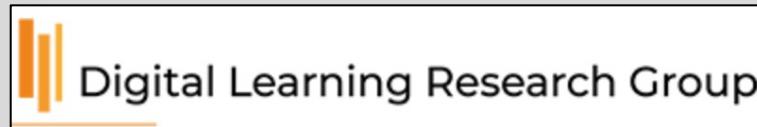


Allgemein-didaktische Unterrichtsmodelle zur Arbeit mit Generativer KI: Himmelstore oder Sackgassen?

Hermann Astleitner

Ao.Univ.-Prof. Mag. Dr.



Vortrag(20+10): Tagung der Sektionen LehrerInnenbildung und LehrerInnenbildungsforschung und
Medienpädagogik der ÖFEB, 25.-26.2.2026, Johannes Kepler Universität Linz.
NO COPY!

I. Hintergrund: EDUCATIONAL THEORY BUILDING LAB

Methoden



Themen

Journal of Research on Computing in Education >
Volume 27, 1995 - Issue 3

Original Articles
A Model for Motivationally Adaptive Computer-Assisted Instruction
Hermann Astleitner & John M. Keller
Pages 270-280 | Published online: 25 Feb 2014
Cite this article | DOI: <https://doi.org/10.1080/08886504.1995.10782132>

SageJournals

Search this journal ▾ Enter search terms... Advanced search

Browse by discipline ▾ Information for ▾

Journal of Educational Computing Research

Impact Factor: 4.9 / 5-Year Impact Factor: 6.0

Restricted access | Research article | First published December 1995 | Request permissions

Learning Strategies for Unstructured Hypermedia—A Framework for Theory, Research, and Practice
Hermann Astleitner and Detlev Leutner View all authors and affiliations

SPRINGER NATURE Link

Find a journal Publish with us Track your research Search

Home > Instructional Science > Article

Designing Emotionally Sound Instruction: The FEASP-Approach
Published: May 2000
Volume 28, pages 169–198, (2000) Cite this article

APA PsycNet® AMERICAN PSYCHOLOGICAL ASSOCIATION

SEARCH ▾ BROWSE ▾

Theory: Designing task-based learning sequences. A categorical model

EXPORT Add To My List Database: APA PsycInfo Chapter

Citation
Astleitner, H. (2007). Theory: Designing task-based learning sequences. A categorical model of task attributes. In H. Astleitner & H.-J. Herber (Eds.), *Task- and standard-based learning: An instructional psychology perspective* (pp. 9–34). Peter Lang Publishing.

Cognitive Overhead in Hypertext Learning Reexamined: Overcoming the Myths
ARTICLE
Joerg Zumbach, University of Heidelberg, Germany
DOI:
Journal of Educational Multimedia and Hypermedia Volume 15, Number 4, October 2006 ISSN 1055-8896 Publisher: Association for the Advancement of Computing in Education (AACE), Waynesville, NC USA

Journal of Instructional Research | Volume 11 | 2022

CLASSROOM ASSIGNMENTS FOR FOSTERING RESILIENCE—AN INSTRUCTIONAL DESIGN MODEL ON AFFECTIVE PERSONALITY DEVELOPMENT
Hermann Astleitner, Paris Lodron University of Salzburg

Higher Education Research & Development

Submit an article ▾ About this journal Browse all articles & issues ▾ Follow this journal ▾

Stay informed when new research is published in this journal
New content alerts

Article
What makes a collegial higher education student? How personality and goal orientations contribute to prosocial behavior >
Hermann Astleitner & Jörg Zumbach
Pages: 1840-1854
Published online: 20 May 2023

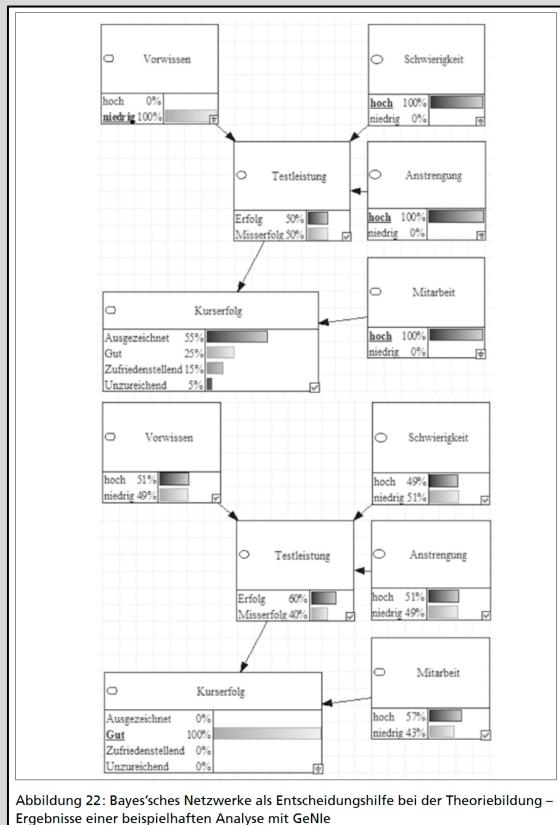
Active Learning in Higher Education
Volume 26, Issue 1, March 2025, Pages 231-254
© The Author(s) 2024, Article reuse guidelines
<https://doi.org/10.1177/14697874241233605>

