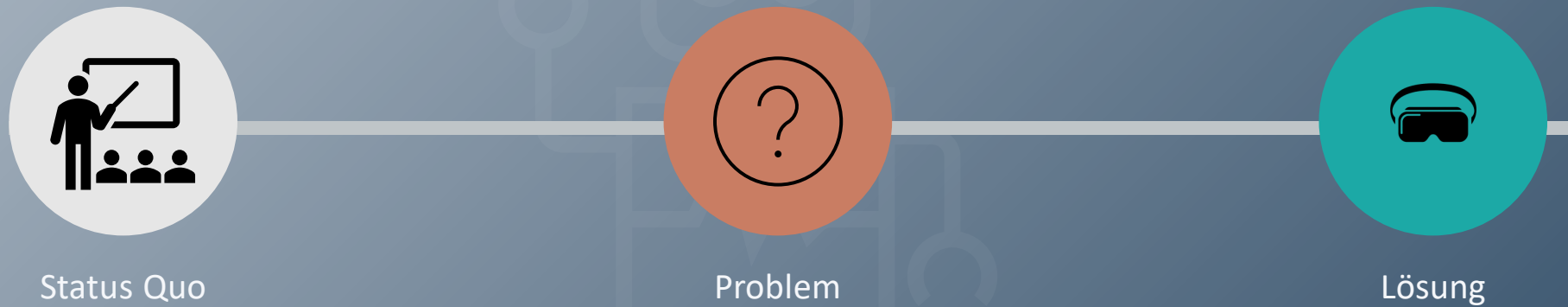


Virtuelle Identitäten, Reale Beratung

Zur transformativen Rolle von Avataren und Künstlicher Intelligenz
beim Beratungstraining in Virtual Reality

Ausgangslage

Ausgangslage



Ausgangslage



Ausgangslage: Theoretischer Überblick

Die Rolle von Avataren

- Steigendes Interesse an Einsatz von Avataren (Chen & Yang, 2013; Hsiao & Chen, 2016)
- Avatare in HCI (Taylor, 2002; Nakamura, 2007)
 - Selbstdiskrepanz-Theorie (Higgins, 1987; Jin, 2012), Proteus-Effekt (Yee & Bailenson, 2007; Nowak & Fox, 2018)
- Anpassbarkeit (Yee & Bailenson, 2009)
 - Psychologische Effekte der Anpassung (Waltemate et al., 2018; Gonzalez-Franco et al., 2024)
- Avatare in VR
 - Bewegung durch Full-Body Tracking (McVeigh-Schultz et al., 2018; Freeman & Maloney, 2021)
 - Zusammenhang zwischen Körperpräsenz und virtueller Identität (Vasalou & Joinson, 2009; Birk et al., 2016)
 - Designfaktoren (Wu et al., 2023; Pauw et al., 2022)



Ausgangslage: Theoretischer Überblick



Debriefing

- Schlüsselmechanismus für das Lernen (Thatcher, 1990; Kriz & Saam, 2007)
- Zwei unterschiedliche Debriefing-Arten (e.g. Dufrene & Young, 2014; Luctkar-Flude et al., 2021; Favolise et al., 2024)
 - Moderiertes Debriefing (The Inascl Board Of Directors, 2011)
 - Selbstdebriefing
- Studien zeigen: Debriefing wirkt (Boet et al., 2011; Tilton, 2013; Tilton et al., 2015; Verkuyl et al., 2018)
- Spannungsfeld: Moderiertes Debriefing & akademisierter Hochschullehre

