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To cite this article: M. Bühler and T. Bednar 2021 *J. Phys.: Conf. Ser.* **2069** 012141

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A review on coupled building physics analyses

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Abstract. This paper reviews methods and tools for coupled building physics analyses in the context of Building Performance Simulations (BPS) with a focus on Building Energy Simulations (BES) and Computational Fluid Dynamics (CFD) as a common application. Furthermore, requirements regarding the necessary information for simulations, data models and coupling are identified. Possibilities of automated simulation model generation, data exchange and the performance of existing multi physics simulation models are analysed and limiting factors are discussed.

1. Introduction

Building Performance Simulations (BPS) can be used to optimize buildings in different areas such as comfort, energy consumption or hygric behavior [1, 2]. BPS involve various analyses like Building Energy Simulations (BES) or Computational Fluid Dynamics (CFD) and cover a variety of disciplines. These reach from heat and mass transfer to thermodynamics, fluid mechanics, control and regulation, environmental science or occupant behavior to analyze whole buildings or neighborhoods.

Although a variety of calculation methods and tools exist, most tools address a specific problem, simplifying the calculation or boundary conditions to reduce complexity or calculation expense [2, 3]. For example, most Building Energy Simulations (BES) like EnergyPlus [4], TRNSYS [5], ESP-r [6] or IDA-ICE [7] assume simplifications like a homogeneous distribution of temperature and pressure in each zone or one-dimensional heat flow through surfaces, which considerably lowers the complexity of the calculation. However, this simplification leads to reduced accuracy and lacks local flow variables needed for comfort or hygric component behavior evaluation. To avoid these problems, CFD can be used to generate very accurate results here [8, 9, 1, 10, 11]. Other applications of CFD include the calculation of convective heat transfer coefficients (CHTC), thermal comfort, wind loads or the investigation of natural or cross ventilation [12, 11, 13, 14, 15].

However, an isolated consideration of a problem is often not adequate to generate sufficiently precise results or to investigate multiple interdependent problems. Therefore calculation methods or tools must be coupled in an interdisciplinary manner. Hence, coupling BES with CFD offers significant added value to the areas of application as mentioned above. [16, 8, 1].

Nevertheless, by combining multiple analyses, new challenges like incompatible or proprietary data models, the high effort required to create the complex computational models, high computing expenses or data exchange between different tools have to be faced. Furthermore, quick and easy access to detailed analyses is a basic requirement for a broad and economic use of



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BPS [17]. In order to apply optimisation algorithms or to develop predictive operating strategies by means of machine learning, automated model generation and high-performance calculation is necessary [18, 19, 20, 21, 22].

2. Coupling Methods

Due to the of performance and capabilities of BES, the system boundaries of CFD are usually defined as the surfaces of the air volume and inlets or outlets of vents, fan coils or windows [23]. The BES handles the conjugate heat transfer with a suitable radiation model, HVAC system models and plant models [24].

2.1. Data exchange

Tian et al. [1] divide the data exchanged between BES and CFD into interface data and state data. Interface data is defined as data at the boundary of the physical domains while state data is defined as data of the respective domain and is mainly used for control purposes of the HVAC. Zhai et al. [23] investigated different combinations of Dirichlet-, Neumann- or Robert-BCs as interface data for the thermal coupling of fluid and solid domains. Surface temperatures (Dirichlet-BC) as BC for the CFD and CHTC, as well as temperature gradients of the air (Robin-BC) as BC for the BES showed to provide the best performance and the most numerically stable coupling and are used in most studies [25].

The coupling of state data is more difficult because HVAC and plant models often differ a lot in the complexity of their structure. Therefore valid descriptions are hard to make [25, 26]. A common approach nowadays is the use of the Functional Mock-up Interface (FMI) [27], a standardised interface for model exchange and co-simulation of dynamic systems [28, 29, 30]. Complex systems can thus be built from modular Functional Mock-up Units (FMU), pre-compiled models with a standardised description of the exchanged data, parameters and values for the initialisation of variables. While there is a large library of FMUs [31, 32] and many BES are compatible with the FMI standard, there is little compatibility on the CFD side [33, 16, 1, 34], especially when it comes to commercial software.