Sage Journals

Article
The social media use of college students: Exploring identity development, learning support, and parallel use
Hermann Astleitner and Sarah Schlick

I. Hintergrund: EDUCATIONAL THEORY BUILDING LAB

Software, Daten, Statistik



The effects of personality and social media experiences on mental health: Examining the mediating role of fear of missing out, ghosting, and vaguebooking

Hermann Astleitner Amrit Bains Sandra Hörmann

Article

We Have Big Data, But Do We Need Big Theory? Review-Based Remarks on an Emerging Problem in the Social Sciences

Philosophy of the Social Sciences
2024, Vol. 54(1) 69–92
© The Author(s) 2023



Article reuse guidelines:
sagepub.com/journals-permissions
DOI: 10.1177/0048393123118825
journals.sagepub.com/home/pos



Hermann Astleitner¹

Will Artificial Intelligence Change the Creativity in Theory Building?

—
An Interdisciplinary Field of Discourse

<https://www.utb.de/doi/book/10.36198/9783838534619>

II. Ausgangspunkt: GenAI-Ambivalenz

GenAI – Ranking

<https://artificialanalysis.ai/leaderboards/models>

Effektivität GenAI

Gu, J., & Yan, Z. (2025). Effects of GenAI interventions on student academic performance: A meta-analysis. *Journal of Educational Computing Research*, 63(6), 1460-1492.

$$g = 0.68 : R^2 = 0.11$$

<https://visible-learning.org/hattie-ranking-influences-effect-sizes-learning-achievement/>

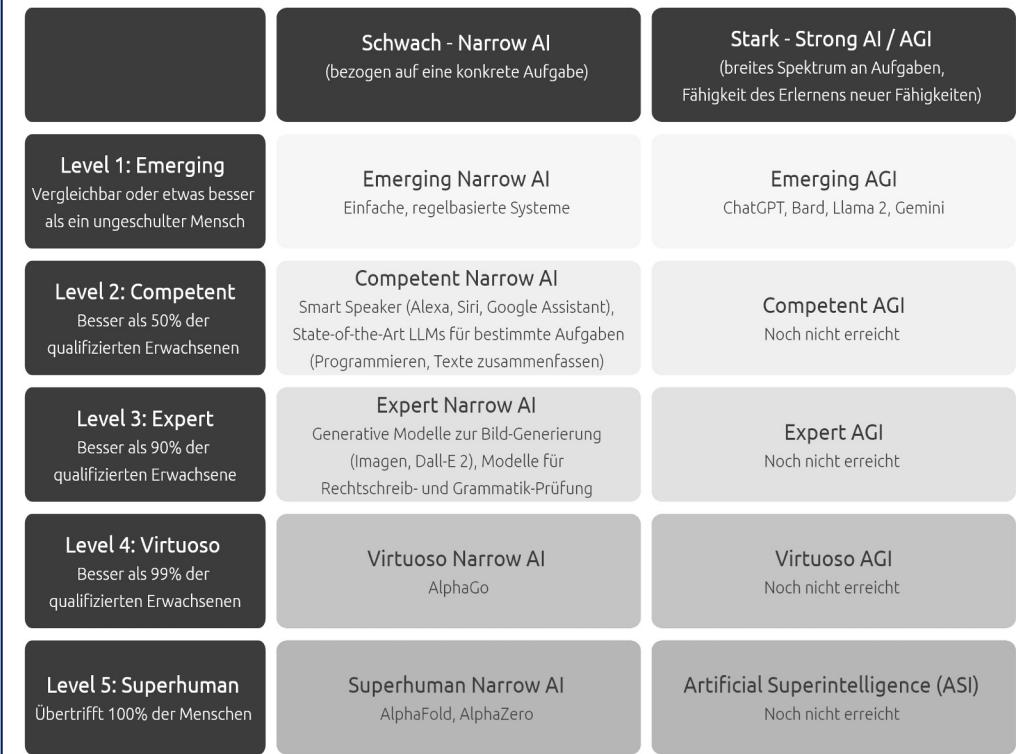
Lehrer:in: 0.32

Lehrer:innenausbildung: 0.12

Co-/Team-Teaching: 0.19

d.velop blog

Starke vs. Schwache KI – Wo stehen wir aktuell?