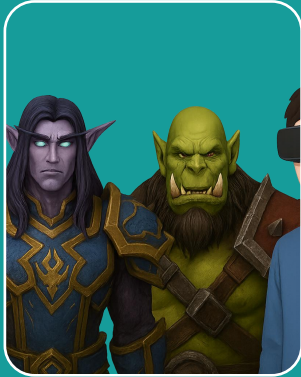
Ausgangslage: Theoretischer Überblick

Chatbots als Debriefler

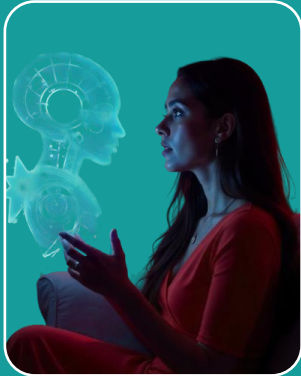
- Zunehmender Einsatz in Bildungskontexten (Hong, 2023; Chan & Hu, 2023)
- Können als Reflexionsmittel eingesetzt werden
 - Problem: Komplexität von Gesprächssituationen (Jin et al., 2024) Sycophancy (Sun & Wang, 2025; Malmqvist, 2024)
- Selbstverantwortung des Lernenden (Järvelä et al., 2023; Molenaar, 2022)
- Probleme beim „Prompten“ → Chatbot = Informationsquelle (Zamfirescu-Pereira et al., 2023; Klar, 2024)
- Chatbot als Debriefler hält seine Rolle
 - Lernende reflektieren Verhalten (Evangelou et al., 2025a)



Ausgangslage: Forschungsfragen

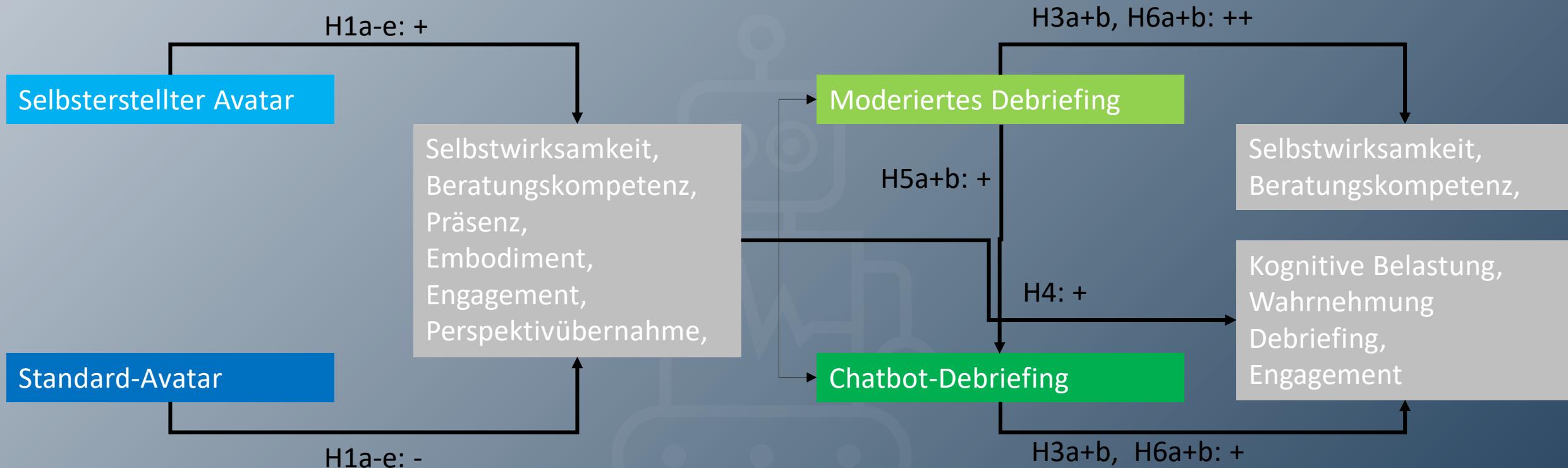


Hat das Erstellen eines eigenen Avatars als digitale Repräsentation eines Selbst positiven Einfluss auf Lerneffekte?



Inwieweit kann künstliche Intelligenz als adäquates Mittel für die Durchführung von Post-Simulations-Debriefings genutzt werden?

