this is observable as a transient appearance of two subpopulations, ON- and OFF-cells, with a drift 951 towards the OFF state. We aim to capture this transient bimodality by a tailored model which 952 describes heterogeneous single-cell trajectories. Our hybrid model combines the simulation speed 953 of differential equations with a stochastic process describing cell division, as well as distributed 954 parameters and measurement noise. We trained the model by comparing the simulated population 955 to the data using the Kolmogorov metric, a shape sensitive distance between distributions. The 956 model reproduces the experimental single-cell data as well as bulk methylation measurements well 957 and is able to predict previously unseen data, including experiments of cyclic ON-OFF-switching 958 with an additional input. Our trained model provides insights into the switching behavior and 959 in particular the mechanisms behind the drift towards the OFF-state on the population and on 960 the single-cell scale. Our analysis suggests that the stochastic nature of the cell division plays an 961 important role in the destabilization of the ON-state, but its effect is only observable over long 962 time. 963

4.12 TFpredict

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Gene-regulatory and signaling networks have matured into substantial instruments for analyz-974 ing biological functions. Constructing such networks highly depends on the availability of detailed 975 information about protein functions, which often needs to be completed, particularly for less well-976 studied organisms. Transcription factors (TFs) play a crucial role in regulating gene expression 977 and are vital for understanding complex biological networks. TFpredict is a cutting-edge super-978 vised machine-learning tool designed to facilitate network building by accurately predicting TF 979 interactions and regulatory pathways. In combination with SABINE, TFpredict can even predict 980 the nature of interactions and binding domains, providing a comprehensive view of the regulatory 981 mechanisms at play. The methodology involves BLAST score extraction, superclass prediction with 982 different classifier models, and the identification of DNA-binding domains with InterProScan. The 983 application of TFpredict in constructing robust and comprehensive biological networks showcases 984 its potential to revolutionize regulatory network analysis. By automating the prediction process, 985 TF predict significantly reduces the time and effort required for network construction. Its ability 986 to predict interactions and binding domains offers a detailed understanding of TF dynamics, fa-987 cilitating the study of complex regulatory pathways. This capability is particularly beneficial for 988

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research on less well-studied organisms, where experimental data may be sparse. TFpredict is a 989 powerful tool in the field of systems biology, enabling researchers to gain deeper insights into TF 990 dynamics and regulatory networks. Its integration into network analysis workflows enhances the 991 accuracy and comprehensiveness of the resulting models, paving the way for discoveries in gene reg-992 ulation. TF predict is available as an open-source Java project on GitHub, providing the scientific 993 community access to its functionalities. The tool can be easily integrated into existing bioinfor-994 matics pipelines, and comprehensive documentation facilitates its use. Researchers are encouraged 995 to contribute to its ongoing development and application, ensuring its continued evolution and 996 relevance in network modeling. 997

4.13 A Computational Pipeline for Evaluating Agreement Between Large-Scale Models and Diverse Datasets 999

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Computational models capturing the function of every gene within a cell, known as Whole-Cell 1003 Models (WCMs), can predict complex, multi-gene phenotypes while reconciling discrepancies in 1004 current understanding (Karr et al., 2012; Macklin et al., 2020). Constructing such models requires 1005 the integration of diverse datasets of varying sizes from different labs and assay types. However, 1006 aggregating these datasets into a model-readable format to scalably identify model-data mismatch 1007 (i.e. knowledge gaps) pose a large challenge for model construction (Szigeti et al., 2018). We are 1008 creating a computational pipeline to rapidly evaluate agreement of a large-scale mechanistic model 1009 of a human epithelial cell (the SPARCED model (Erdem et al., 2022)) with a compendium of data 1010 spanning multiple sources and modalities. Conditions, duration, and results of wet-lab experi-1011 ments are converted into a machine readable format based in-part on PEtab guidelines (Schmi-1012 ester et al., 2021). To ensure this pipeline covers a broad range of potential use case scenarios, we 1013 constructed 13 benchmarks SPARCED has previously been validated against, comprising various 1014 biological conditions, perturbations, and measurement techniques. Initial deployment (i.e. creat-1015 ing new benchmarks) on the LINCS Microenvironment (ME) perturbation dataset (Gross et al., 1016 2022) indicates mixed agreement with Reverse Phase Protein Array (RPPA) data. Further model 1017 agreement is being evaluated with RNAseq, ATACseq, and highly multiplexed immunofluorescence 1018 perturbation data. This pipeline will provide a means to rapidly evaluate how diverse datasets col-1019 lectively compare to model variants, thereby improving the accuracy and scalability of SPARCED 1020 and contributing to the creation of a human Whole-Cell Model. 1021