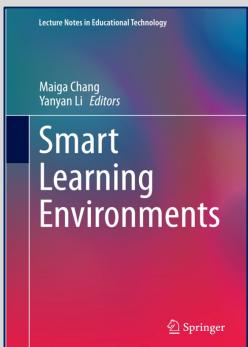


<https://www.d-velop.de/blog/wp-content/uploads/2024/02/starke-vs-schwache-ki.jpg>

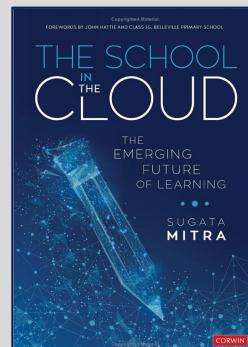
II. Ausgangspunkt: GenAI-Ambivalenz

GenAI-Potential

- Problemlösung/Prävention
(Gestaltung, Diagnose, Intervention, Evaluation)
 - Co-Intelligence
- Augmented Management
 - Big Data Analysis
 - Monitoring
 - Digital Forensics



ohne Lehrer:in

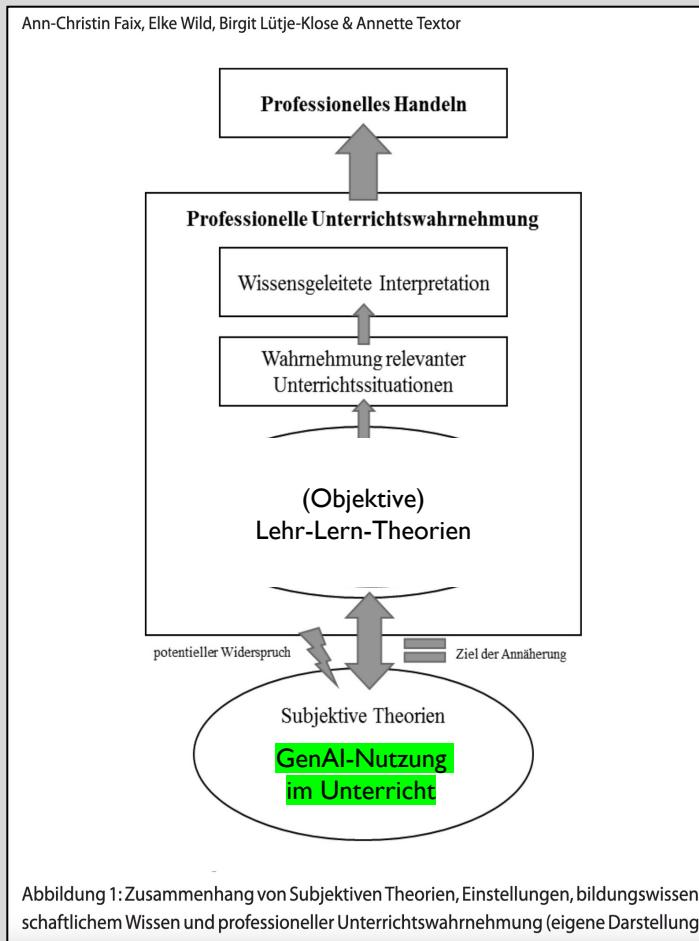


ohne Schulhaus

GenAI-Richtlinien

- Relevanz als Gadget oder Werkzeug?
 - Nutzungsregeln und –kontrolle
- Technologieakzeptanz und –mehrwert
 - Hochwertige Prompts
- Hohes Vorwissen
- Kritische Prüfung
- **Subjektive und objektive Lehr-Lerntheorie**

II. Ausgangspunkt: GenAI-Professionalität



Faix, A. C., Wild, E., Lütje-Klose, B., & Textor, A. (2019). Professionalisierung für inklusiven Unterricht im Rahmen interdisziplinärer und videogestützter Lehrveranstaltungen. *Journal für Psychologie*, 27(2), 71–94. (Modifiziert)

Berkman, E. T., & Wilson, S. M. (2021). So useful as a good theory? The practicality crisis in (social) psychological theory. *Perspectives on Psychological Science*, 16(4), 864-874.



