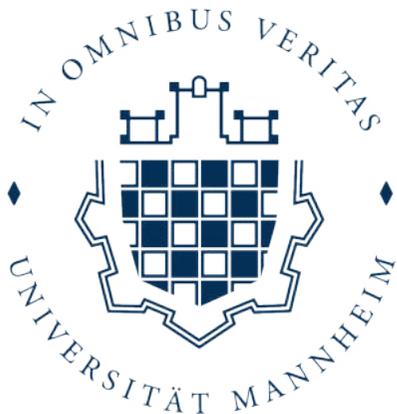


Digitalisierung als Treiber für Change Prozesse an Hochschulen



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UNESCO Co-Chair on Data Science in Higher Education Learning and
Teaching, Curtin University

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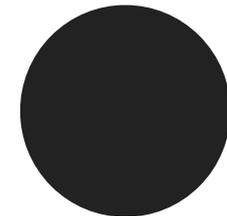
Chair



**Curtin
University**

...
Artificial General Intelligence (20XX)
Generative Artificial Intelligence (2022)
Artificial Intelligence Education (2019)
Adaptive learning (2015)
Digital Badges (2013)
Learning Analytics (2011)
Massive Open Online Courses (2008)
Technology-enhanced learning (2004)
Virtual Campus (2000)
Learning Management System (1999)
Virtual University (1999)
Open Learning (1995)
E-Learning (1993)
Computer-mediated learning (1990)
Computer-assisted learning (1985)
Computer-based learning (1980)
Computer-assisted instruction (1960)
Computer-assisted instruction (1960)
Computer-assisted learning (1980)
Computer-assisted learning (1985)
Computer-assisted learning (1990)

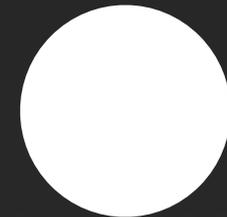
Die Technik rast, die
(Hochschul-) **Didaktik humpelt
hinterher.**



3

Seel, N. M., & Ifenthaler, D. (2009). Online lernen und lehren. Reinhardt Verlag. <https://doi.org/10.36198/9783838532882>

Folgen wir – im Kontext der
Hochschulbildung – **blind** einem nicht
hinterfragten **technik-optimistischen**
Hype?





Bereitstellung von **hochwertiger tertiären Bildung** mit angemessener Finanzierung und der **Nutzung von Technologie** entwickeln, einschließlich Internet, Massive Open Online Courses (MOOCs) und anderen Modalitäten, die den *akzeptierten Qualitätsstandards* entsprechen, **um den Bildungszugang zu verbessern.**

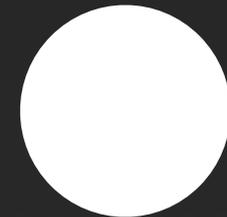




Warum sollte die *Digitalisierung der Hochschulbildung* bei der **Überwindung von Bildungsungleichheit** und **Benachteiligung** erfolgreicher sein als frühere Maßnahmen und Reformen?



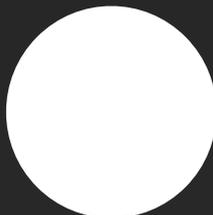
Selwyn, N. (2016). *Is technology good for education?* Polity Press.



Continent	% Course participants (N)	% Forum participants (N)
North America	32% (15,481)	32% (1,145)
South America	7% (3496)	10% (372)
Europe	28% (13,412)	25% (918)
Asia	26% (12,723)	24% (879)
Africa	5% (2453)	7% (234)
Oceania	2% (1125)	2% (83)

Highest level of educational attainment	% Course participants with this level (N)	% Forum participants (N)
Some high school	2% (103)	1% (26)
Completed high school	4% (282)	3% (66)
Some college	12% (805)	10% (204)
Bachelor's degree	42% (3,050)	43% (849)
Master's/professional degree	36% (2,832)	40% (781)
Doctorate	4% (265)	3% (58)

Gillani, N., & Eynon, R. (2014). Communication patterns in massively open online courses. *The Internet and Higher Education*, 23, 18–26. <https://doi.org/10.1016/j.iheduc.2014.05.004>



Die Abschlussquoten reichen von 0.9 % bis 36.1 %, mit einem Medianwert von 6.5 % – wobei **Abschlussquoten von 5 % typisch sind.**



Jordan, K. (2014). Initial trends in enrolment and completion of Massive Open Online Courses. *International Review of Research in Open and Distance Learning*, 15(1). <https://doi.org/10.19173/irrodl.v15i1.1651>

Course title	Provider	Institution	Section A (≤ 15 pts.)	Section B (≤ 30 pts.)	Section C ^a (≤ 30 pts.)	Overall score (≤ 75 pts.)
Business Foundations	edX	University of British Columbia	14	24	18	56
Commercialization of Social Enterprises	Future Learn	Free University of Bruxelles	11	27	18	56
Operations Management	Coursera	University of Illinois	15	23	18	56
Innovation Management	Future Learn	University of Leeds	13	27	16	56
Fundamentals of Project Planning and Management	Future Learn	University of Virginia	14	22	20	56
Reputation Management in a Digital World	edX	Curtin University	14	24	16	54
Business Model Implementation	edX	Delft University of Technology	13	24	16	53
Global Impact: Cultural Psychology	Coursera	University of Illinois at Urbana-Champaign	13	22	18	53
Leading and Managing People-Centered Change	Future Learn	Durham University	11	24	18	53
Ethics for Managers	Canvas Network	Santa Clara University	14	21	18	53

Note. ^a Raw points in Section C weighted with factor 2.

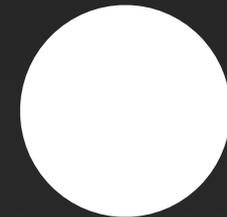
Section	(A)		(B)		(C) ^a		Total ^b	
	<i>M</i>	<i>SD</i>	<i>M</i>	<i>SD</i>	<i>M</i>	<i>SD</i>	<i>M</i>	<i>SD</i>
Provider/Platform								
Canvas Network (<i>n</i> = 10)	12.3	1.7	17.0	3.8	13.2	4.2	42.5	8.1
Coursera (<i>n</i> = 25)	11.5	2.1	16.2	5.6	12.2	3.8	39.9	10.0
edX (<i>n</i> = 24)	12.3	2.2	18.5	4.3	13.2	2.3	43.9	6.8
FutureLearn (<i>n</i> = 19)	12.1	1.8	19.3	4.4	13.7	2.8	45.0	6.6
iversity (<i>n</i> = 8)	11.3	2.4	18.0	3.3	16.0	1.9	45.3	3.2
Open2Study (<i>n</i> = 10)	9.9	.3	7.5	.9	11.4	2.5	28.8	3.5
Udacity (<i>n</i> = 5)	8.8	1.6	9.8	4.6	8.4	3.6	27.0	9.5
<i>F</i> -value; η^2	3.93**; $\eta^2 = .200$		11.11**; $\eta^2 = .415$		3.94**; $\eta^2 = .201$		9.19**; $\eta^2 = .370$	
Region								
North America (<i>n</i> = 38)	11.7	2.4	17.2	5.2	12.3	3.8	41.3	10.0
Europe (<i>n</i> = 37)	11.9	1.8	17.8	4.9	14.1	2.9	43.7	7.5
Asia (<i>n</i> = 8)	11.1	2.6	14.1	4.9	11.8	2.9	37.0	8.8
Australia (<i>n</i> = 17)	10.7	1.6	12.3	6.4	12.2	2.3	35.2	9.1
Africa (<i>n</i> = 1)	12.0	--	16.0	--	8.00	--	36.0	--
<i>F</i> -value; η^2	1.050; $\eta^2 = .042$		3.795**; $\eta^2 = .137$		2.453*; $\eta^2 = .093$		3.123*; $\eta^2 = .115$	
Authoring Institution								
Academic (<i>n</i> = 86)	11.7	2.0	17.1	5.2	13.2	3.3	41.9	8.5
Nonacademic (<i>n</i> = 15)	10.7	2.4	12.2	5.9	12.2	3.3	33.9	10.6
<i>T</i> -value; η^2	1.659; $\eta^2 = .027$		3.266**; $\eta^2 = .097$		2.463*; $\eta^2 = .058$		3.274**; $\eta^2 = .098$	

Note. ^{a,b} Analysis based on weighted scores.
* $p < .05$. ** $p < .01$.

Egloffstein, M., Koegler, K., & Ifenthaler, D. (2019). Instructional quality of business MOOCs: Indicators and initial findings. *Online Learning*, 23(4), 85–105. <https://doi.org/10.24059/olj.v23i4.2091>



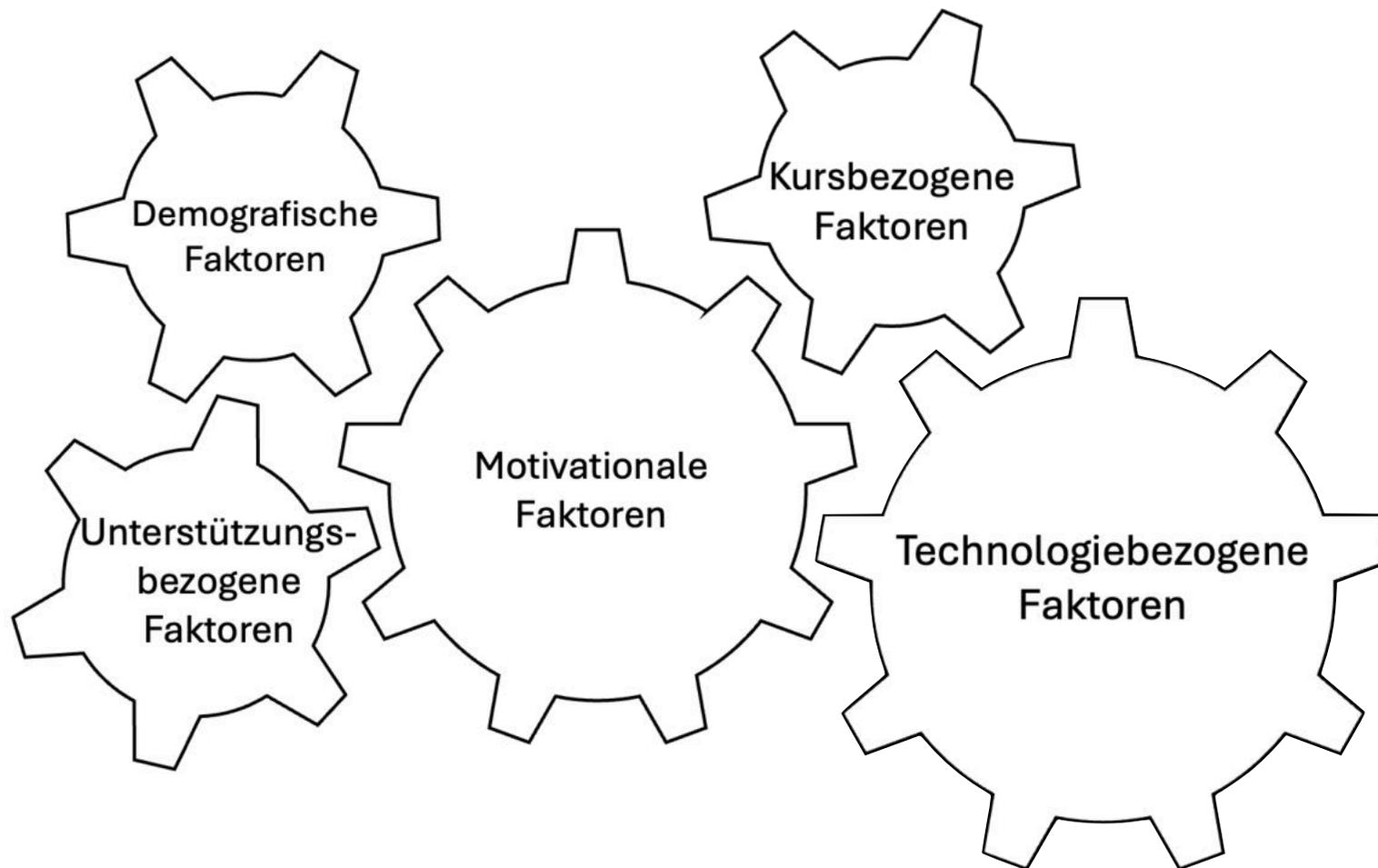
Die **sozialen Nachteile verschwinden nicht einfach**, wenn man über das Internet lernt.



11



McMillan Cottom, T. (2020). Where platform capitalism and racial capitalism meet: The sociology of race and racism in the digital society. *Sociology of Race and Ethnicity*, 6(4), 441–449. <https://doi.org/10.1177/2332649220949473>



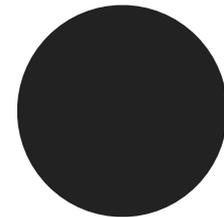
Rahmani, A. M., Groot, W., & Rahmani, H. (2024). Dropout in online higher education: a systematic literature review. *International Journal of Educational Technology in Higher Education*, 21, 19. <https://doi.org/10.1186/s41239-024-00450-9>

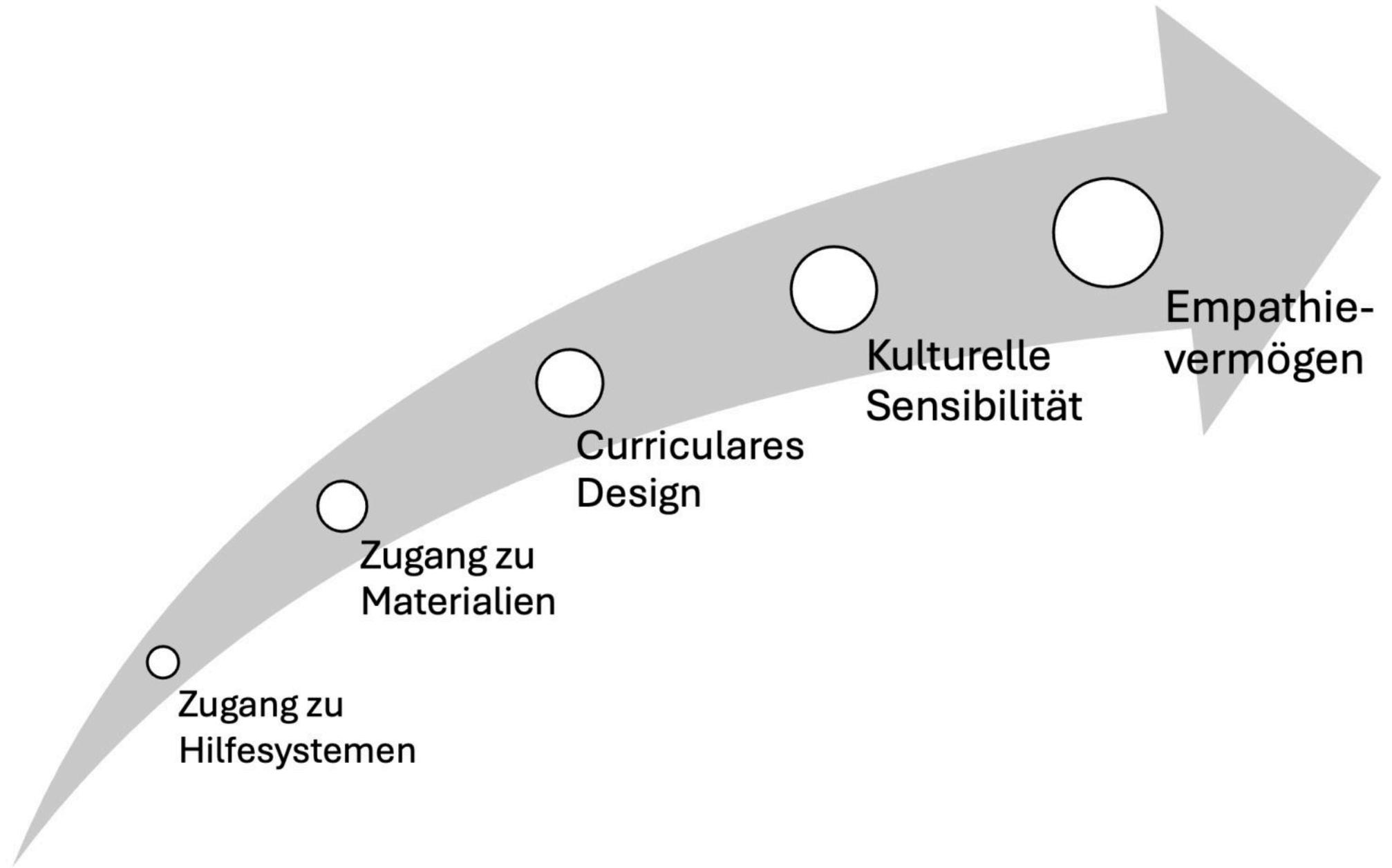


Die **soziale Ängstlichkeit von Studierenden** in digitalen Lernumgebungen an Hochschulen existiert zwischen Lernenden und Lernenden sowie zwischen Lernenden und Lehrenden.