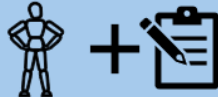
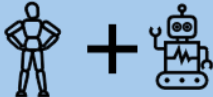
Hypothesen



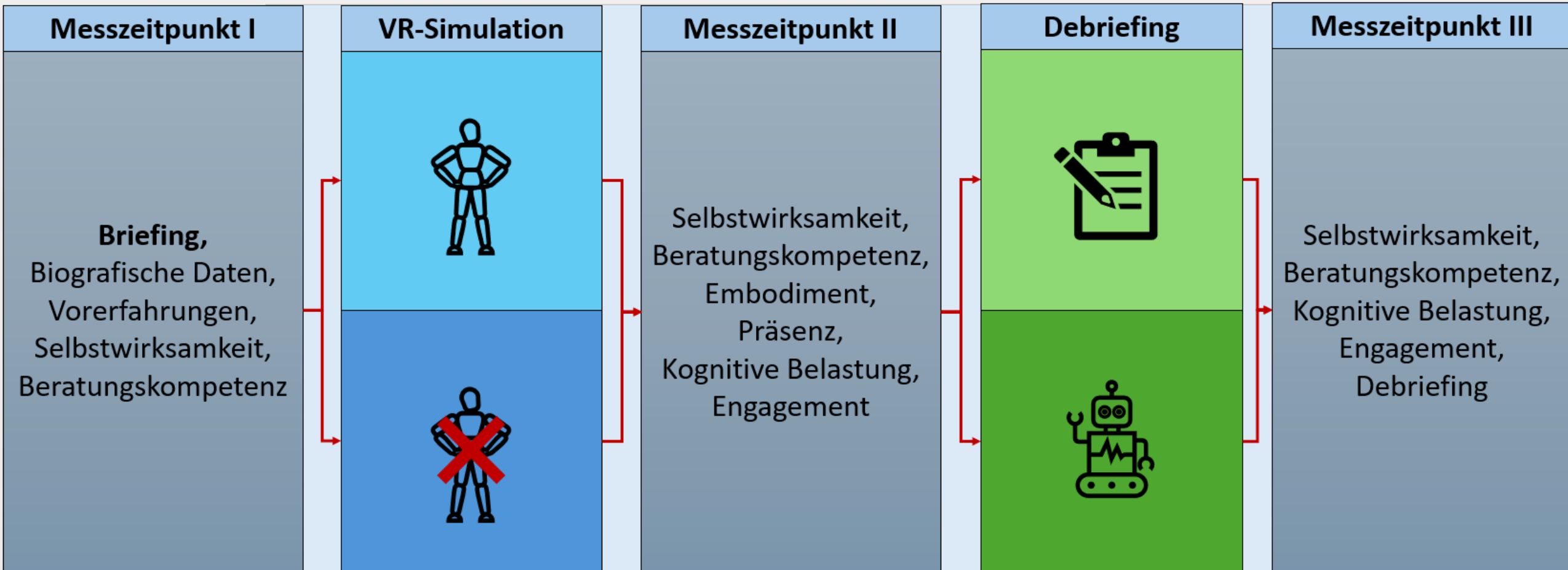


Forschungsdesign: Pilotstudie

Forschungsdesign: Pilotstudie

Eigener Avatar + Angeleitetes, strukturiertes Debriefing 	Eigener Avatar + KI-geleitetes Debriefing 
Kein eigener Avatar + Angeleitetes, strukturiertes Debriefing 	Kein eigener Avatar + KI-geleitetes Debriefing 

Forschungsdesign: Pilotstudie





Datenanalyse: Pilotstudie

Datenanalyse: Pilotstudie



45 Studierende



-  +  - 11
-  +  - 12
-  +  - 12
-  +  - 10

age $M = 22$; $SD = 8.7$

39 weiblich, 6
männlich

Ergebnisse

Datenanalyse: Avatar

	Min	Md	Max	M	SD	α
Self-efficacy (t1)	2.38	4.1	5.75	4.1	0.82	.89
Self-efficacy (t2)	1.5	4.13	5.75	4.2	0.91	.92
Counseling competence (t1)	3.86	5.4	7.29	5.4	0.83	.94
Counseling competence (t2)	2.23	5.59	7.52	5.6	1.14	.96
Embodiment	1.0	3.83	7.0	3.83	1.32	.93
Perspective-Taking	2.17	5.5	7.0	5.34	1.06	.9
Engagement	2.0	4.25	5.0	4.12	0.86	.92
Presence	1.13	3.87	5.8	3.85	1.04	.94