III. Fragestellung und Methodik

Fragestellung

In der vorliegenden Arbeit geht es um die Frage, ob es **allgemein-didaktische Lehr-Lern-Theorien** zur Arbeit mit generativer Künstlicher Intelligenz (GenAI wie z.B. ChatGPT) gibt und wenn ja, welche Qualität diese aufweisen.

Drei Typen von Qualitäten werden in einem Modell integriert und repräsentieren einen **innovativen Bewertungsrahmen** für Research-and-Development-Vorhaben im Bereich der Unterrichtsforschung.

Methodik

Methodisch wird ein narrativer **Literaturreview** durchgeführt (vgl. Baumeister & Leary, 1997). Als Ergebnis dieses Literaturreviews sind eine Reihe von didaktischen Modellen identifiziert worden, die mit dem **Bewertungsrahmen evaluiert** werden. Dabei wird festgestellt, ob ein bestimmtes Qualitätsmerkmal gegeben ist oder nicht.

IV. Evaluationskriterien: Qualitätsmerkmale einer Lehr-Lerntheorie

Klarheit, Konsistenz, Sparsamkeit, Überprüfbarkeit, empirische Angemessenheit, Produktivität, Generalisierbarkeit, Integration, Nutzen, Praktikabilität und Wirkung (Prochaska et al. 2008)

Theoretisch	Empirisch	Praktisch
T1. Lernen unter Lehren	E1. Keine Evidenz: Subjektive /Wiss. Theorie	P1. Handlungsempfehlungen
T2. Lerneffekte/Entwicklungsziele	E2. Schwache Evidenz: Qualitative Studien, Befragungen	P2. Didaktikbeispiele
T3. Lehr-/Lernmethoden	E3. Mittelstarke Evidenz: (Quasi-)experimentelle Studien	P3. Unterrichtsmaterialien
T4. Bedingungen der Anwendung	E4. Starke Evidenz: Langzeittest im Feld/Metaanalyse	P4. Fortbildungsmaßnahmen

Reigeluth, C. M. (2013). What is instructional-design theory and how is it changing? In C. M. Reigeluth (Ed.), *Instructional-design theories and models* (pp. 5–29). Routledge.

John, K. S., & McNeal, K. S. (2017). The strength of evidence pyramid: One approach for characterizing the strength of evidence of geoscience education research (GER) community claims. *Journal of Geoscience Education*, 65(4), 363–372.

Burke, L. A., & Hutchins, H. M. (2008). A study of best practices in training transfer and proposed model of transfer. *Human Resource Development Quarterly*, 19(2), 107–128.

V. Ergebnisse

V.I. Multidimensionales Engagement

Table 1. Instructional Strategies for Fostering Multidimensional Engagement

Levels	Goals	General Strategies in Elements of Learning
Cognitive Engagement		
Knowledge	Stimulating rehearsal/repetition and encoding	-Repeat knowledge to be learned in intervals -Use memory aids
Comprehension	Making thinking explicit and visible	-Think aloud -Use visual representations
Convergent thinking	Varying task-contexts	-Make tasks the core elements of instruction -Diversify tasks
Evaluation	Promoting critical-analytic thinking	-Stimulate multiperspective reasoning -Cultivate standard-based knowledge revision
Synthesis	Supporting divergent thinking/idea generation	-Assist in theory building -Foster system-thinking
Motivational Engagement		
Attention	Allowing choice-making and classroom structuring	-Permit to select/modify task assignments -Use activity schedules
Relevance	Generating utility value and multiple perspectives	-Communicate and self-generate utility value information -Strive for multiple goals
Interest	Being cool and dynamic	-Use popular topics -Use changing topics
Identification	Stimulating mastery orientation and positivity	-Focus on individual progress -Increase booster thoughts and behaviors
Intrinsic motivation	Enhancing fantasy and curiosity	-Establish game-like activities -Stimulate discovery learning
Social-emotional Engagement		
Self-assertion	Considering prosocial contexts	-Foster identity building -Balance power
Entertainment	Covering enjoyment and emotional needs	-Offer sensations -Include moving experiences
Belongingness	Promoting acceptance and commitment	-Emphasize similarities and complementarities -Include service learning activities
Adaptiveness	Practicing mindfulness	-Forcing perspective recognition -Reducing prejudice and stereotyping
Security	Establishing nonthreatening atmospheres	-Striving for secure attachment -Building resilience

Chi, M. T., & Wylie, R. (2014). The ICAP framework: Linking cognitive engagement to active learning outcomes. *Educational Psychologist*, 49(4), 219–243.