Ifenthaler, D., Cooper, M., Daniela, L., & Sahin, M. (2023). Social anxiety in digital learning environments: an international perspective and call to action. *International Journal of Educational Technology in Higher Education*, 20, 50. <https://doi.org/10.1186/s41239-023-00419-0>

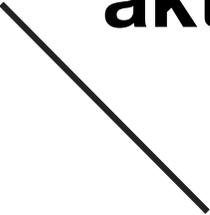




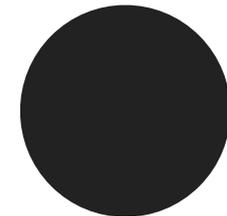
Tereshko, L. M., Weiss, M. J., Cross, S., & Neang, L. (2024). Culturally diverse student engagement in online higher education: A review. *Journal of Behavioral Education*. <https://doi.org/10.1007/s10864-024-09554-8>



Auf der Ebene der Beziehungen zwischen Lehrenden und Lernenden geht es darum, Dominanz durch **Dialog** zu ersetzen, Hierarchie durch **Kooperation** und **Kollegialität** sowie Passivität durch **aktives Lernen** und **Problemlösung**.

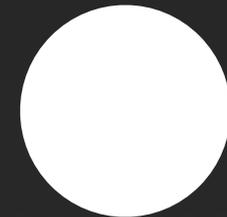


Lynch, K., & Baker, J. (2005). Equality in education: An equality of condition perspective. *Theory and Research in Education*, 3(2), 131–164. <https://doi.org/10.1177/1477878505053298>





Digitalisierung der Hochschulbildung kann
ein Teil der Lösung sein, um *soziale
Benachteiligung* zu überwinden, aber sie
sollte **niemals als *die* Lösung**
angesehen werden.



**Hochschulbildung
berechenbar
machen?**

**Nachhaltige digitale
Transformation der
Hochschulbildung?**

1

2

3

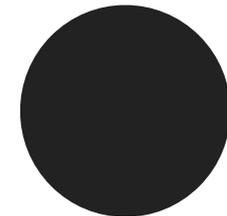
4

**Hochschulbildung
demokratischer
gestalten?**

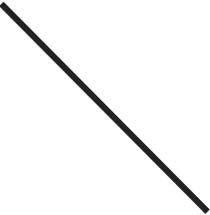
**Hochschulbildung
personalisierter
gestalten?**



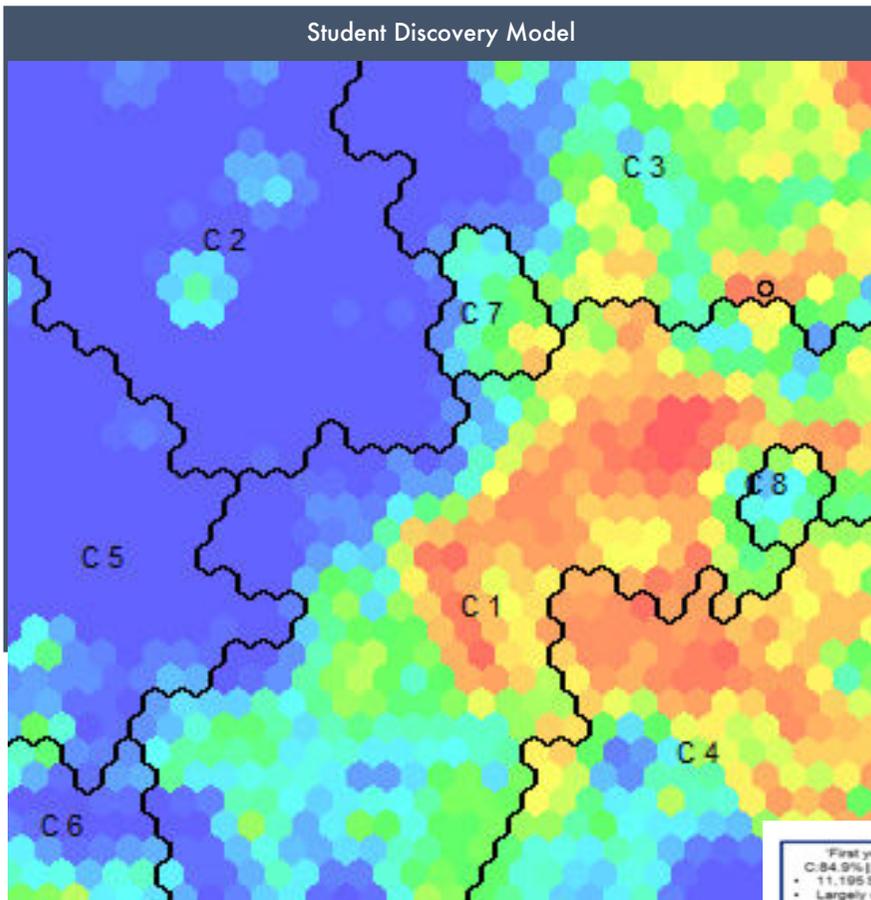
Im Gegensatz zu Menschen können
Data Analytics Ansätze eine um
vielfaches **größere Datenmengen in**
Echtzeit verarbeiten.



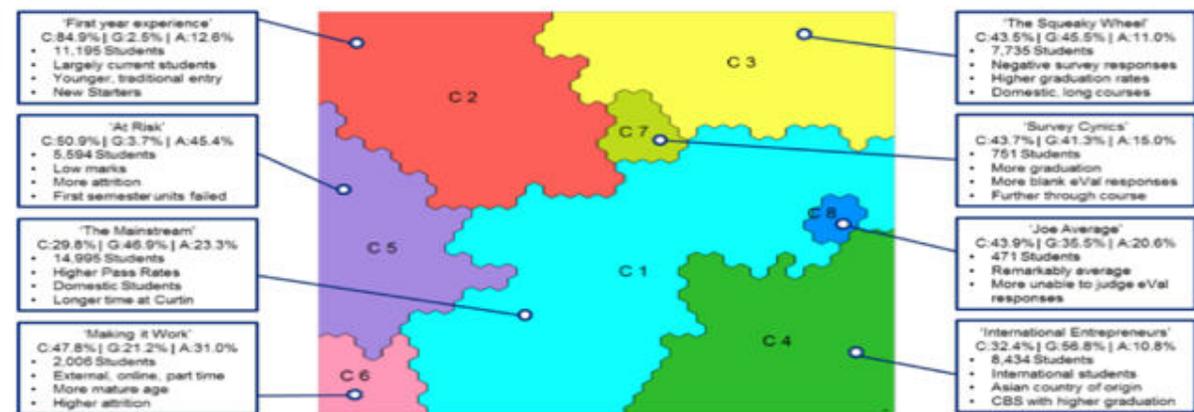
18



De Laat, M., Joksimovic, S., & Ifenthaler, D. (2020). Artificial intelligence, real-time feedback and workplace learning analytics to support in situ complex problem-solving: a commentary. *International Journal of Information and Learning Technology*, 37(5), 267–277. <https://doi.org/10.1108/IJILT-03-2020-0026>



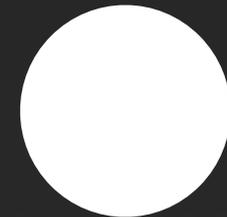
Data Analytics identifizieren **Zusammenhänge in großen Datenmengen** (Big Data), die für den Menschen nicht intuitiv zu erfassen oder zu verstehen wären, um Entscheidungen zu treffen.

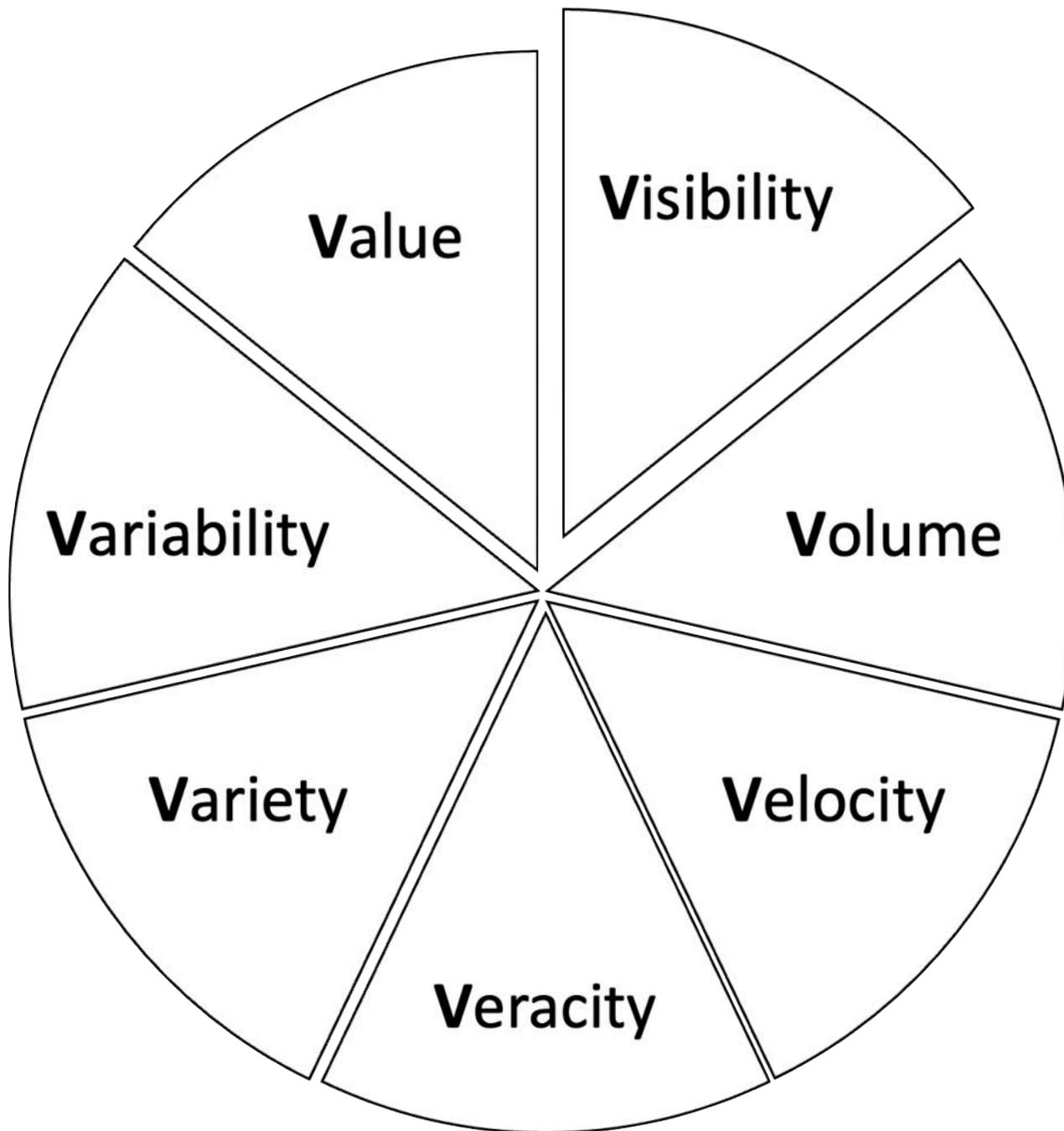


Henry, M., Gibson, D. C., Flodin, C., & Ifenthaler, D. (2020). Learning innovations for identifying and developing talent for university. In D. Parrish & J. Joyce-McCoach (Eds.), *Innovations in higher education - Cases on transforming and advancing practice* (pp. 1–17). IntechOpen. <https://doi.org/10.5772/intechopen.81380>



Daten werden immer **aktiv produziert**,
strukturiert und **analysiert** – Daten sind
nicht einfach natürliche und wesentliche
Elemente, die auf **neutrale und**
objektive Weise aus der Welt abstrahiert
werden.

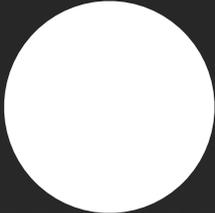




Niculescu, V. (2020). On the impact of high performance computing in big data analytics for medicine. *Applied Medical Informatics*, 42(1), 9–18.
<https://ami.info.umfcluj.ro/index.php/AMI/article/view/766>



Herausforderungen für die
Implementierung von Data Analytics in
Organisationen sind die **Interaktion** und
Fragmentation von Daten sowie deren
kontextuellen Eigenarten.



22



Gašević, D., Dawson, S., Rogers, T., & Gašević, D. (2016). Learning analytics should not promote one size fits all: The effects of instructional conditions in predicting academic success. *Internet and Higher Education*, 28, 68–84. <https://doi.org/10.1016/j.iheduc.2015.10.002>

Table 1. Student profile – comparison of institutions predicting pass/fail rates

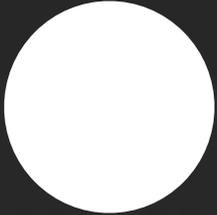
Institution	N	R ²	Adjusted R ²	R ² -SVR	Predictive accuracy (SVM)
UNI1	244494	0.4635	0.4633***	0.4889	0.817
UNI2	217039	0.4528	0.4526***	0.4603	0.796
UNI3	127218	0.431	0.4306***	0.4595	0.796
UNI4	114432	0.372	0.3716***	0.3807	0.766
UNI5	88026	0.4379	0.4374***	0.4430	0.807
UNI6	84510	0.3641	0.3635***	0.3530	0.763
UNI7	76278	0.434	0.4334***	0.4604	0.803
UNI8	73043	0.3718	0.3711***	0.3562	0.783
<i>SD</i>		0.096	0.097	0.126	0.024

Note. * $p < .05$, ** $p < .01$, *** $p < .001$

Ifenthaler, D., & Widanapathirana, C. (2014). Development and validation of a learning analytics framework: Two case studies using support vector machines. *Technology, Knowledge and Learning*, 19(1–2), 221–240. <https://doi.org/10.1007/s10758-014-9226-4>



Datenqualität kann nicht durch
Algorithmen gelöst werden.