2.2. Synchronization schemes

Due to the fact that interface and state data vary with time, it is theoretically necessary to run each BES and CFD simulation for each time step. Even during each time step, iteration between BES and CFD may be needed to reach a convergence [35]. Tian et al. [24] summarized coupling methods with a focus on synchronising data, which is mainly synchronised in three ways: Static coupling, dynamic coupling and bin coupling. The amount of time for these different strategies, as well as their simulations speed and accuracy vary broadly. To determine which coupling strategy is used, building characteristics, the required results and accuracy have to be taken into account. [25, 35].

3. Tools

Rodríguez et al. [25] and Harish et al. [10] give an overview of current studies on BES-CFD co-simulations and the used tools. Most commonly used BES tools are TRANSYS and EnergyPlus, while ANSYS Fluent [36] is predominantly used for CFD. For the coupling of the tools in the mentioned studies, a customer-specific solution is used in most case, in which the CFD simulation is executed by the BES by means of scripts and the results are read in by the BES via files. For the more advanced coupling via FMI, Zuo et al. [34] developed the FMU "Rooms.CFD", available in the Modelica Buildings library, which was used in a study to successfully couple the Modelica Buildings library and the fast fluid dynamics (FFD) program by Jin et al. [37]. Qiao et al. [16] also used this FMU to couple openFOAM and Modelica with a CFD-side adaptation.

4. Performance

As CFD simulations have a high computational effort, fully dynamically coupled simulations for longer periods are not possible, yet [38, 39]. Consequently, the development of methods to reduce the computational effort of CFD while maintaining accuracy has become a major area of research [39]. Hosain et al. [40], Feng et al. [41] and Morozova et al. [39] reviewed such methods, with Fast Fluid Dynamics (FFD) and Reduced Order Models (ROM) in particular appearing promising and being increasingly used in current work [19, 42, 43, 44, 45, 46]. Machine learning is also a heavily researched field and has already been successfully used to accelerate CFD [47, 48, 49]. With these methods, calculation in real time or even faster is possible [49, 39]. In addition to the use of new computational methods, the appropriate choice of the level of detail (LOD) of geometry is also a decisive factor. On the one hand relevant details should be taken into account [50, 51, 52, 53], but on the other hand complex geometries need a finer discretization and make high-quality meshing more difficult [54]. Furthermore, the computational time can also be significantly reduced by use of GPUs [55, 39], high performance computing [56] or distributed computing [57, 1].

5. Automated model generation

The automated generation of simulation models for CFD-BES co-simulations is a complex process that can be divided into the creation of a model for the BES with the building, HVAC system and equipment and the CFD model. The current standard and research focus is the use of Building Information Models (BIM) with the open exchange formats IFC and gbXML. Wetter et al. [27] describe workflows and tools for generating BIM models for BES from BIM (BIM to BES) in Annex 60, which have been implemented and further developed in numerous studies [58, 59, 60, 61, 62].

The CFD model requires a different kind of geometry than the one of the BES. The geometry for the CFD model has to be watertight and in most cases simplified to be able to generate a mesh with a high quality [50, 51, 53]. The meshing can then be performed considering the mesh quality [54, 63, 64] and boundary conditions can be assigned. Lee et al. [50] and Delavar et al. [65] describe a BIM-to-CFD workflow and its application but manual work will still be a part of that method.

6. Limitations

While the performance of BES coupled with CFD puts a limit to accuracy and the cases in which it can be applied, the high amount of work required to create the simulation models is also a major hurdle to widespread use. With BIM as the current standard, it is currently not possible to create stable and reliable simulation models. While Annex 60 lists some limitations, numerous studies also deal with problems in practical application. Common problems are missing data [66, 59, 67, 68, 10], data loss [69], geometry errors [69, 60, 70] or inconsistencies [66, 71]. Furthermore, there is still a high level of manual effort required to create the simulation models, which is error-prone and tedious due to the high level of complexity [10].

7. Conclusion

This paper has provided an overview of methods and tools for coupling BPS with a focus on BES and CFD. CFD is a discipline with a high computational cost as a limiting factor. Therefore, coupling has to be balanced between performance and accuracy through the choice of the synchronization scheme, solver method and LOD depending on the building characteristics and the required results. Current research focus is therefore the development of faster CFD methods such as FFD, ROM or machine learning. Further research is needed here, especially to assess which method is suitable for a certain accuracy standard and the respective application.