4.14 Standard compliant data and model management for systems medicine projects 1022

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Large collaborative projects need to share data during and after, within and beyond the con-1027 sortium. FAIRDOM-SEEK (https://fairdomseek.org/) is an open-source software for storing, 1028 cataloguing, sharing and reusing research outcomes designed to support the principles of FAIR 1029 (Findable, Accessible, Interoperable, and Reusable) research data management. Originally devel-1030 oped for the needs of systems biology of microorganisms, SEEK is used in numerous projects of 1031 systems biology, systems medicine, and related domains. All data types can be handled and the use 1032 of files or references to files is possible. Users can change the visibility of files and references, making 1033 it a platform for projects and data publication. Its properties make it an interoperability resource 1034 for combining different tools for scientific work and subsequent publication of the outcomes. The 1035 systems medicine approach to quantification and characterization of large complex systems involves 1036 integration of multipledata types (e.g. genomics, proteomics, metabolomics, phenomics, images, 1037 patient related data, etc.), stored in several specialized systems used within one project. LiSvM-1038 Cancer for example, uses REDCap (https://www.project-redcap.org/) as a clinical data system 1039 that manages information about patients and samples; openBIS (https://openbis.ch/) as pri-1040 mary system for experimental raw data and its metadata; Nextcloud (ttps://nextcloud.com/) 1041 for short-term raw data exchange; and OMERO for microscopic images. The harmonisation and 1042 integration of (meta)data between these platforms is mandatory to make the data comparable 1043 and publishable in open data repositories. Here, we describe our experience in combining multiple 1044 open-source data repository systems for the benefit of large collaborative system medicine projects. 1045

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Molecular mechanisms of biological systems and pathways may be described and represented 1050 graphically under the form of molecular maps. In a molecular map, nodes represent biological 1051 entities such as (pools of) bio-molecules and edges relationships between these entities. Molecular 1052 maps may be represented and exchanged using standard graphical languages and formats (e.g., 1053 SBGN, SBGN-ML, SBML) that are supported by a handful of editors (e.g., SBGN-ED, Newt, 1054 CellDesigner) and libraries (e.g., libSBGN, libsSBML, SBMLDiagrams). While these tools allow 1055 users to easily build and save maps as images or in standard exchange formats, they are not 1056 well suited to work with the content of maps programmatically. Here we introduce MomaPy, a 1057 Python library that allows users to perform a wide variety of tasks on maps, including reading, 1058 comparing and rendering them efficiently. At its core, MomaPy separates the model of a map 1059 (what is represented) from its layout (how it is represented) à la SBML+layout/render, easing the 1060 navigation of their biological content. MomaPy currently supports SBGN PD, SBGN AF, and 1061 CellDesigner maps, and may be easily extended to support other types of maps and additional 1062 tasks to be performed. 1063

4.16 Eulerian Parameter Inference: Modelling of Single-Cell Data

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Single-cell measurement techniques and spatial omics call for modelling approaches capable 1067 of capturing inherent cell population heterogeneity. In-silico models must be adequately param-1068 eterised to reflect the data and enable accurate predictions beyond data reproduction. We here 1069 present Eulerian Parameter Inference (EPI), a probabilistic inference method based on a change 1070 of variables. The input of EPI is 1) a deterministic simulation model, such as reaction rate equa-1071 tions or other ordinary differential equations, and 2) data exhibiting large variations, for instance, 1072 single-cell gene-expression data. EPI translates all information from the data into distributed 1073 model parameters. Further, the employed change of variables formulation allows for point-wise 1074 evaluation of the inferred parameter density. Each evaluation only requires one forward simulation 1075 of the model and its Jacobian. In particular, we do not require an explicit formulation of the 1076 inverse mapping from the data to the parameters. We demonstrate EPI's capabilities on diverse 1077 models ranging from algebraic equations to ordinary and even partial differential equation systems, 1078 thereby proving its practical applicability. The eulerpi Python package is available on the Python 1079 Package Index PyPI. It provides all necessary functionalities and only requires a model and a 1080 data sample as user input. We hope this easy-to-use package will facilitate EPI's applicability in 1081 numerous and diverse research groups. 1082

Figure 1: Participants at this year's COMBINE.

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