24



Ifenthaler, D. (2022). Data mining and analytics in the context of workplace learning: benefits and affordances. In M. Goller, E. Kyndt, S. Paloniemi, & C. Damsa (Eds.), *Methods for researching professional learning and development. Challenges, applications, and empirical illustrations* (pp. 313–327). Springer. https://doi.org/10.1007/978-3-031-08518-5_14

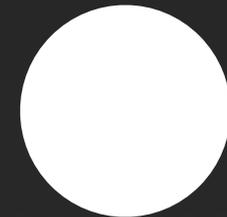
Table 2. Student profile – comparison of areas of study predicting pass/fail rates

Areas of study	<i>N</i>	R ²	Adjusted R ²	R ² -SVR	Predictive accuracy (SVM)
Arts & Humanities	386059	0.4299	0.4297	0.45039	0.799
Business	269410	0.4054	0.4053	0.4360	0.780
Education	157693	0.4887	0.4885	0.5049	0.824
Law & Justice	84663	0.4900	0.4896	0.5166	0.827
IT	57371	0.3732	0.3726	0.3586	0.776
Science & Engineering	57234	0.4228	0.422	0.4234	0.800
<i>SD</i>		0.107	0.107	0.129	0.027

Note. * $p < .05$, ** $p < .01$, *** $p < .001$

Ifenthaler, D., & Widanapathirana, C. (2014). Development and validation of a learning analytics framework: Two case studies using support vector machines. *Technology, Knowledge and Learning*, 19(1–2), 221–240. <https://doi.org/10.1007/s10758-014-9226-4>

Jede Form der **Quantifizierung**
menschlicher Aktivitäten ist
notwendigerweise **reduktionistisch**.



26

Gourlay, L. (2024). Surveillance and datafication in higher education: Documentation of the human. *Postdigital Science and Education*, 6, 1039–1048. <https://doi.org/10.1007/s42438-022-00352-x>

Table 2. Summary of learning analytics indicators mapped to three data profiles

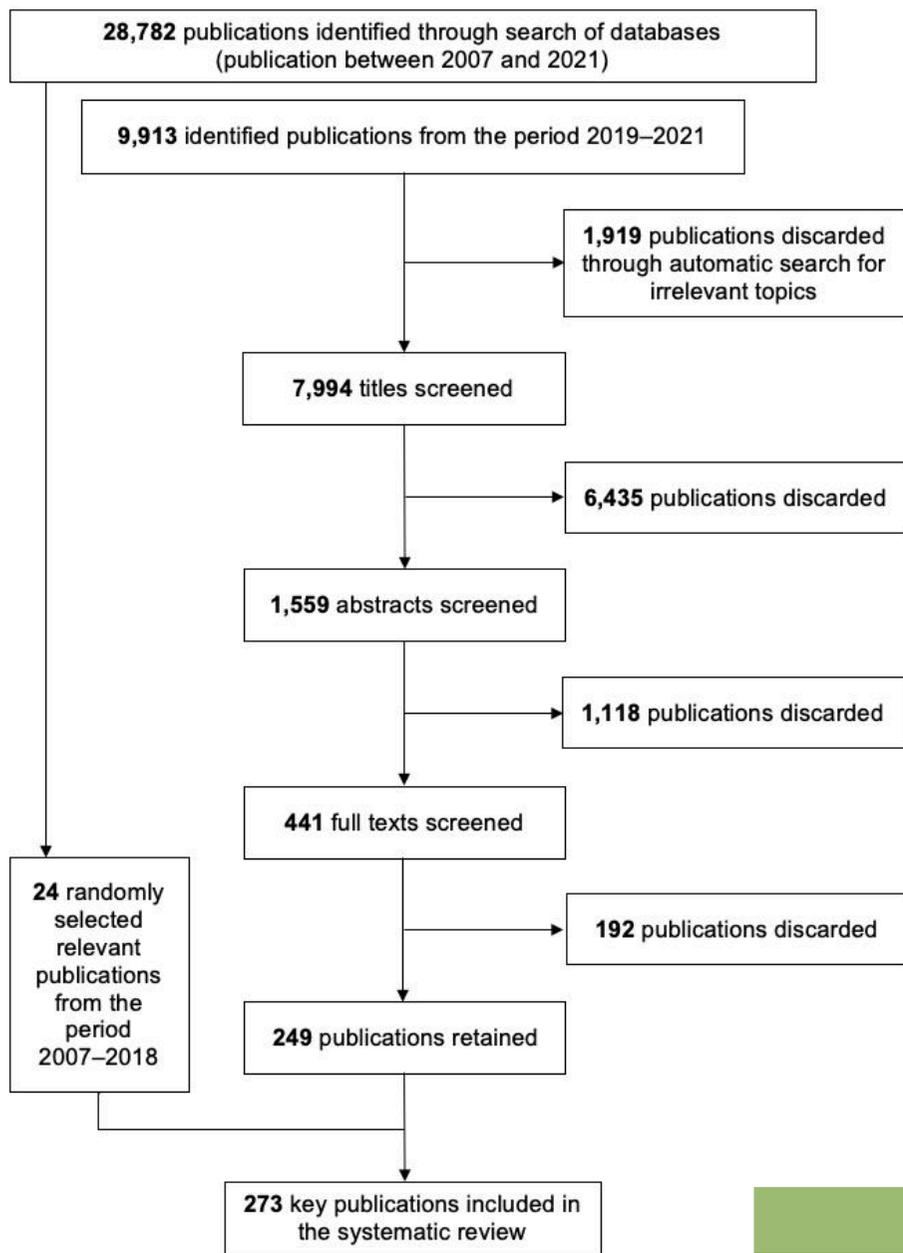
	Student profile	Learning profile	Curriculum profile
Students answers/ grades	N/A	Content access (video/ audio trace data) pen trace data (self-)assessment (score, grade, completion) data	N/A
Students social learning behaviour/ engagement	Prior academic performance prior competence/skills demographic background social behaviour trait self-report survey current workload study pattern	Course access (login) content access discussion/forum (length, quality) trace data engagement trace data (self-)assessment (score, grade, completion) data	N/A
At-risk/ low-performers	Prior academic performance prior competence/skills demographic background socioeconomic background academic goals technology preparedness Completed/ withdrawn courses motivation/interest prior learning behaviour prior academic institutions enrolment history/ mode/ load	Course access (login) content access assignment submission engagement trace data discussion/forum (length, quality) trace data (Self-)assessment (score, grade, completion) data final grade reflection/ feedback access social network usage	Course characteristics course survey
Student performance	Prior academic performance demographic background socioeconomic background enrolment history/ mode/ load counselling activities psychological test outcomes	(Self-)assessment (score, grade, completion) data final grade course access content access discussion/forum (length, quality) trace data engagement trace data	N/A
Course completion	Prior academic performance demographic background completed/ withdrawn courses enrolment history/ mode/ load	Course access (login) content access discussion/forum (length, quality) trace data engagement trace data (self-)assessment (score, grade, completion) data	N/A

Yau, J., & Ifenthaler, D. (2020). Reflections on different learning analytics indicators for supporting study success. *International Journal of Learning Analytics and Artificial Intelligence for Education*, 2(2), 4–23. <https://doi.org/10.3991/ijai.v2i2.15639>



daten-
getrieben

Ifenthaler, D. (2021). Learning analytics for school and system management. In OECD (Ed.), *OECD digital education outlook 2021: pushing the frontiers with artificial intelligence, blockchain and robots* (pp. 161–172). OECD Publishing. <https://doi.org/10.1787/d535b828-en>



GEFÖRDERT VOM



Bundesministerium
für Bildung
und Forschung

bibb Bundesinstitut für
Berufsbildung

**228 Indikatoren mit Fokus
auf den Kontext des Lernens**

Hemmler, Y., & Ifenthaler, D. (2022). Four perspectives on personalized and adaptive learning environments for workplace learning. In D. Ifenthaler & S. Seufert (Eds.), *Artificial intelligence education in the context of work* (pp. 27–39). Springer. https://doi.org/10.1007/978-3-031-14489-9_2

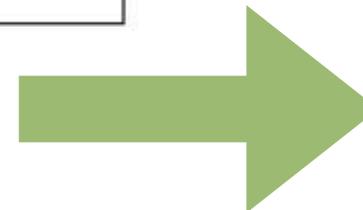
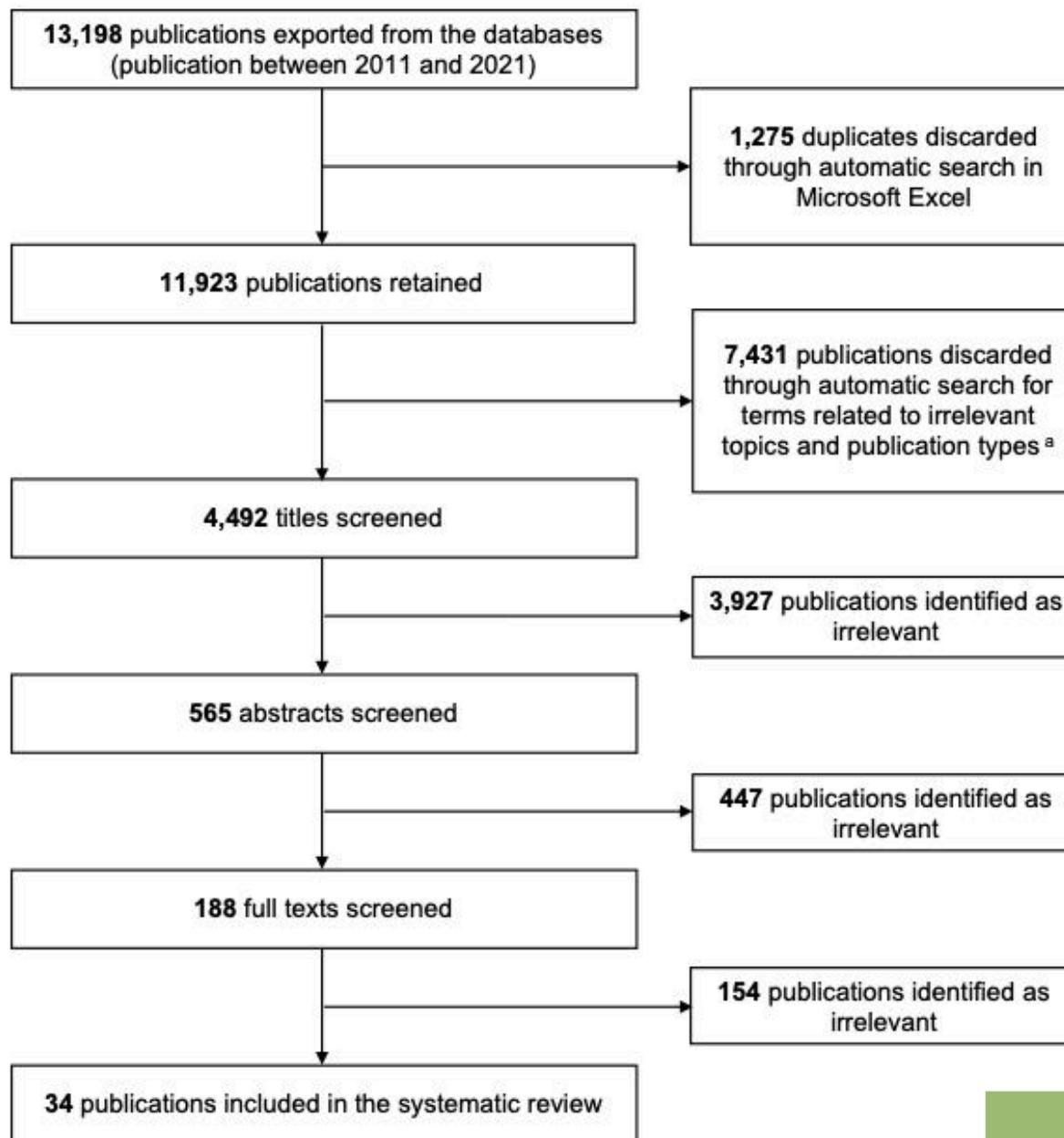


GEFÖRDERT VOM



Bundesministerium
für Bildung
und Forschung

bibb Bundesinstitut für
Berufsbildung



**43 Indikatoren
mit Fokus auf
Lernergebnisse**

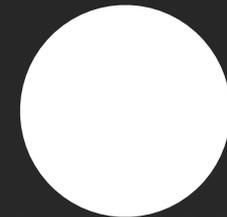
Hemmler, Y. M., Rasch, J., & Ifenthaler, D. (2023). A categorization of workplace learning goals for multi-stakeholder recommender systems: A systematic review. *TechTrends*, 67, 68–111. <https://doi.org/10.1007/s11528-022-00777-y>



Wir wissen relativ wenig darüber, wie sich die **(Lern-)Kulturen** und Prozesse an Universitäten auf die **Einführung von Datenpraktiken auswirken** können, insbesondere in Bezug auf Verwaltungsdaten.



Brooks, R., & Timms, J. (2024). Institutional constraints to higher education datafication: an English case study. *Higher Education*. <https://doi.org/10.1007/s10734-024-01363-2>





Bildungsdatenkompetenz

(Educational Data Literacy) ist ethisch verantwortliches sammeln, managen, analysieren, verstehen, interpretieren und anwenden von Daten aus dem Kontext des Lernen und Lehrens.



**Hochschulbildung
berechenbar
machen?**

**Nachhaltige digitale
Transformation der
Hochschulbildung?**

1

2

3

4

**Hochschulbildung
demokratischer
gestalten?**

**Hochschulbildung
personalisierter
gestalten?**

INTRODUCTION

If computerized instruction is ever to have a large impact on education, computer-assisted instruction (CAI) systems must have the flexibility and skill of a human teacher. In developing the SCHOLAR CAI system Carbonell (1) took a first step toward an intelligent tutorial CAI system. In SCHOLAR, knowledge was not stored as text, but in an interrelated network of facts and concepts, so that the knowledge could be used in a variety of ways. In short, the attempt was to structure information like a human knowledge, so that the program could use its knowledge as flexibly as a human tutor does.