Automated model generation has only been possible in some areas so far, as the current implementation of BIM in software requires a high level of manual effort to generate physically correct models with a suitable LOD. Here, compatibility between different software and the implementation of standards is an important direction for future research.

References

- [1] W. Tian, X. Han, W. Zuo, and M. D. Sohn, “Building energy simulation coupled with cfd for indoor environment: A critical review and recent applications,” *Energy and Buildings*, vol. 165, pp. 184–199, 2018, ISSN: 0378-7788. DOI: 10.1016/j.enbuild.2018.01.046.
- [2] P. de Wilde, *Building performance analysis*. Place of publication not identified: Wiley, 2018, ISBN: 1119341906. DOI: 10.1002/9781119341901.
- [3] D. B. Crawley, J. W. Hand, M. Kummert, and B. T. Griffith, “Contrasting the capabilities of building energy performance simulation programs,” *Building and Environment*, vol. 43, no. 4, pp. 661–673, 2008, ISSN: 03601323. DOI: 10.1016/j.buildenv.2006.10.027.
- [4] Crawley, “Energyplus: Energy simulation program,” *ASHRAE Journal*, vol. 42, no. 4, p. 49, 2000.
- [5] *Trnsys 17 a transient system simulation program, volume 4: Mathematical reference, solar energy laboratory, university of wisconsin-madison, transsolar energietechnik gmbh, cstb-centre scientifique et technique du bâtiment, tess—thermal energy systems specialists*.
- [6] P. A. Strachan, G. Kokogiannakis, and I. A. Macdonald, “History and development of validation with the esp-r simulation program,” *Building and Environment*, vol. 43, no. 4, pp. 601–609, 2008, ISSN: 03601323. DOI: 10.1016/j.buildenv.2006.06.025.
- [7] *IDA Indoor climate and energy*. 2013.
- [8] X. Shan, W. Xu, Y.-K. Lee, and W.-Z. Lu, “Evaluation of thermal environment by coupling cfd analysis and wireless-sensor measurements of a full-scale room with cooling system,” *Sustainable Cities and Society*, vol. 45, pp. 395–405, 2019, ISSN: 2210-6707. DOI: 10.1016/j.scs.2018.12.011.
- [9] M. Shirzadi, P. A. Mirzaei, and M. Naghashzadegan, “Development of an adaptive discharge coefficient to improve the accuracy of cross-ventilation airflow calculation in building energy simulation tools,” *Building and Environment*, vol. 127, pp. 277–290, 2018, ISSN: 03601323. DOI: 10.1016/j.buildenv.2017.10.019.
- [10] V. Harish and A. Kumar, “A review on modeling and simulation of building energy systems,” *Renewable and Sustainable Energy Reviews*, vol. 56, pp. 1272–1292, 2016, ISSN: 1364-0321. DOI: 10.1016/j.rser.2015.12.040.
- [11] S. Gilani, H. Montazeri, and B. Blocken, “Cfd simulation of stratified indoor environment in displacement ventilation: Validation and sensitivity analysis,” *Building and Environment*, vol. 95, pp. 299–313, 2016, ISSN: 03601323. DOI: 10.1016/j.buildenv.2015.09.010.
- [12] M. T. Kahsay, G. T. Bitsuamlak, and F. Tariku, “Cfd simulation of external chtc on a high-rise building with and without façade appurtenances,” *Building and Environment*, vol. 165, p. 106350, 2019, ISSN: 03601323. DOI: 10.1016/j.buildenv.2019.106350.
- [13] R. Ramponi and B. Blocken, “Cfd simulation of cross-ventilation for a generic isolated building: Impact of computational parameters,” *Building and Environment*, vol. 53, pp. 34–48, 2012, ISSN: 03601323. DOI: 10.1016/j.buildenv.2012.01.004.
- [14] B. Blocken, T. Stathopoulos, J. Carmeliet, and J. L. Hensen, “Application of computational fluid dynamics in building performance simulation for the outdoor environment: An overview,” *Journal of Building Performance Simulation*, vol. 4, no. 2, pp. 157–184, 2011, ISSN: 1940-1507. DOI: 10.1080/19401493.2010.513740.