$F(1,43) = 0.26, p = .615 \rightarrow$ Selbstwirksamkeit

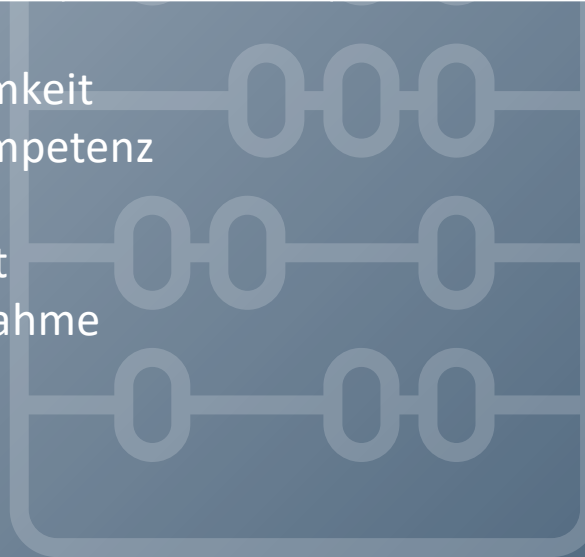
$F(1,43) = 0.32, p = .574 \rightarrow$ Beratungskompetenz

$t(42.99) = 0.72, p = .476 \rightarrow$ Präsenz

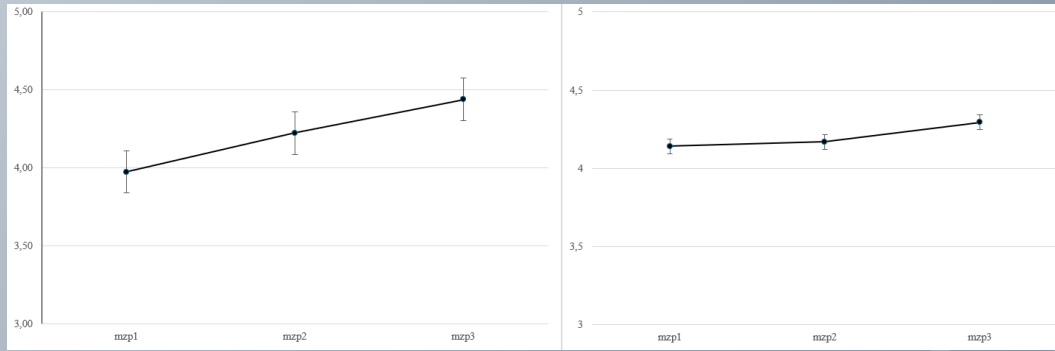
$t(39.52) = 0.26, p = .796 \rightarrow$ Embodiment

$W = 301.5, p = .274 \rightarrow$ Perspektivübernahme

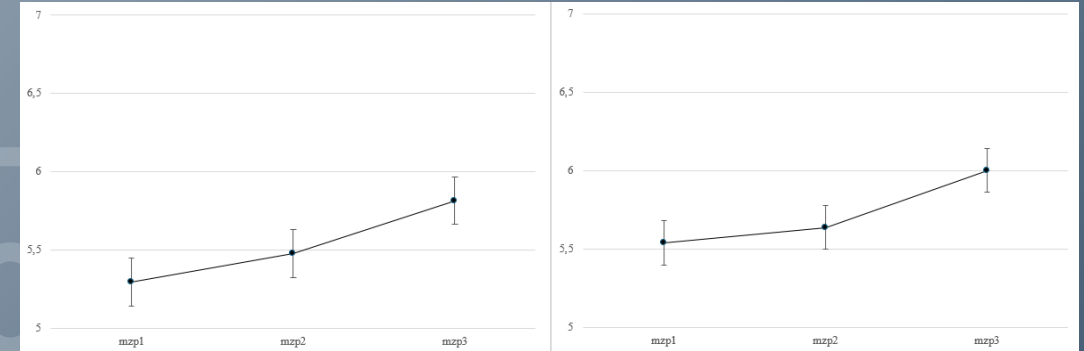
$W = 268.5, p = .728 \rightarrow$ Engagement



Datenanalyse: Debriefing



Average increase in self-efficacy for moderated debriefing (left) and self-debriefing (right). While the original Likert scale ranged from 1 to 6, only the relevant range from 3 to 5 is displayed here (mzp = measurement point).