Personalisiertes Lernen

bezeichnet die Anpassung
des Lernangebots an die

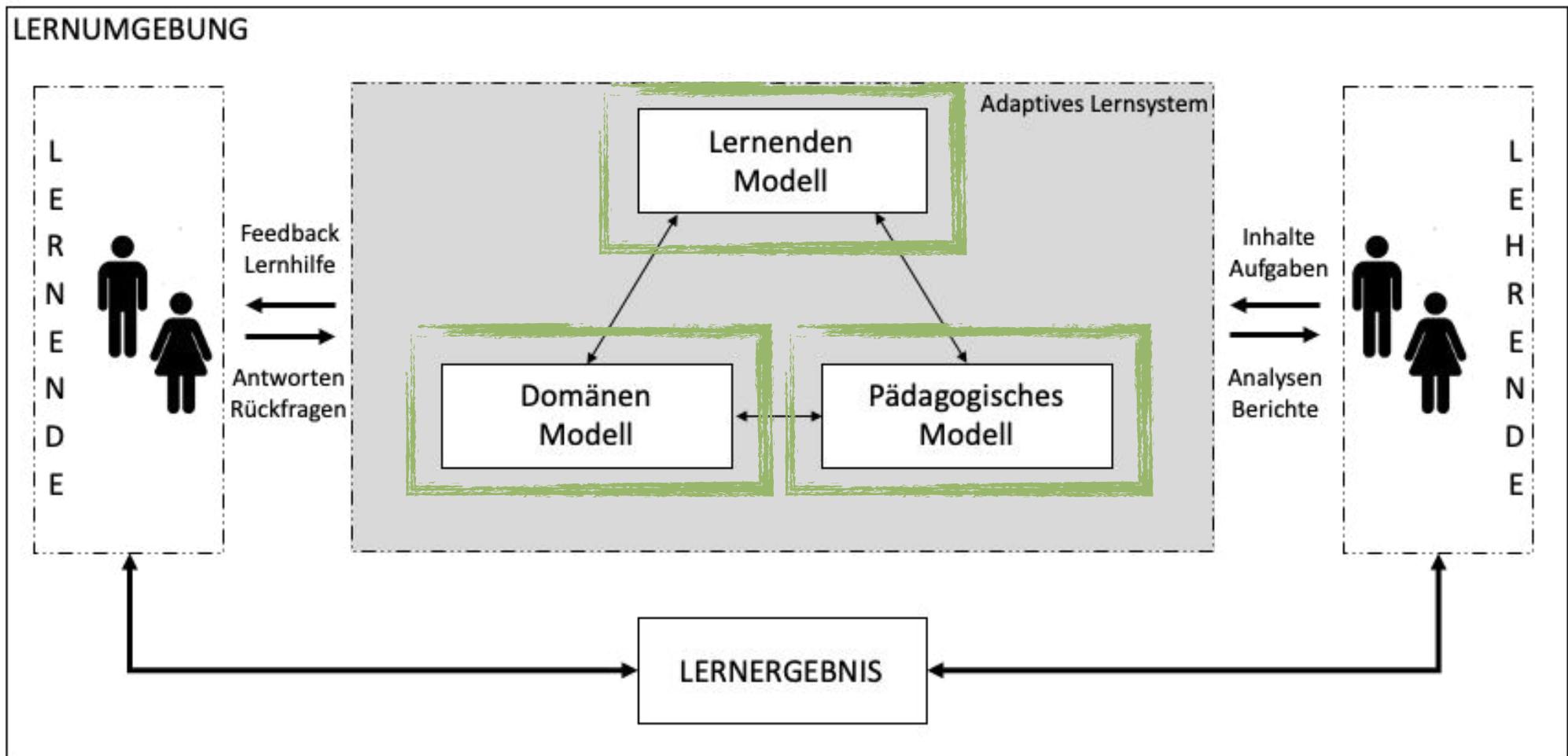
individuellen

Voraussetzungen und

Bedürfnisse (z. B.
Erwartungen, Vorkenntnisse,
Fähigkeiten, Präferenzen,
Lernstrategien) der
Lernenden.

Definitionen zu **adaptivem
Lernen** fokussieren auf die
Analyse großer Mengen an
Bildungsdaten (Big Data)
sowie den Einsatz KI-
basierter Technologien zur
Anpassung des
Lernangebots **an
individuelle Lernprozesse.**

Hemmler, Y., & Ifenthaler, D. (2022). Personalisierte und adaptive Lernumgebungen für Online-Weiterbildungen. In S. Schumann, S. Seeber, & S. Abele (Eds.), *Digitale Transformation in der Berufsbildung. Konzept, Befunde und Herausforderungen* (pp. 145-164). wbv. <https://doi.org/10.3278/9783763971381>



San Pedro, M. O. Z., & Baker, R. S. (2021). Knowledge inference models used in adaptive learning. In A. A. Davier, R. J. Mislevy, & J. Hao (Eds.), *Computational psychometrics: new methodologies for a new generation of digital learning and assessment. methodology of educational measurement and assessment* (pp. 61–77). Springer. https://doi.org/10.1007/978-3-030-74394-9_5m

Ein vielversprechender Ansatz besteht darin, **Lernende die Anpassung vornehmen zu lassen** und die Änderungen der Lernenden zu berücksichtigen, um den adaptiven Algorithmus zu aktualisieren.

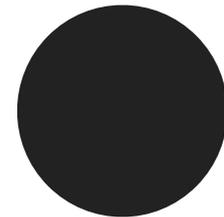


Table 8 Research methodological components (outcome variables, research design and data collection methods) in adaptive learning studies

Research methodological components	Number of studies	Percentage
<i>Outcome variables</i>		
Cognitive (e.g., achievement)	41	67.2
Affective (e.g., attitude)	23	37.7
Behavior (e.g., time spent)	25	41.0
Other	6	9.8
<i>Research design</i>		
Experimental	27	44.3
Non-experimental	11	18.0
Qualitative	6	9.8
Mixed-method	17	27.9
<i>Data collection methods</i>		
Test data	43	70.5
Survey	40	65.6
Extant data (Email, recording, discussion data)	25	41.0
Interview	9	14.7
Observation	5	8.2
Clickstream data/log files	4	6.6



Martin, F., Chen, Y., Moore, R. L., & Westine, C. D. (2020). Systematic review of adaptive learning research designs, context, strategies, and technologies from 2009 to 2018. *Educational Technology Research and Development*, 68, 1903–1929. <https://doi.org/10.1007/s11423-020-09793-2>

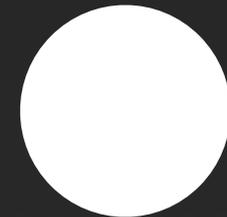
Table 11 Adaptive targets (content and instructional characteristics) in adaptive learning environments

What was adapted?	Adaptive target	Number of studies
Content	Adaptive content	9
Assessment	Adaptive feedback	8
Assessment	Adaptive course topic and question difficulty	4
Navigation	Adaptive learning sequence	5
Navigation	Adaptive learning path	5
Navigation	Adaptive pacing	1
Navigation	Adaptive navigation	1
Presentation	Adaptive caption filtering	1
Presentation	Adaptive material format and presentation	4

Martin, F., Chen, Y., Moore, R. L., & Westine, C. D. (2020). Systematic review of adaptive learning research designs, context, strategies, and technologies from 2009 to 2018. *Educational Technology Research and Development*, 68, 1903–1929. <https://doi.org/10.1007/s11423-020-09793-2>



Valide Indikatoren zur Unterstützung von
Algorithmen sind die **Voraussetzung für**
adaptive Lernumgebungen in der
Hochschulbildung.

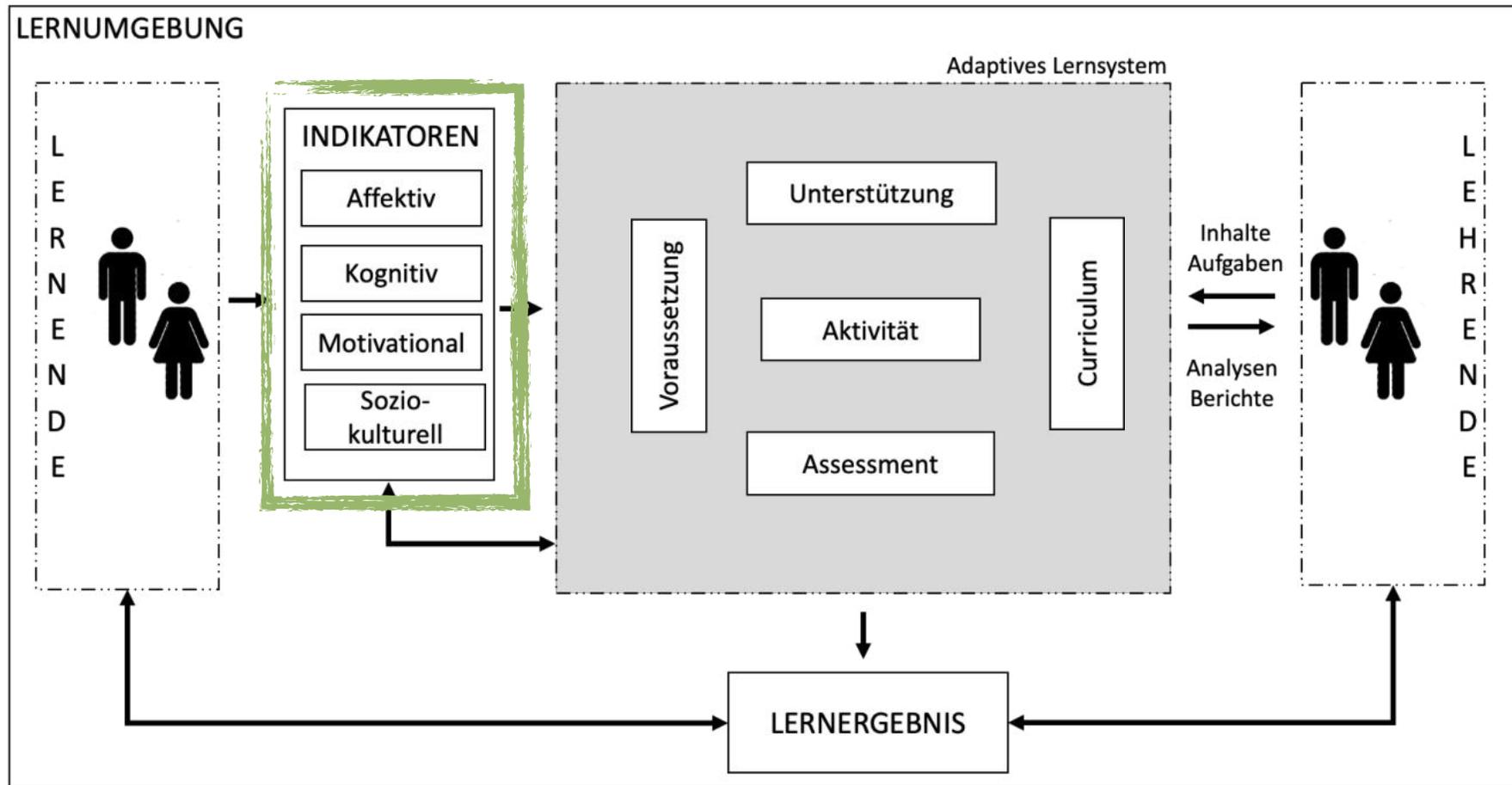


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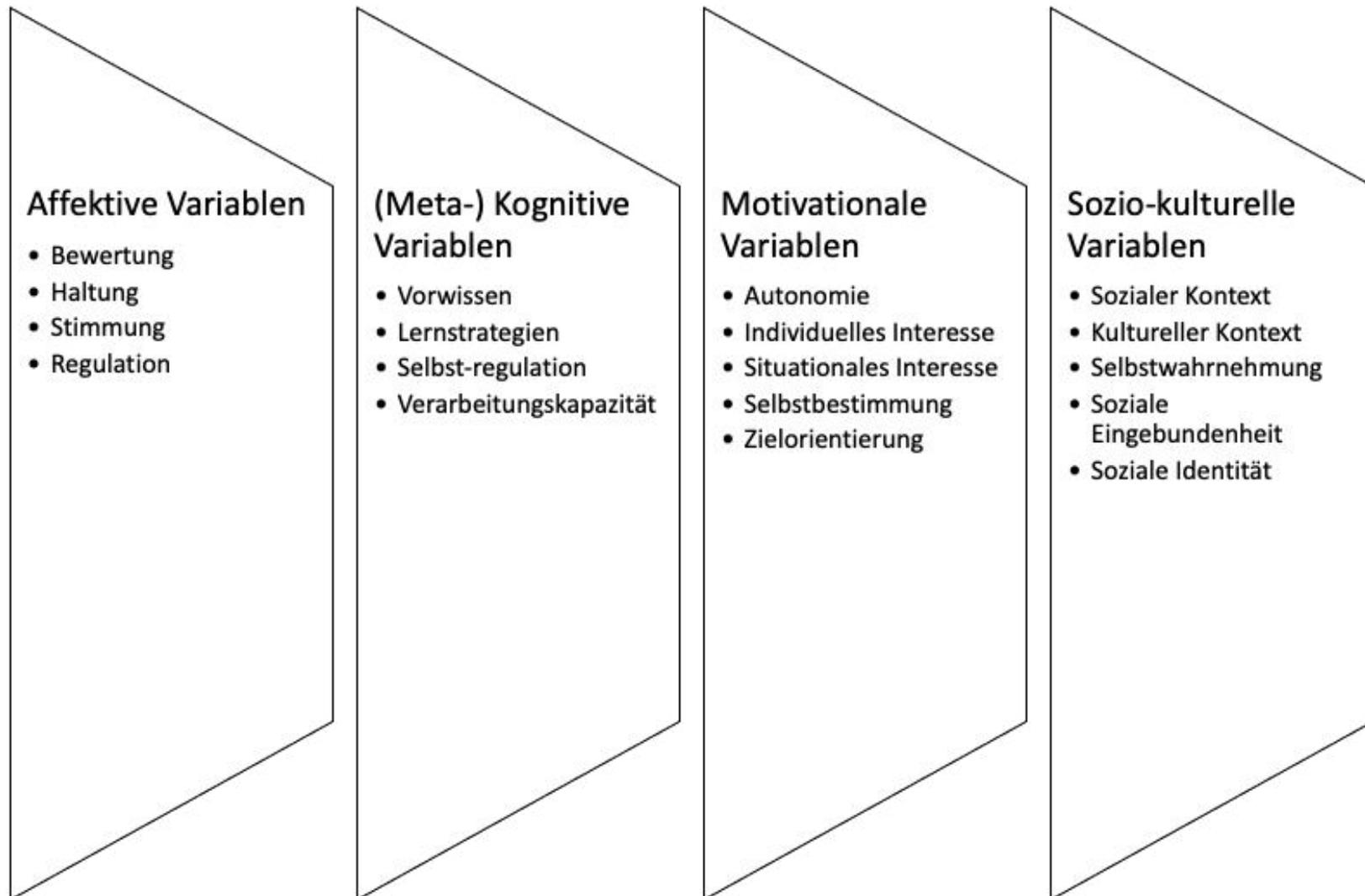


Yau, J., & Ifenthaler, D. (2020). Reflections on different learning analytics indicators for supporting study success. *International Journal of Learning Analytics and Artificial Intelligence for Education*, 2(2), 4–23. <https://doi.org/10.3991/ijai.v2i2.15639>