- [15] Q. CHEN, “Ventilation performance prediction for buildings: A method overview and recent applications,” *Building and Environment*, vol. 44, no. 4, pp. 848–858, 2009, ISSN: 03601323. DOI: 10.1016/j.buildenv.2008.05.025.
- [16] H. Qiao, X. Han, S. Nabi, and C. R. Laughman, “Coupled simulation of a room air-conditioner with cfd models for indoor environment,” in *Proceedings of the 13th International Modelica Conference, Regensburg, Germany, March 4–6, 2019*, ser. Linköping Electronic Conference Proceedings, Linköping University Electronic Press, 2019, pp. 265–274. DOI: 10.3384/ecp19157265.
- [17] X. Shi and W. Yang, “Performance-driven architectural design and optimization technique from a perspective of architects,” *Automation in Construction*, vol. 32, pp. 125–135, 2013, ISSN: 0926-5805. DOI: 10.1016/j.autcon.2013.01.015.
- [18] S.-J. Cao, “Challenges of using cfd simulation for the design and online control of ventilation systems,” *Indoor and Built Environment*, vol. 28, no. 1, pp. 3–6, 2019, ISSN: 1423-0070. DOI: 10.1177/1420326X18810568.
- [19] Marzullo, “A comparative study of computational algorithms used in the automatic generation of reduced-order models from cfd simulations,” 2017. DOI: 10.13140/RG.2.2.16929.92008.
- [20] V. Machairas, A. Tsangrassoulis, and K. Axarli, “Algorithms for optimization of building design: A review,” *Renewable and Sustainable Energy Reviews*, vol. 31, pp. 101–112, 2014, ISSN: 1364-0321. DOI: 10.1016/j.rser.2013.11.036.
- [21] A.-T. Nguyen, S. Reiter, and P. Rigo, “A review on simulation-based optimization methods applied to building performance analysis,” *Applied Energy*, vol. 113, pp. 1043–1058, 2014, ISSN: 0306-2619. DOI: 10.1016/j.apenergy.2013.08.061.
- [22] R. Evins, “A review of computational optimisation methods applied to sustainable building design,” *Renewable and Sustainable Energy Reviews*, vol. 22, pp. 230–245, 2013, ISSN: 1364-0321. DOI: 10.1016/j.rser.2013.02.004.
- [23] Z. Zhai and Q. Chen, “Solution characters of iterative coupling between energy simulation and cfd programs,” *Energy and Buildings*, vol. 35, no. 5, pp. 493–505, 2003, ISSN: 0378-7788. DOI: 10.1016/S0378-7788(02)00156-1.
- [24] W. Tian and W. Zuo, “Literature review and research needs to couple building energy and airflow simulation,” 2013.
- [25] M. Rodríguez-Vázquez, I. Hernández-Pérez, J. Xamán, Y. Chávez, M. Gijón-Rivera, and J. M. Belman-Flores, “Coupling building energy simulation and computational fluid dynamics: An overview,” *Journal of Building Physics*, vol. 44, no. 2, pp. 137–180, 2020, ISSN: 1744-2591. DOI: 10.1177/1744259120901840.
- [26] C. Gomes, C. Thule, D. Broman, P. G. Larsen, and H. Vangheluwe, “Co-simulation: A survey,” *ACM Computing Surveys*, vol. 51, no. 3, pp. 1–33, 2018, ISSN: 0360-0300. DOI: 10.1145/3179993.
- [27] Michael Wetter and Christoph van Treeck, *Iea ebc annex 60: New generation computing tools for building and community energy systems*, 2017.
- [28] G. Schweiger, C. Gomes, G. Engel, I. Hafner, J. Schoeggel, A. Posch, and T. Nouidui, “An empirical survey on co-simulation: Promising standards, challenges and research needs,” *Simulation Modelling Practice and Theory*, vol. 95, pp. 148–163, 2019, ISSN: 1569-190X. DOI: 10.1016/J.SIMPAT.2019.05.001.
- [29] M. Mitterhofer, G. F. Schneider, S. Stratbücker, and K. Sedlbauer, “An fmi-enabled methodology for modular building performance simulation based on semantic web technologies,” *Building and Environment*, vol. 125, pp. 49–59, 2017, ISSN: 03601323. DOI: 10.1016/j.buildenv.2017.08.021.