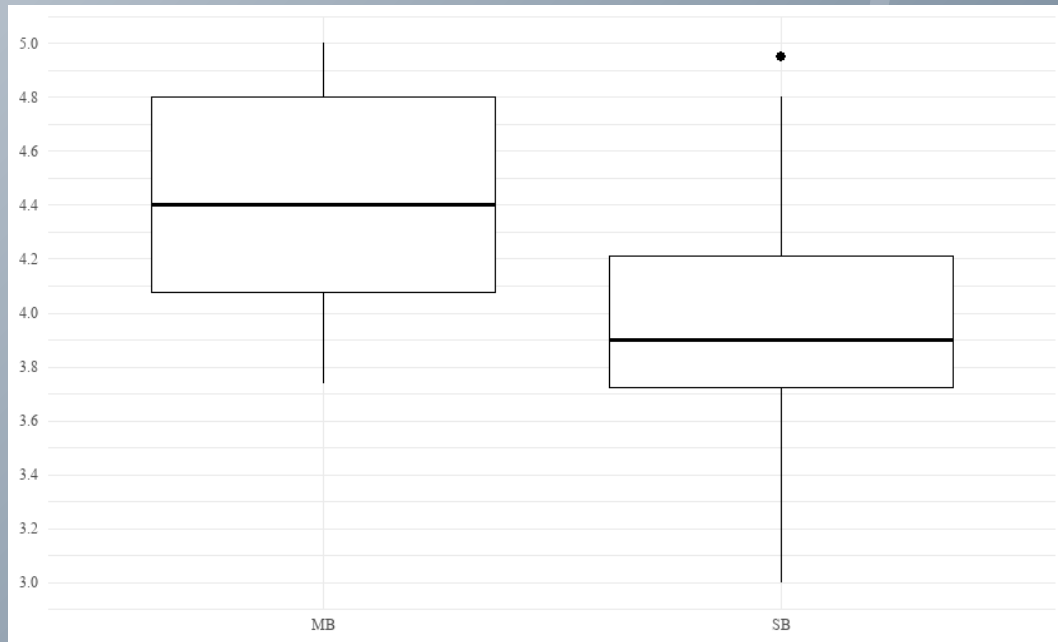
Average increase in counseling competence for moderated debriefing (left) and self-debriefing (right). While the original Likert scale ranged from 1 to 8, only the relevant range from 5 to 7 is displayed here (mzp = measurement point).

$t(44) = 3.47, p < .001 \rightarrow \text{MZP 1 zu MZP 3}$
 $t(44) = 3.94, p < .0001 \rightarrow \text{MZP 2 zu MZP 3}$

$t(44) = 4.27, p < .001 \rightarrow \text{MZP 1 zu MZP 3}$
 $t(44) = 5.56, p < .001 \rightarrow \text{MZP 2 zu MZP 3}$

$F(1,43) = 0.001, p = .971 \rightarrow \text{Selbstwirksamkeit}$
 $F(1,43) = 0.46, p = .500 \rightarrow \text{Beratungskompetenz}$