Makro-Perspektive adaptiver Lernumgebungen

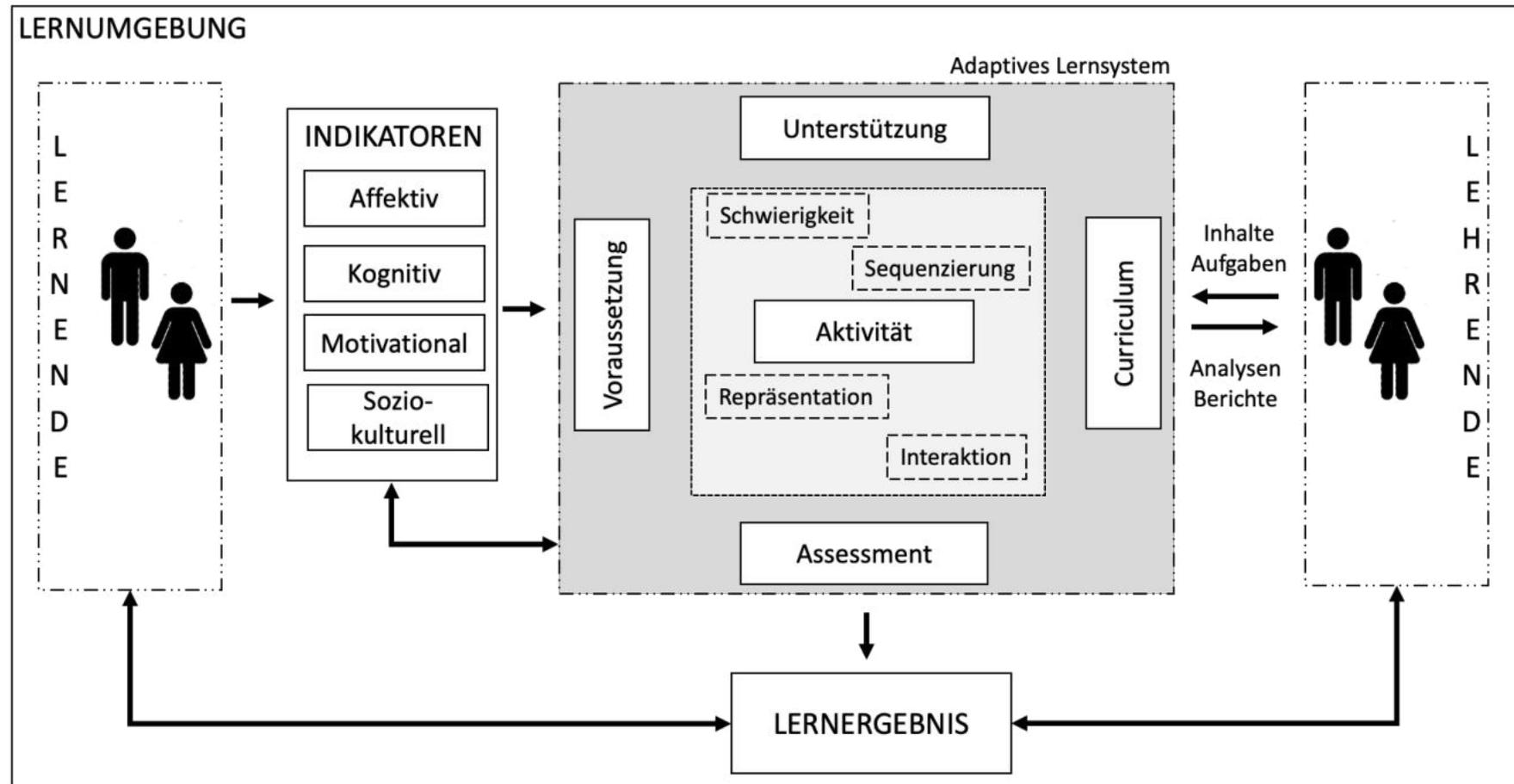


Hemmler, Y., & Ifenthaler, D. (2022). Four perspectives on personalized and adaptive learning environments for workplace learning. In D. Ifenthaler & S. Seufert (Eds.), *Artificial intelligence education in the context of work* (pp. 27–39). Springer. https://doi.org/10.1007/978-3-031-14489-9_2



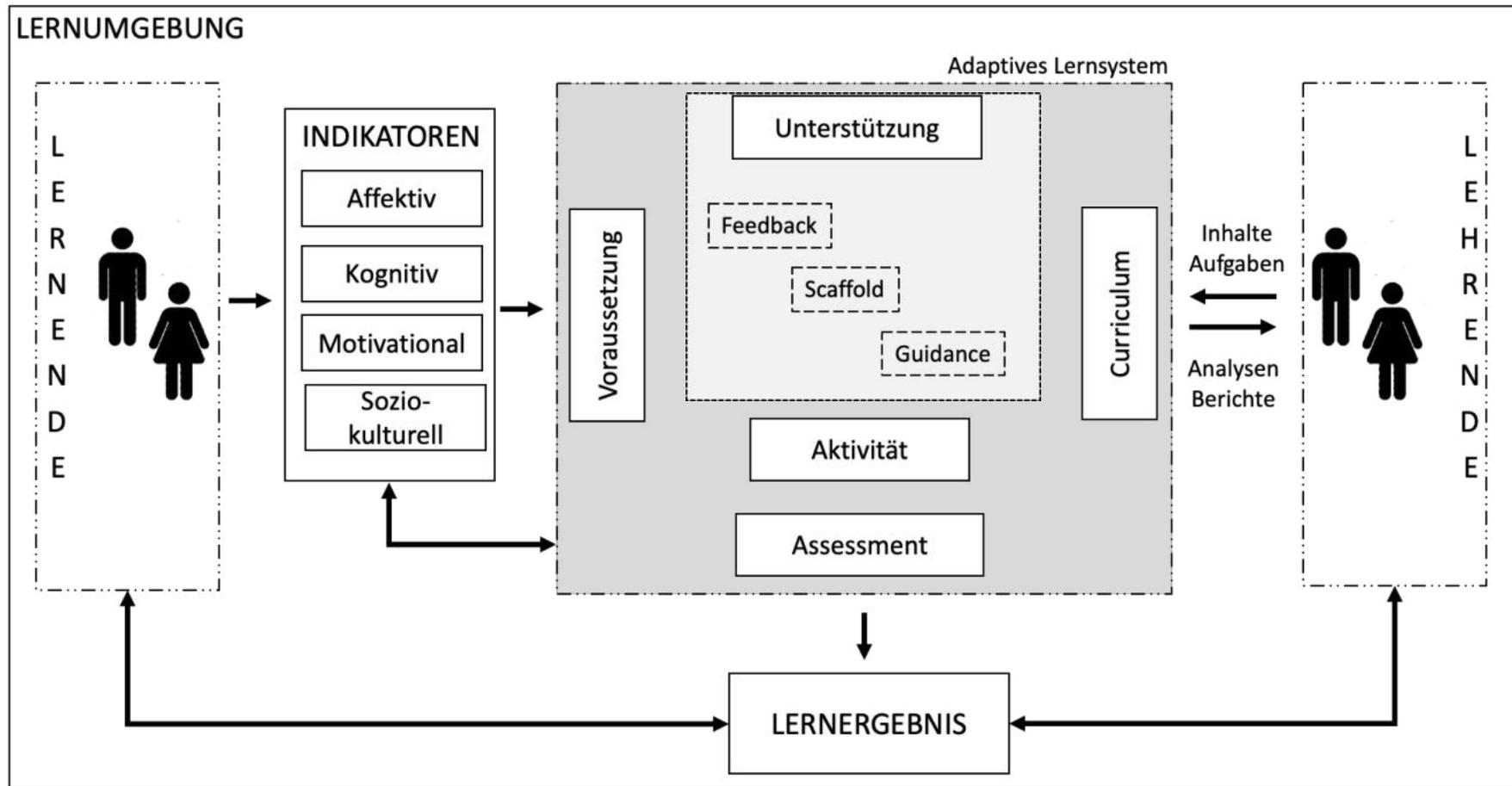
Plass, J. L., & Pawar, S. (2020). Toward a taxonomy of adaptivity for learning. *Journal of Research on Technology in Education*, 52(3), 275–300. <https://doi.org/10.1080/15391523.2020.1719943>

Mikro-Perspektive: **Aktivität**



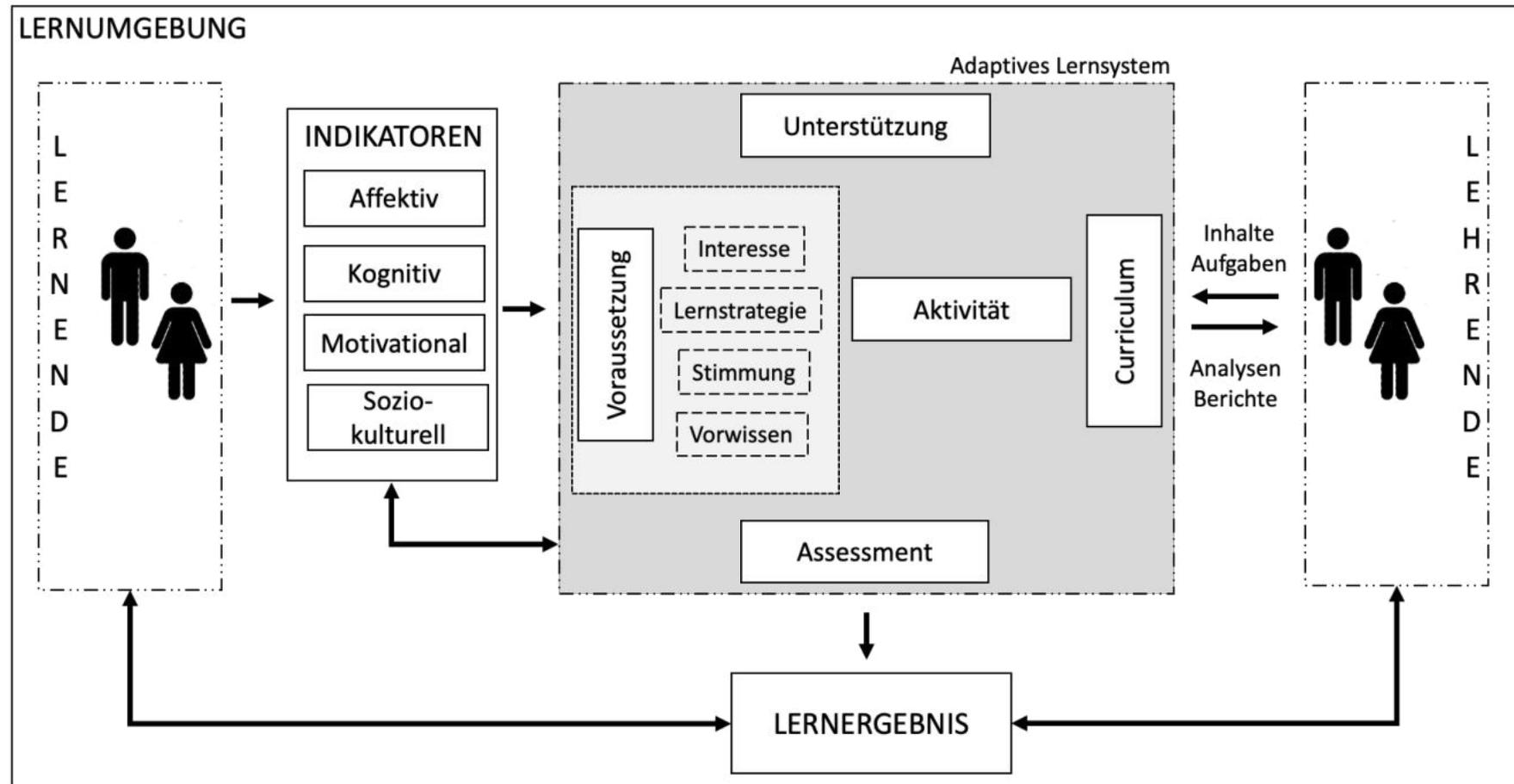
Hemmler, Y., & Ifenthaler, D. (2022). Four perspectives on personalized and adaptive learning environments for workplace learning. In D. Ifenthaler & S. Seufert (Eds.), *Artificial intelligence education in the context of work* (pp. 27–39). Springer. https://doi.org/10.1007/978-3-031-14489-9_2

Mikro-Perspektive: Unterstützung



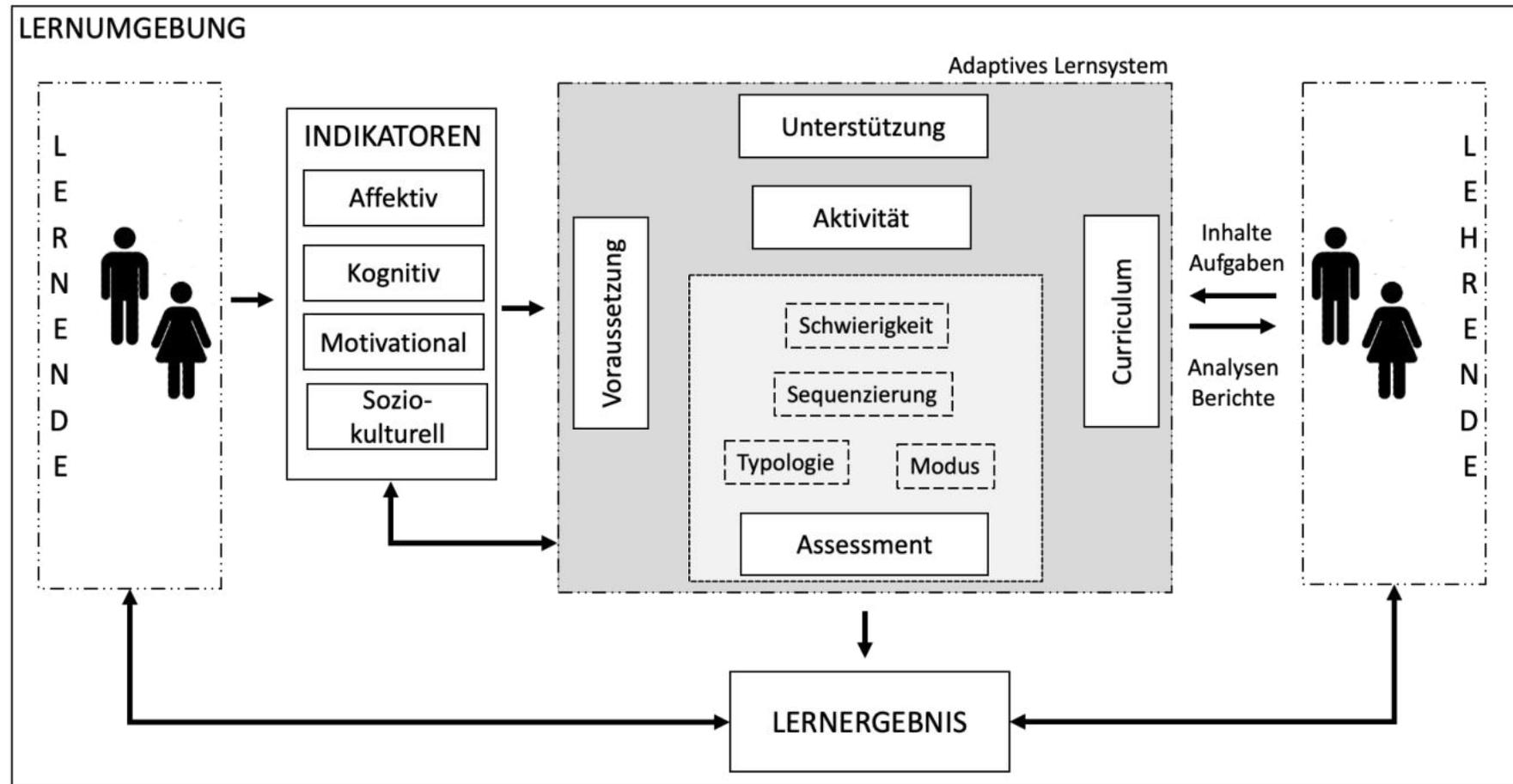
Hemmler, Y., & Ifenthaler, D. (2022). Four perspectives on personalized and adaptive learning environments for workplace learning. In D. Ifenthaler & S. Seufert (Eds.), *Artificial intelligence education in the context of work* (pp. 27–39). Springer. https://doi.org/10.1007/978-3-031-14489-9_2