- [30] T. Nouidui, M. Wetter, and W. Zuo, “Functional mock-up unit for co-simulation import in energyplus,” *Journal of Building Performance Simulation*, vol. 7, no. 3, pp. 192–202, 2014, ISSN: 1940-1507. DOI: 10.1080/19401493.2013.808265.
- [31] C. Nytsch-Geusen, J. Huber, M. Ljubijunkic, and J. Rädler, “Modelica-buildingsystems – a simulation library of complex building energy systems,” *BauSIM, Berlin*, 2012.
- [32] M. Wetter, “A modelica-based model library for building energy and control systems,” 2009.
- [33] *Tools — functional mock-up interface*, 25.02.2021.
- [34] W. Zuo, M. Wetter, W. Tian, D. Li, M. Jin, and Q. Chen, “Coupling indoor airflow, hvac, control and building envelope heat transfer in the modelicabuildingslibrary,” *Journal of Building Performance Simulation*, vol. 9, no. 4, pp. 366–381, 2015, ISSN: 1940-1507. DOI: 10.1080/19401493.2015.1062557.
- [35] Z. Zhai, Q. Chen, P. Haves, and J. H. Klems, “On approaches to couple energy simulation and computational fluid dynamics programs,” *Building and Environment*, vol. 37, no. 8-9, pp. 857–864, 2002, ISSN: 03601323. DOI: 10.1016/S0360-1323(02)00054-9.
- [36] *Ansys fluent 15.0 user’s guide*, 2013.
- [37] M. Jin, W. Zuo, and Q. Chen, “Simulating natural ventilation in and around buildings by fast fluid dynamics,” *Numerical Heat Transfer, Part A: Applications*, vol. 64, no. 4, pp. 273–289, 2013, ISSN: 1521-0634. DOI: 10.1080/10407782.2013.784131.
- [38] S. Kato, “Review of airflow and transport analysis in building using cfd and network model,” *Japan Architectural Review*, vol. 1, no. 3, pp. 299–309, 2018, ISSN: 2475-8876. DOI: 10.1002/2475-8876.12051.
- [39] N. Morozova, R. Capdevila, and F. Trias, *Towards Real-Time CFD Simulation of Indoor Environment*. 2018.
- [40] M. L. Hosain and R. B. Fdhila, “Literature review of accelerated cfd simulation methods towards online application,” *Energy Procedia*, vol. 75, pp. 3307–3314, 2015, ISSN: 1876-6102. DOI: 10.1016/j.egypro.2015.07.714.
- [41] Z. Feng, C. W. Yu, and S.-J. Cao, “Fast prediction for indoor environment: Models assessment,” *Indoor and Built Environment*, vol. 28, no. 6, pp. 727–730, 2019, ISSN: 1423-0070. DOI: 10.1177/1420326X19852450.
- [42] W. Tian, T. A. Sevilla, W. Zuo, and M. D. Sohn, “Coupling fast fluid dynamics and multizone airflow models in modelica buildings library to simulate the dynamics of hvac systems,” *Building and Environment*, vol. 122, pp. 269–286, 2017, ISSN: 03601323. DOI: 10.1016/j.buildenv.2017.06.013.
- [43] V. Harish and A. Kumar, “Reduced order modeling and parameter identification of a building energy system model through an optimization routine,” *Applied Energy*, vol. 162, pp. 1010–1023, 2016, ISSN: 0306-2619. DOI: 10.1016/j.apenergy.2015.10.137.
- [44] W. Liu, M. Jin, C. Chen, R. You, and Q. Chen, “Implementation of a fast fluid dynamics model in openfoam for simulating indoor airflow,” *Numerical Heat Transfer, Part A: Applications*, vol. 69, no. 7, pp. 748–762, 2016, ISSN: 1521-0634. DOI: 10.1080/10407782.2015.1090780.
- [45] W. Tian, T. A. Sevilla, and W. Zuo, “A systematic evaluation of accelerating indoor airflow simulations using cross-platform parallel computing,” *Journal of Building Performance Simulation*, vol. 10, no. 3, pp. 243–255, 2016, ISSN: 1940-1507. DOI: 10.1080/19401493.2016.1212933.