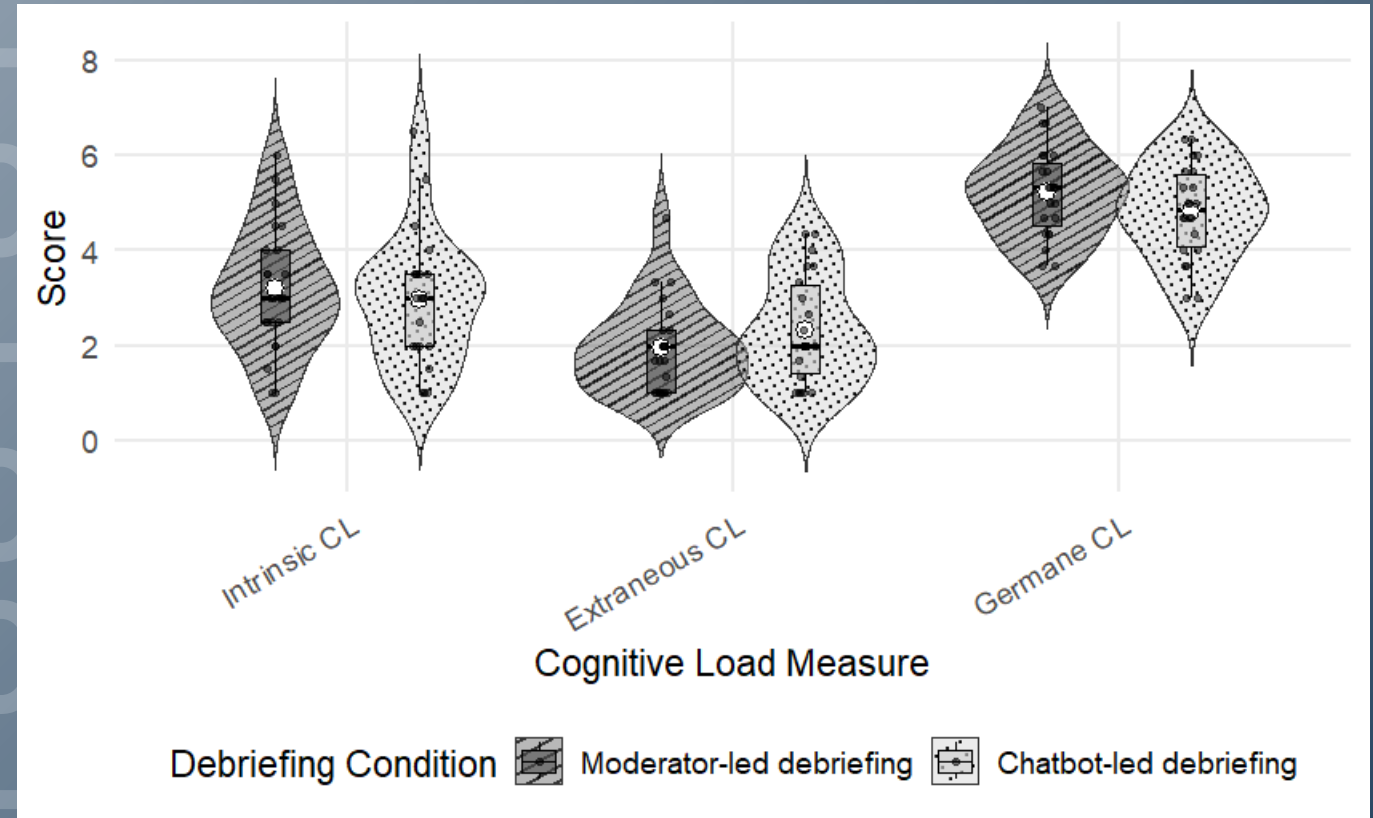
Datenanalyse: Debriefing



$t(39.806) = 3.68, p = .0007$

Datenanalyse: Cognitive Load

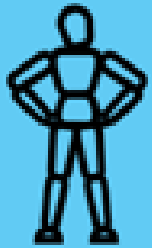
$t(43) = 0.59, p = .557 \rightarrow \text{ICL}$
 $U = 204, p = .267 \rightarrow \text{ECL}$
 $t(43) = 1.40, p = .169 \rightarrow \text{GCL}$