Mikro-Perspektive: Voraussetzung



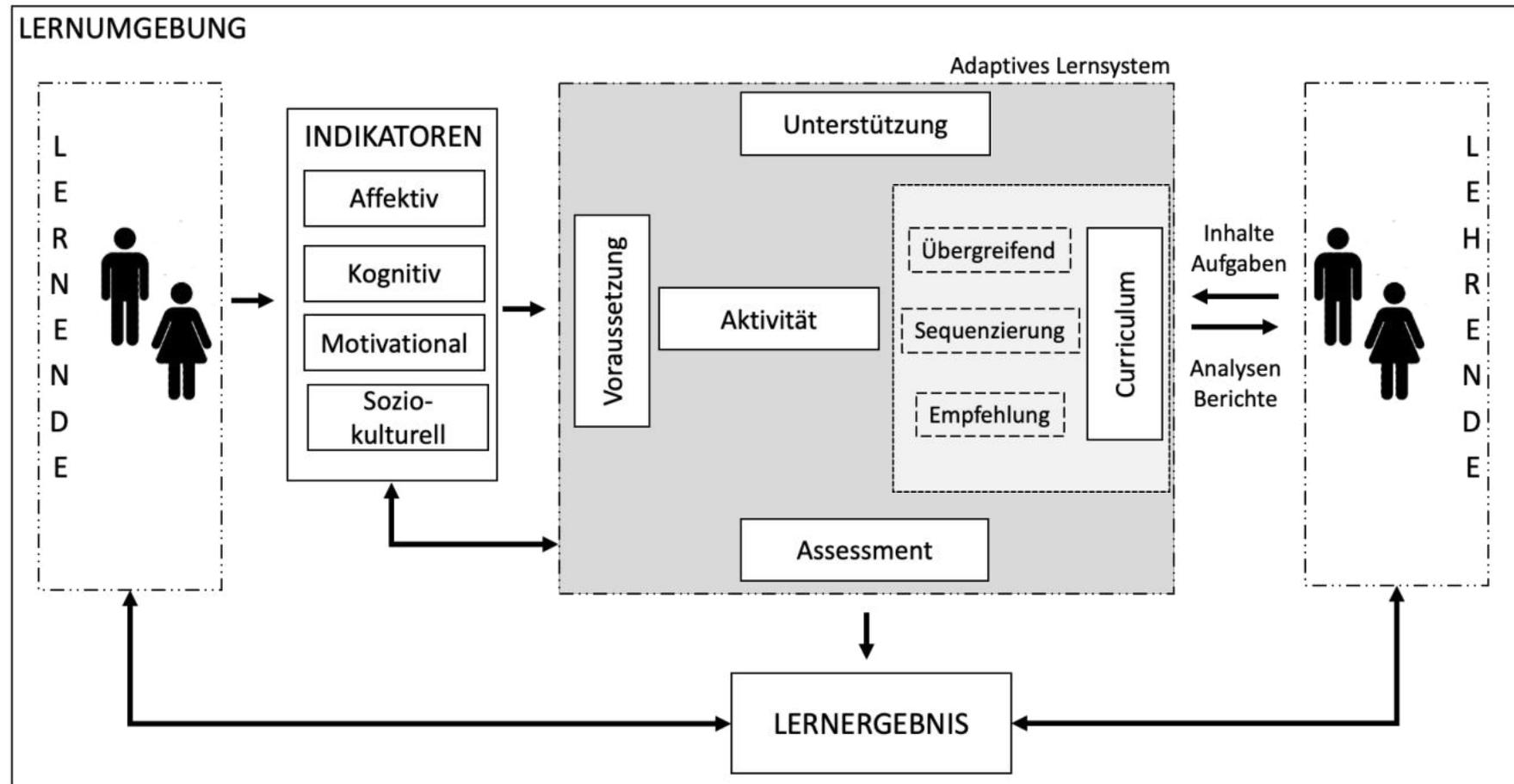
Hemmler, Y., & Ifenthaler, D. (2022). Four perspectives on personalized and adaptive learning environments for workplace learning. In D. Ifenthaler & S. Seufert (Eds.), *Artificial intelligence education in the context of work* (pp. 27–39). Springer. https://doi.org/10.1007/978-3-031-14489-9_2

Mikro-Perspektive: Assessment



Hemmler, Y., & Ifenthaler, D. (2022). Four perspectives on personalized and adaptive learning environments for workplace learning. In D. Ifenthaler & S. Seufert (Eds.), *Artificial intelligence education in the context of work* (pp. 27–39). Springer. https://doi.org/10.1007/978-3-031-14489-9_2

Mikro-Perspektive: Curriculum



Hemmler, Y., & Ifenthaler, D. (2022). Four perspectives on personalized and adaptive learning environments for workplace learning. In D. Ifenthaler & S. Seufert (Eds.), *Artificial intelligence education in the context of work* (pp. 27–39). Springer. https://doi.org/10.1007/978-3-031-14489-9_2

ADAPTIVES SYSTEM

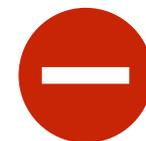
ADAPTIERBARKEIT

ADAPTIVE
UNTERSTÜTZUNG

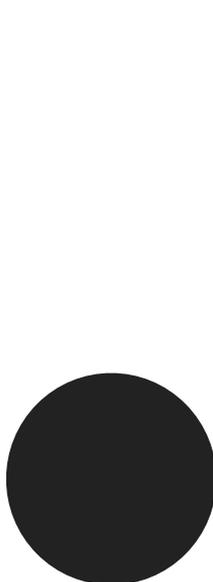
AUTOMATISIERTE
ADAPTION



Selbstregulation



Hemmler, Y., & Ifenthaler, D. (2022). Personalisierte und adaptive Lernumgebungen für Online-Weiterbildungen. In S. Schumann, S. Seeber, & S. Abele (Eds.), *Digitale Transformation in der Berufsbildung. Konzept, Befunde und Herausforderungen* (pp. 145-164). wbv. <https://doi.org/10.3278/9783763971381>

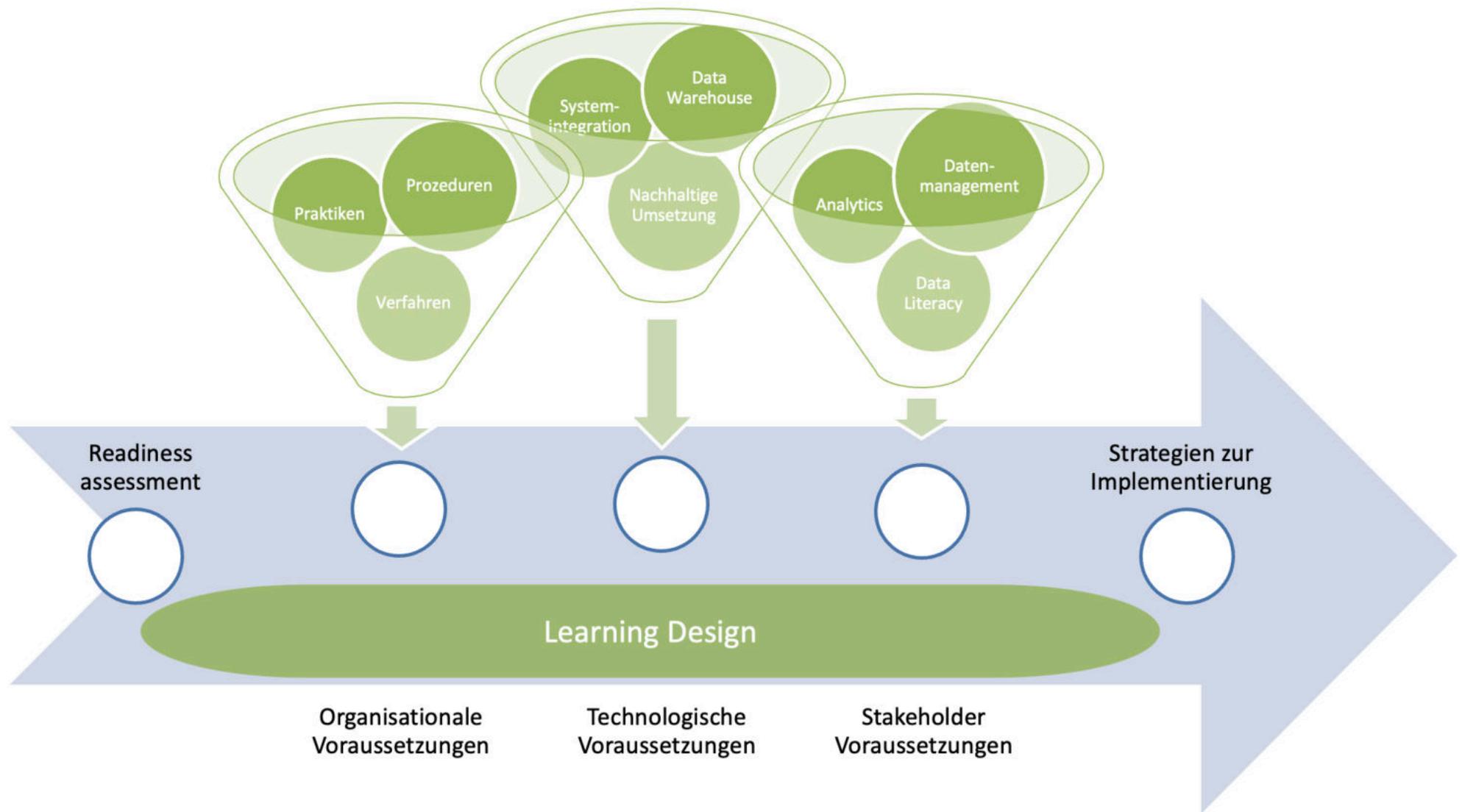


Die Implementierung von adaptiven
Lernumgebungen erfordert eine
Weiterentwicklung von **Systemen**,
Prozessen und **Stakeholdern**.

49



Ifenthaler, D. (2017). Are higher education institutions prepared for learning analytics? *TechTrends*, 61(4), 366–371.
<https://doi.org/10.1007/s11528-016-0154-0>



Ifenthaler, D. (2020). Change management for learning analytics. In N. Pinkwart & S. Liu (Eds.), *Artificial intelligence supported educational technologies* (pp. 261–272). Springer. https://doi.org/10.1007/978-3-030-41099-5_15

**Hochschulbildung
berechenbar
machen?**

**Nachhaltige digitale
Transformation der
Hochschulbildung?**

1

2

3

4

**Hochschulbildung
demokratischer
gestalten?**

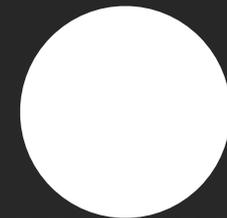
**Hochschulbildung
personalisierter
gestalten?**

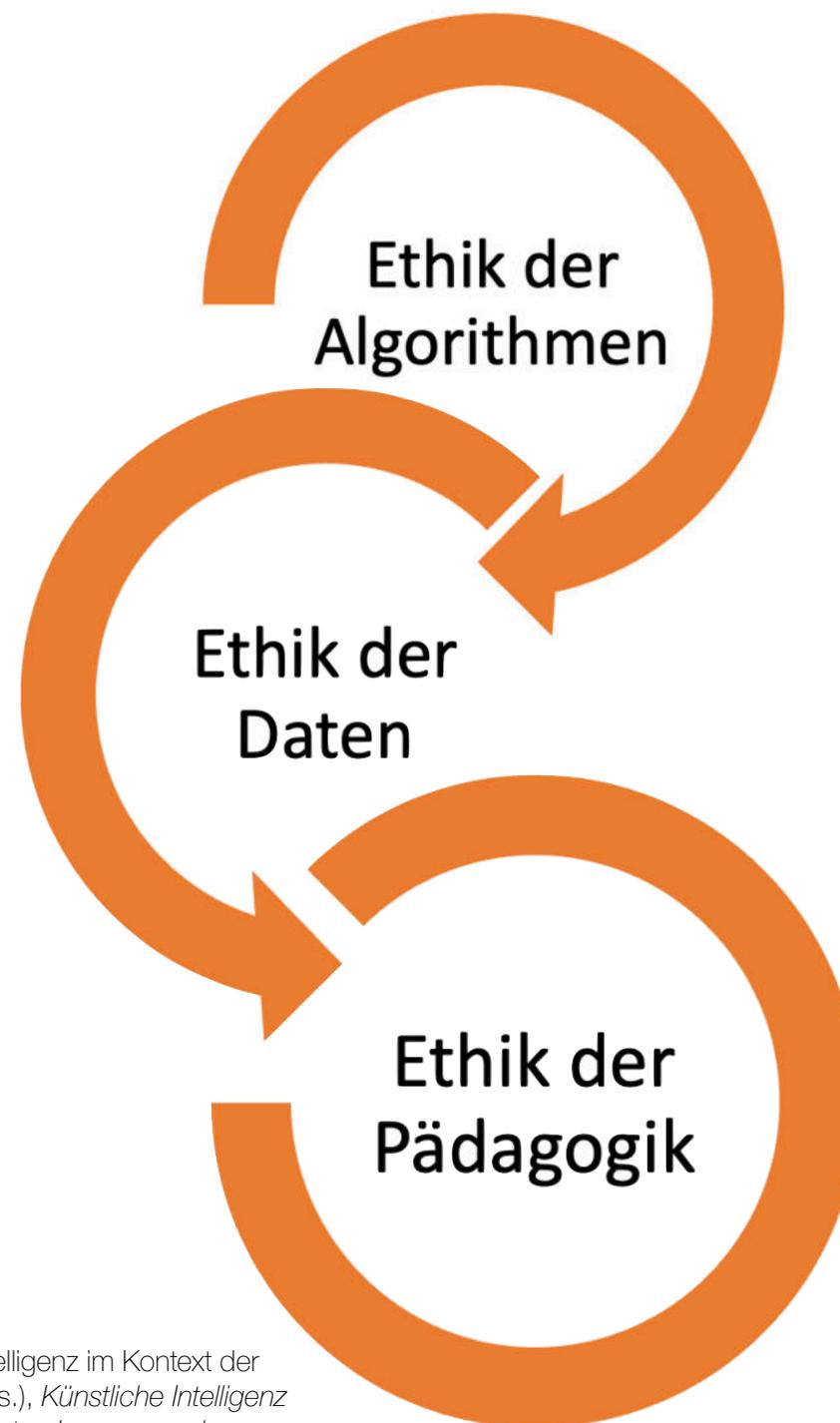


Digitale Technologien sind **nicht neutral**, sondern formen durch ihre bloße Existenz **Überzeugungen** und **Wünsche** sowie verändern die **kollektive Intentionalität**.