- [46] D. T. Mullen, M. M. Keane, M. Geron, and R. Monaghan, “Automatic extraction of reduced-order models from cfd simulations for building energy modelling,” *Energy and Buildings*, vol. 99, pp. 313–326, 2015, ISSN: 0378-7788. DOI: 10.1016/j.enbuild.2015.04.015.
- [47] W. Tian, T. A. Sevilla, D. Li, W. Zuo, and M. Wetter, “Fast and self-learning indoor airflow simulation based on in situ adaptive tabulation,” *Journal of Building Performance Simulation*, vol. 11, no. 1, pp. 99–112, 2017, ISSN: 1940-1507. DOI: 10.1080/19401493.2017.1288761.
- [48] H. Zhou, Y. C. Soh, and X. Wu, “Integrated analysis of cfd data with k-means clustering algorithm and extreme learning machine for localized hvac control,” *Applied Thermal Engineering*, vol. 76, pp. 98–104, 2015, ISSN: 1359-4311. DOI: 10.1016/j.applthermaleng.2014.10.004.
- [49] Q. Fang, Z. Li, Y. Wang, M. Song, and J. Wang, “A neural-network enhanced modeling method for real-time evaluation of the temperature distribution in a data center,” *Neural Computing and Applications*, vol. 31, no. 12, pp. 8379–8391, 2019, ISSN: 0941-0643. DOI: 10.1007/s00521-019-04508-y.
- [50] M. Lee, G. Park, H. Jang, and C. Kim, “Development of building cfd model design process based on bim,” *Applied Sciences*, vol. 11, no. 3, p. 1252, 2021. DOI: 10.3390/app11031252.
- [51] G. Park, C. Kim, M. Lee, and C. Choi, “Building geometry simplification for improving mesh quality of numerical analysis model,” *Applied Sciences*, vol. 10, no. 16, p. 5425, 2020. DOI: 10.3390/app10165425.
- [52] F. Xu, J. Yang, and X. Zhu, “A comparative study on the difference of cfd simulations based on a simplified geometry and a more refined bim based geometry,” *AIP Advances*, vol. 10, no. 12, p. 125318, 2020. DOI: 10.1063/5.0031907.
- [53] Z. Ali, J. Tyacke, R. Watson, P. G. Tucker, and S. Shahpar, “Efficient preprocessing of complex geometries for cfd simulations,” *International Journal of Computational Fluid Dynamics*, vol. 33, no. 3, pp. 98–114, 2019, ISSN: 1029-0257. DOI: 10.1080/10618562.2019.1606421.
- [54] M. Lee, G. Park, C. Park, and C. Kim, “Improvement of grid independence test for computational fluid dynamics model of building based on grid resolution,” *Advances in Civil Engineering*, vol. 2020, pp. 1–11, 2020, ISSN: 1687-8086. DOI: 10.1155/2020/8827936.
- [55] J. Lai, H. Li, and Z. Tian, “Cpu/gpu heterogeneous parallel cfd solver and optimizations,” in *Proceedings of the 2018 International Conference on Service Robotics Technologies*, Unknown, Ed., ser. ACM Other conferences, New York, NY: ACM, 2018, pp. 88–92, ISBN: 9781450364348. DOI: 10.1145/3208833.3208847.
- [56] S. Lee, J. Gounley, A. Randles, and J. S. Vetter, “Performance portability study for massively parallel computational fluid dynamics application on scalable heterogeneous architectures,” *Journal of Parallel and Distributed Computing*, vol. 129, pp. 1–13, 2019, ISSN: 07437315. DOI: 10.1016/j.jpdc.2019.02.005.
- [57] G. Chen, J. Wang, Z. Zhang, Z. Tian, L. Li, H. Kang, and Y. Jin, “Distributed-parallel cfd computation for all fuel assemblies in pwr core,” *Annals of Nuclear Energy*, vol. 141, p. 107340, 2020, ISSN: 0306-4549. DOI: 10.1016/j.anucene.2020.107340.
- [58] C. Nytsch-Geusen, J. Rädler, M. Thorade, and C. Ribas Tugores, “Bim2modelica - an open source toolchain for generating and simulating thermal multi-zone building models by using structured data from bim models,” in *Proceedings of the 13th International Modelica Conference, Regensburg, Germany, March 4–6, 2019*, ser. Linköping Electronic Conference Proceedings, Linköping University Electronic Press, 2019, pp. 33–38. DOI: 10.3384/ecp1915733.