Fazit

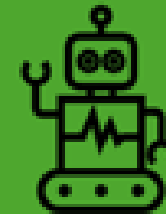
Fazit



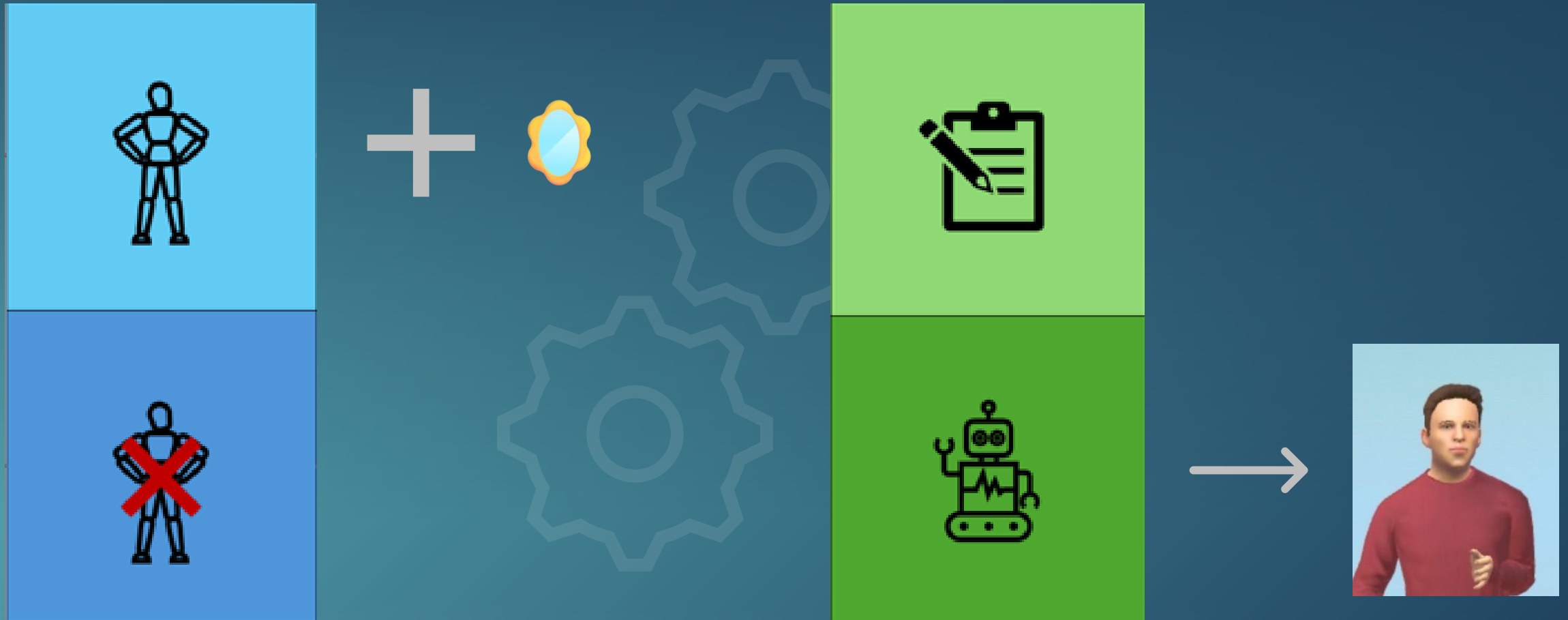
Relevanz eines Avatars in *high fidelity Simulation*



KI als adäquates Mittel



Fazit



Literaturverzeichnis

- [Bo11] Boet, Sylvain; Bould, M. Dylan; Bruppacher, Heinz R.; Desjardins, François; Chandra, Deven B.; Naik, Viren N.: Looking in the mirror: Self-debriefing versus instructor debriefing for simulated crises*. Critical Care Medicine, 39(6):1377–1381, June 2011.
- [J] Chan, Cecilia Ka Yuk; Hu, Wenjie: Students' voices on generative AI: perceptions, benefits, and challenges in higher education. International Journal of Educational Technology in Higher Education, 20(1):43, July 2023.
- [Co68] Cohen, Jacob: Weighted kappa: Nominal scale agreement provision for scaled disagreement or partial credit. Psychological Bulletin, 70(4):213–220, 1968. Place: US Publisher: American Psychological Association.
- [DY14] Dufrene, Claudine; Young, Anne: Successful debriefing — Best methods to achieve positive learning outcomes: A literature review. Nurse Education Today, 34(3):372–376, March 2014.
- [He09] Hertel, Silke: Beratungskompetenz von Lehrern. Waxmann Verlag, 2009.
- [Ho23] Hong, Wilson Cheong Hin: The impact of ChatGPT on foreign language teaching and learning: opportunities in education and research. Journal of Educational Technology and Innovation, 5(1):37–45, March 2023. Number: 1.
- [Ji24] Jin, Hyoungwook; Lee, Seonghee; Shin, Hyungyu; Kim, Juho: Teach AI How to Code: Using Large Language Models as Teachable Agents for Programming Education. In: Proceedings of the 2024 CHI Conference on Human Factors in Computing Systems. CHI '24, Association for Computing Machinery, New York, NY, USA, pp. 1–28, May 2024. Number: 652.
- [JNH23] Järvelä, Sanna; Nguyen, Andy; Hadwin, Allyson: Human and artificial intelligence collaboration for socially shared regulation in learning. British Journal of Educational Technology, 54(5):1057–1076, 2023. _eprint: <https://onlinelibrary.wiley.com/doi/pdf/10.1111/bjet.13325>.
- [KI24] Klar, Maria: How should we teach chatbot interaction to students? A pilot study on perceived affordances and chatbot interaction patterns in an authentic K-12 setting. Gesellschaft für Informatik e.V., p. 10.18420/delfi2024, 2024.
- [KR20] Kuckartz, Udo; Rädiker, Stefan: Fokussierte Interviewanalyse mit MAXQDA. Schritt Schritt, pp. 55–74, 2020.
- [Ku19] Kuckartz, Udo: Qualitative Content Analysis: From Kracauer's Beginnings to Today's Challenges. Forum Qualitative Sozialforschung / Forum: Qualitative Social Research, 20(3):20, 2019.
- [KS07] Kriz, Willy Ch; Saam, Nicole J: Großgruppenplanspiele als Interventionsmethode. Planspiele für die Organisationsentwicklung. Schriftenreihe: Wandel und Kontinuität in Organisationen, 2007.
- [Lu21] Luctkar-Flude, Marian; Tyerman, Jane; Verkuyl, Marg; Goldsworthy, Sandra; Harder, Nicole; Wilson-Keates, Barbara; Kruizinga, Julia; Gumapac, Nathaniel: Effectiveness of Debriefing Methods for Virtual Simulation: A Systematic Review. Clinical Simulation in Nursing, 57:18–30, August 2021.
- [Ma24] Malmqvist, Lars: Sycophancy in Large Language Models: Causes and Mitigations, November 2024. arXiv:2411.15287 [cs].