Searl, J. R. (1971). *The philosophy of language*. Oxford University Press.

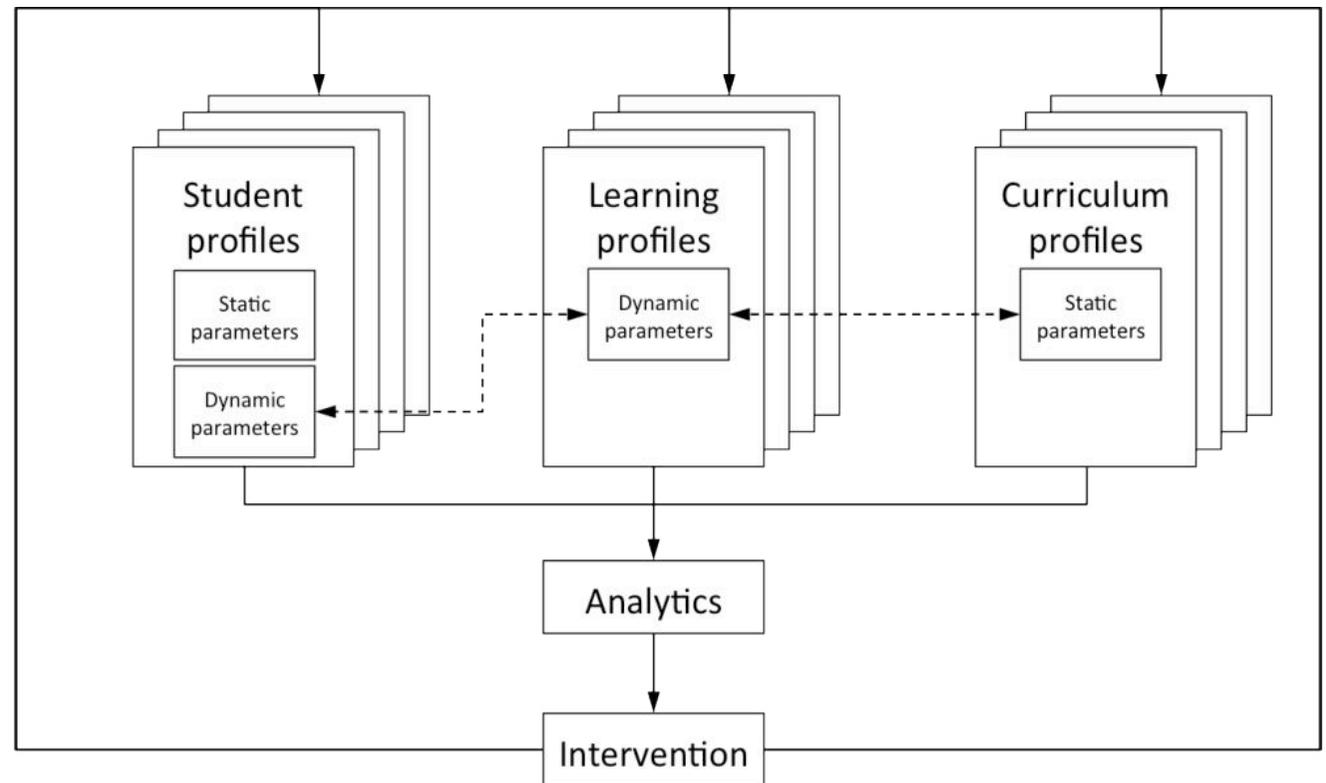


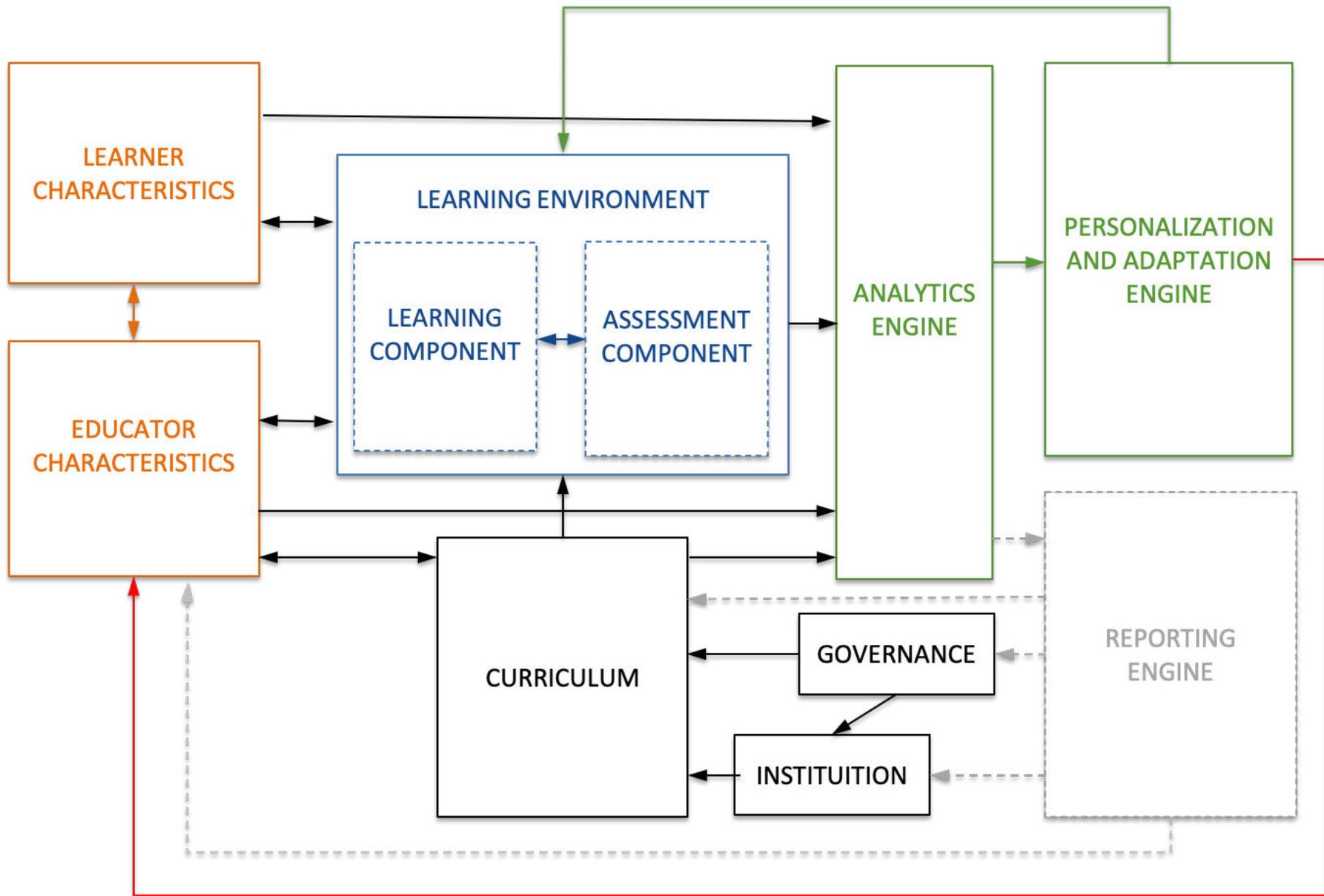


Ifenthaler, D. (2023). Ethische Perspektiven auf künstliche Intelligenz im Kontext der Hochschule. In T. Schmohl, A. Watanabe, & K. Schelling (Eds.), *Künstliche Intelligenz in der Hochschulbildung. Chancen und Grenzen des KI-gestützten Lernens und Lehrens* (pp. 71–86). transcript-Verlag. <https://doi.org/10.14361/9783839457696>

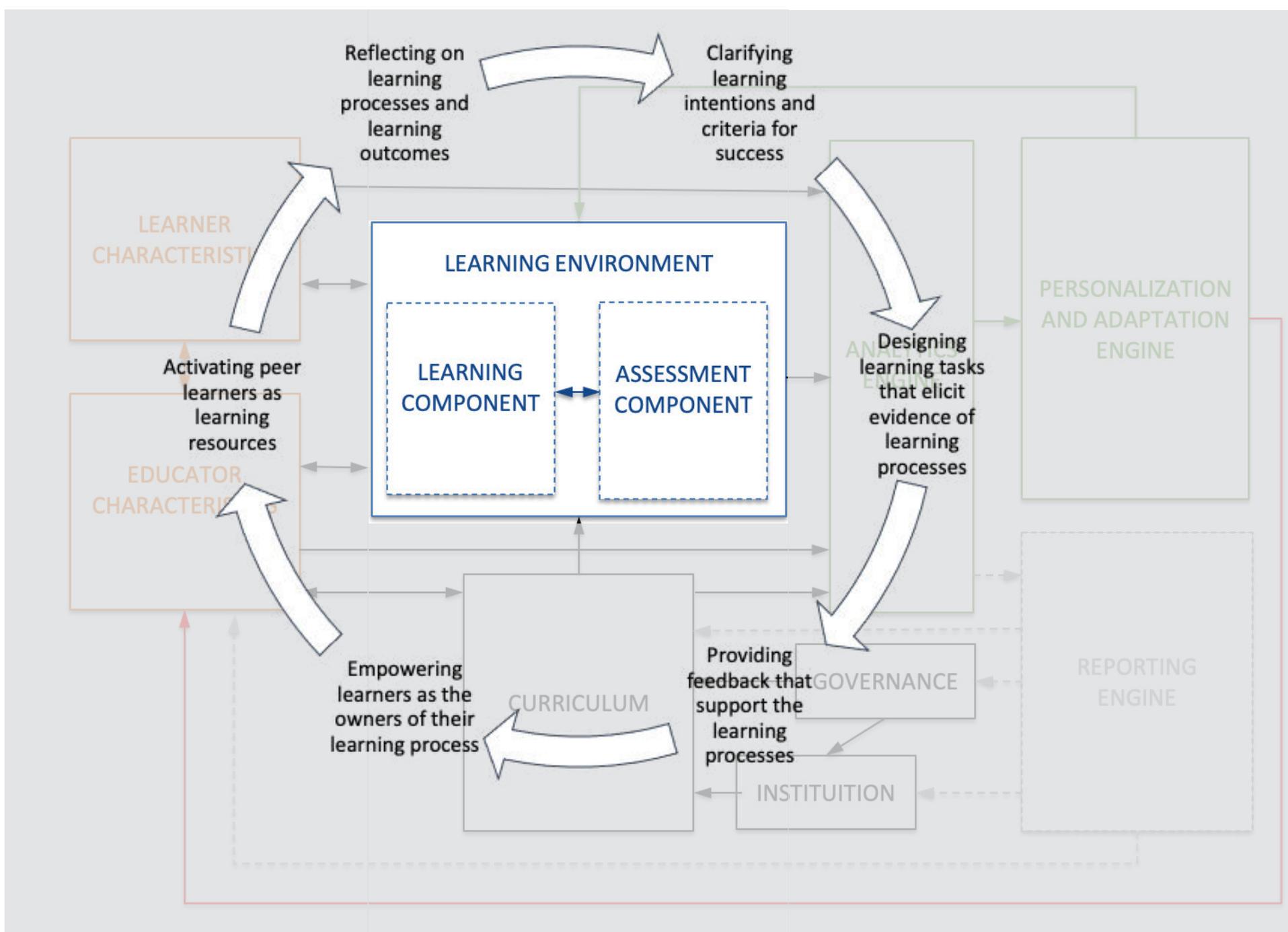
01

Daten- und Lern- Architektur.