- [59] A. Andriamamonjy, D. Saelens, and R. Klein, “An automated ifc-based workflow for building energy performance simulation with modelica,” *Automation in Construction*, vol. 91, pp. 166–181, 2018, ISSN: 0926-5805. DOI: 10.1016/j.autcon.2018.03.019.
- [60] S. Pinheiro, R. Wimmer, J. O’Donnell, S. Muhic, V. Bazjanac, T. Maile, J. Frisch, and C. van Treeck, “Mvd based information exchange between bim and building energy performance simulation,” *Automation in Construction*, vol. 90, pp. 91–103, 2018, ISSN: 0926-5805. DOI: 10.1016/j.autcon.2018.02.009.
- [61] G. N. Lilis, G. I. Giannakis, and D. V. Rovas, “Automatic generation of second-level space boundary topology from ifc geometry inputs,” *Automation in Construction*, vol. 76, pp. 108–124, 2017, ISSN: 0926-5805. DOI: 10.1016/j.autcon.2016.08.044.
- [62] W. Jeong, J. B. Kim, M. J. Clayton, J. S. Haberl, and W. Yan, “A framework to integrate object-oriented physical modelling with building information modelling for building thermal simulation,” *Journal of Building Performance Simulation*, vol. 9, no. 1, pp. 50–69, 2016, ISSN: 1940-1507. DOI: 10.1080/19401493.2014.993709.
- [63] B. Fabritius and G. Tabor, “Improving the quality of finite volume meshes through genetic optimisation,” *Engineering with Computers*, vol. 32, no. 3, pp. 425–440, 2015, ISSN: 1435-5663. DOI: 10.1007/s00366-015-0423-0.
- [64] M. Tomac and D. Eller, “From geometry to cfd grids—an automated approach for conceptual design,” *Progress in Aerospace Sciences*, vol. 47, no. 8, pp. 589–596, 2011, ISSN: 03760421. DOI: 10.1016/j.paerosci.2011.08.005.
- [65] M. Delavar, G. T. Bitsuamlak, J. K. Dickinson, and L. M. F. Costa, “Automated bim-based process for wind engineering design collaboration,” *Building Simulation*, vol. 13, no. 2, pp. 457–474, 2019, ISSN: 1996-8744. DOI: 10.1007/s12273-019-0589-2.
- [66] J. T. O’Donnell, M. van Dessel, and T. Maile, “Bim to building energy performance simulation: An evaluation of current industry transfer processes,” in *Proceedings of Building Simulation 2019: 16th Conference of IBPSA*, V. Corrado, E. Fabrizio, A. Gasparella, and F. Patuzzi, Eds., ser. Building Simulation Conference proceedings, IBPSA, 2020, pp. 92–99, ISBN: 9781775052012. DOI: 10.26868/25222708.2019.210241.
- [67] T. Hong, Y. Chen, Z. Belafi, and S. D’Oca, “Occupant behavior models: A critical review of implementation and representation approaches in building performance simulation programs,” *Building Simulation*, vol. 11, no. 1, pp. 1–14, 2018, ISSN: 1996-8744. DOI: 10.1007/s12273-017-0396-6.
- [68] I. Gaetani, P.-J. Hoes, and J. L. Hensen, “Occupant behavior in building energy simulation: Towards a fit-for-purpose modeling strategy,” *Energy and Buildings*, vol. 121, pp. 188–204, 2016, ISSN: 0378-7788. DOI: 10.1016/j.enbuild.2016.03.038.
- [69] E. Kamel and A. M. Memari, “Review of bim’s application in energy simulation: Tools, issues, and solutions,” *Automation in Construction*, vol. 97, pp. 164–180, 2019, ISSN: 0926-5805. DOI: 10.1016/j.autcon.2018.11.008.
- [70] D. Ladenhauf, K. Battisti, R. Berndt, E. Eggeling, D. W. Fellner, M. Gratzl-Michlmair, and T. Ullrich, “Computational geometry in the context of building information modeling,” *Energy and Buildings*, vol. 115, pp. 78–84, 2016, ISSN: 0378-7788. DOI: 10.1016/j.enbuild.2015.02.056.
- [71] M. H. Elnabawi and N. Hamza, “Investigating building information model (bim) to building energy simulation (bes): Interoperability and simulation results,” *IOP Conference Series: Earth and Environmental Science*, vol. 397, p. 012013, 2019, ISSN: 1755-1315. DOI: 10.1088/1755-1315/397/1/012013.