Literaturverzeichnis

- [Mo22] Molenaar, Inge: The concept of hybrid human-AI regulation: Exemplifying how to support young learners' self-regulated learning. Computers and Education: Artificial Intelligence, 3, January 2022. Number: 100070.
- [PFS16] Palaganas, Janice C.; Fey, Mary; Simon, Robert: Structured Debriefing in Simulation- Based Education. AACN Advanced Critical Care, 27(1):78–85, February 2016.
- [Re12] Reed, Shelly J.: Debriefing Experience Scale: Development of a Tool to Evaluate the Student Learning Experience in Debriefing. Clinical Simulation in Nursing, 8(6):e211– e217, July 2012.
- [SW25] Sun, Yuan; Wang, Ting: Be Friendly, Not Friends: How LLM Sycophancy Shapes User Trust, February 2025. arXiv:2502.10844 [cs].
- [Th90] Thatcher, Donald C: Promoting learning through games and simulations. Simulation & Gaming, 21(3):262–273, 1990.
- [Th11] The Inascl Board Of Directors: Standard VI: The Debriefing Process. Clinical Simulation in Nursing, 7(4):S16–S17, August 2011.
- [Ti13] Tilton, Kathleen J: Non-acute Care Clinical for BSN Students Chronic Illness Care and Diabetes Self-management Support. Doctor of Nursing Practice Systems Change Projects., 2013.
- [TTH15] Tilton, Kathleen J.; Tiffany, Jone; Hoglund, Barbara A.: Non-Acute-Care Virtual Simulation: Preparing Students to Provide Chronic Illness Care:. Nursing Education Perspectives, 36(6):394–395, 2015.
- [Ve18] Verkuyl, Margaret; Atack, Lynda; McCulloch, Tara; Liu, Linda; Betts, Lorraine; Lapum, Jennifer L.; Hughes, Michelle; Mastrilli, Paula; Romaniuk, Daria: Comparison of Debriefing Methods after a Virtual Simulation: An Experiment. Clinical Simulation in Nursing, 19:1–7, June 2018.
- [Za23] Zamfirescu-Pereira, J.D.; Wong, Richmond Y.; Hartmann, Bjoern; Yang, Qian: Why Johnny Can't Prompt: How Non-AI Experts Try (and Fail) to Design LLM Prompts. In: Proceedings of the 2023 CHI Conference on Human Factors in Computing Systems. CHI '23, Association for Computing Machinery, New York, NY, USA, pp. 1–21, April 2023. Number: 437.

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