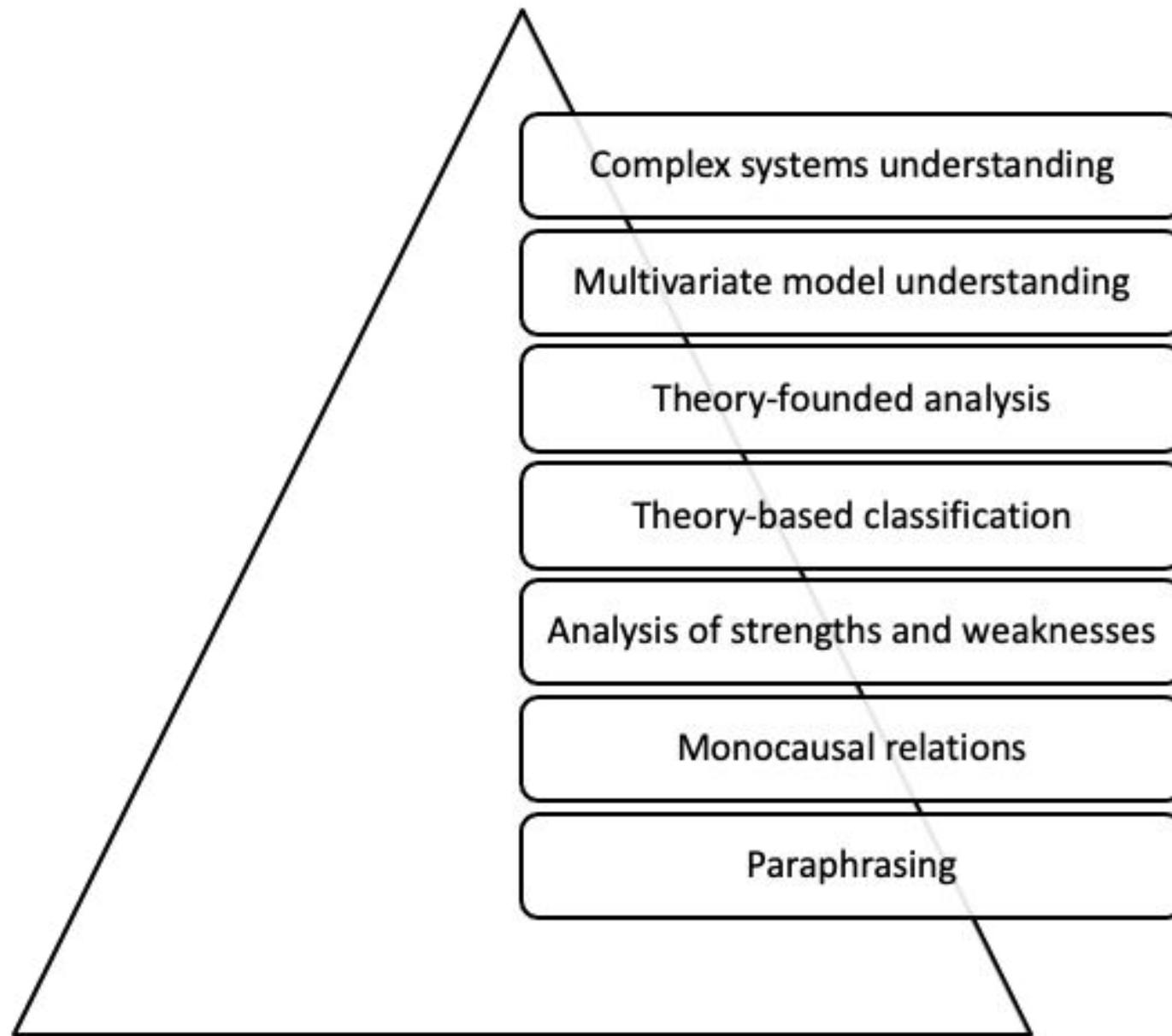




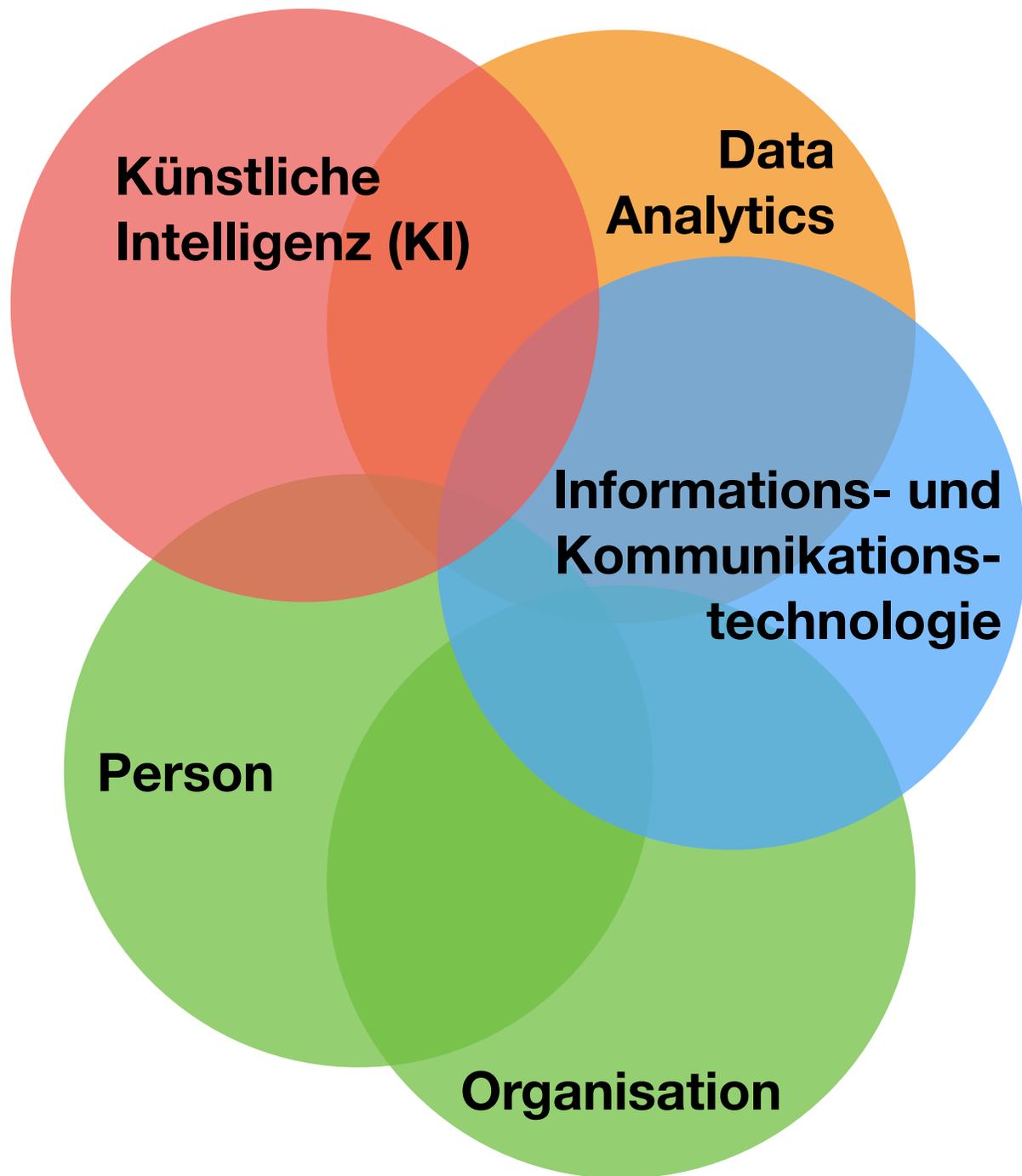
Ifenthaler, D., & Greiff, S. (2021). Leveraging learning analytics for assessment and feedback. In J. Liebowitz (Ed.), *Online learning analytics* (pp. 1–18). Auerbach Publications. <https://doi.org/10.1201/9781003194620>



Ifenthaler, D., & Greiff, S. (2021). Leveraging learning analytics for assessment and feedback. In J. Liebowitz (Ed.), *Online learning analytics* (pp. 1–18). Auerbach Publications. <https://doi.org/10.1201/9781003194620>



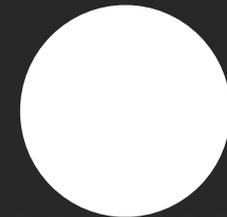
Pirnay-Dummer, P., Ifenthaler, D., & Seel, N. M. (2012). Designing model-based learning environments to support mental models for learning. In D. H. Jonassen & S. Land (Eds.), *Theoretical foundations of learning environments* (2 ed., pp. 66–94). Routledge.



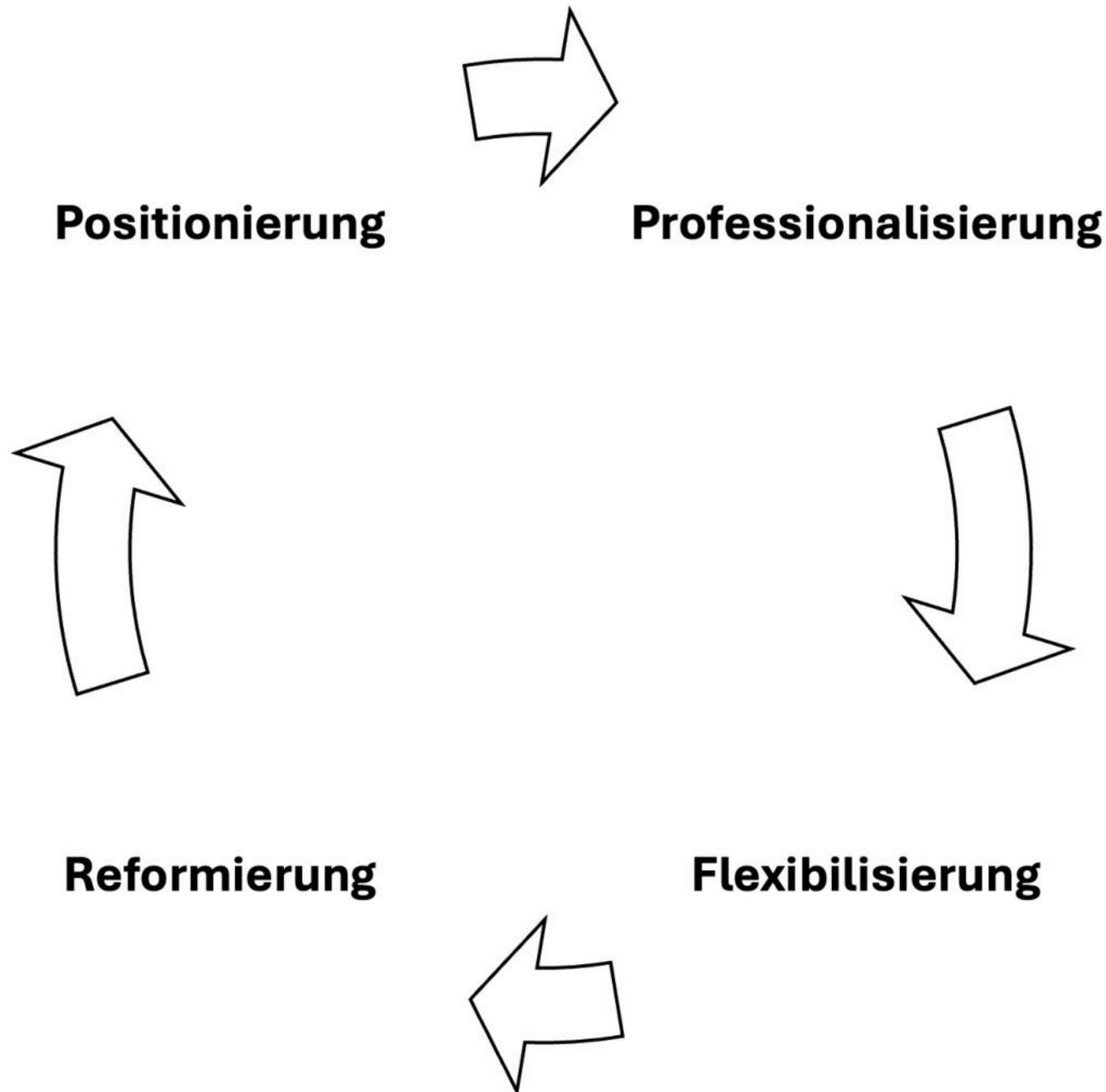
02

Nachhaltige
Lernkultur.

Die **Organisation will oft anderes vom Menschen** als der Mensch von der Organisation.

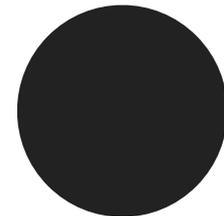


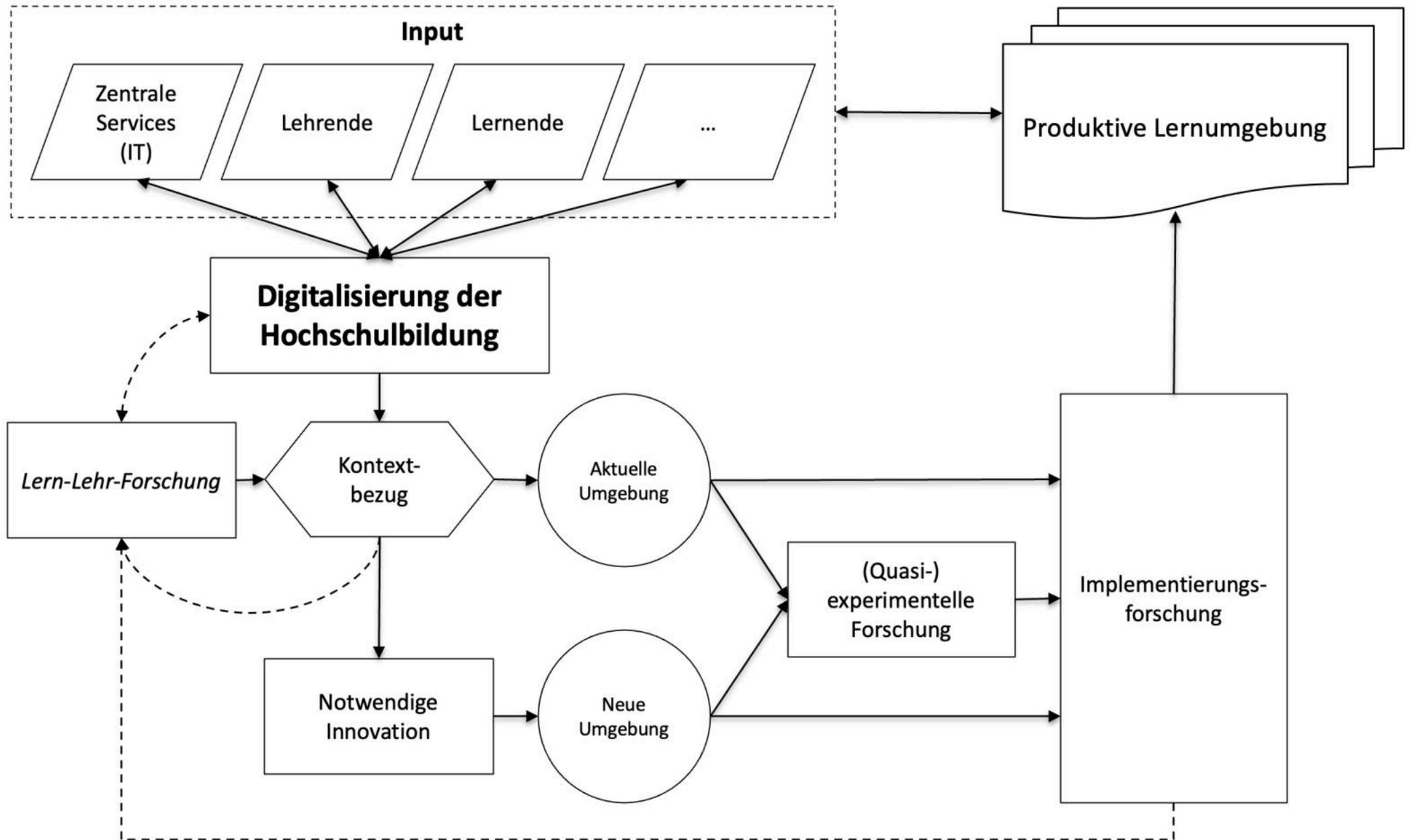
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Euler, D., & Seufert, S. (2011). Change Management in der Hochschullehre: Die nachhaltige Implementierung von e-Learning-Innovationen. Zeitschrift für Hochschulentwicklung. <https://doi.org/10.3217/zfhd03/02>

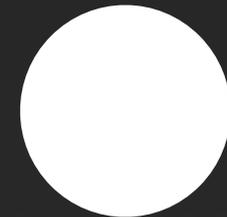
Strategien und **Rahmenwerke**, die eine **angemessene Nutzung** der Digitalisierung in Hochschulen gewährleisten, sind der Schlüssel für deren **nachhaltigen Einführung** und **Nutzung**.



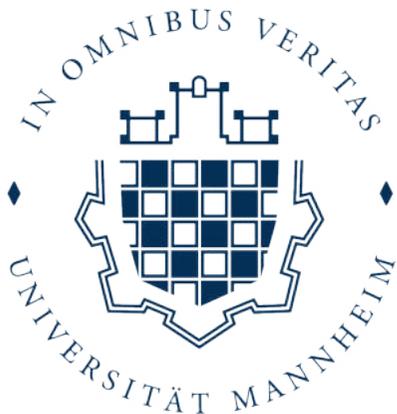


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Energieeffiziente Hard- und Software
sowie **gemeinsam genutzte**
Infrastrukturen, die auf **erneuerbarer**
Energie basieren sind entscheidend für
ihren **nachhaltigen Betrieb** und ihre
Skalierung im Kontext der
Hochschulbildung.



Digitalisierung als Treiber für Change Prozesse an Hochschulen



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