Astleitner, H. (2018). Multidimensional Engagement in Learning - An Integrated Instructional Design Approach. *Journal of Instructional Research*, 7, 6–32.



Yuan, L., & Liu, X. (2025). The effect of artificial intelligence tools on EFL learners' engagement, enjoyment, and motivation. *Computers in Human Behavior*, 162, 108474.

Bewertung

T1: Ja

T2: Ja

T3: Ja

T4: Nein

E1: Ja

E2: Nein

E3: Nein

E4: Nein

P1: Ja

P2: Ja

P3: Nein

P4: Nein

V. Ergebnisse

V.2. Lehrzieltaxonomien

RECOMMENDATION		AI CAPABILITIES	DISTINCTIVE HUMAN SKILLS
CREATE	Review	Suggest a range of alternatives, enumerate potential drawbacks and advantages, describe successful real-world cases	Formulate original solutions incorporating human judgement, collaborate spontaneously
EVALUATE	Review	Identify pros and cons of various courses of action, develop rubrics	Engage in metacognitive reflection, holistically appraise ethical consequences of alternative courses of action
ANALYZE	Version 1.0 (2023) Amend	Compare and contrast data, infer trends and themes, compute, predict	Critically think and reason within the cognitive and affective domains, interpret and relate to authentic problems, decisions, & choices
APPLY	Review	Make use of a process, model, or method to illustrate how to solve a quantitative inquiry	Operate, implement, conduct, execute, experiment, and test in the real world; apply creativity and imagination to idea & solution development
UNDERSTAND	Review	Describe a concept in different words, recognize a related example, translate	Contextualize answers within emotional, moral, or ethical considerations
REMEMBER	Amend	Recall factual information, list possible answers, define a term, construct a basic chronology	Recall information in situations where technology is not readily accessible

Bewertung

T1: Ja

T2: Ja

T3: Ja

T4: Nein

E1: Ja

E2: Nein

E3: Nein

E4: Nein

P1: Ja

P2: Ja

P3: Ja

P4: Ja



Bloom's Taxonomy Revisited

Use this table as a reference for evaluating and making changes to aligned course activities and assessments (or, where possible, learning outcomes) that account for generative Artificial Intelligence (AI) tool capabilities and distinctive human skills.

All course activities and assessments will benefit from **review** given the capabilities of AI tools; those at the **Remember** and **Analyze** levels may be more likely to need **amendment**.

Version 1.0 (2023)



Attribution 4.0 International (CC BY 4.0)

<https://ecampus.oregonstate.edu/faculty/artificial-intelligence-tools/blooms-taxonomy-revisited-v1-2023.pdf>

<https://www.iqesonline.net/unterrichten/aufgaben/werkzeuge-kompetenzrad-fragewuerfel-aufgabenmap/>

V. Ergebnisse

V.3. Self-Determination Theory

Table 2 Learning activities using ChatGPT and their impact on the SDT needs and SRL phases derived through the participants' consensus

Learning Activities	Autonomy	Competence	Relatedness	Forethought	Performance	Self-reflection
When teachers design learning activities, students use ChatGPT to ...						
#1 Search information	X			X		
#2 Get examples	X			X		
#3 Check their answers		X				X
#4 Generate review questions to check for their understanding		X				X
#5 Create new problems for practice		X				X
#6 Create challenging problems		X				X
#7 Get insight into complex problems	X				X	
#8 Ask ideas for their improvement		X				X
#9 Make lists or outlines	X			X		
#10 Summarize their own work			X		X	
#11 Ask for definitions		X		X		
#12 Generate questions for discussions	X				X	
#13 Generate questions for essays	X				X	
#14 Get feedback for their work		X				X
#15 Practice peer feedback		X			X	
#16 Prepare for tough conversations		X			X	
#17 Visualize a problem			X		X	
When teachers design learning activities, students are expected to ...						
#18 Anticipate ChatGPT's outputs		X				X
#19 Grade ChatGPT's outputs		X				X
#20 Debate with ChatGPT	X				X	

Chiu, T. K. (2024). A classification tool to foster self-regulated learning with generative artificial intelligence by applying self-determination theory: a case of ChatGPT. *Educational Technology Research and Development*, 72(4), 2401–2416.

Bewertung

T1: Ja

T2: Ja

T3: Ja

T4: Nein

E1: Ja

E2: Ja

E3: Nein

E4: Nein

P1: Ja

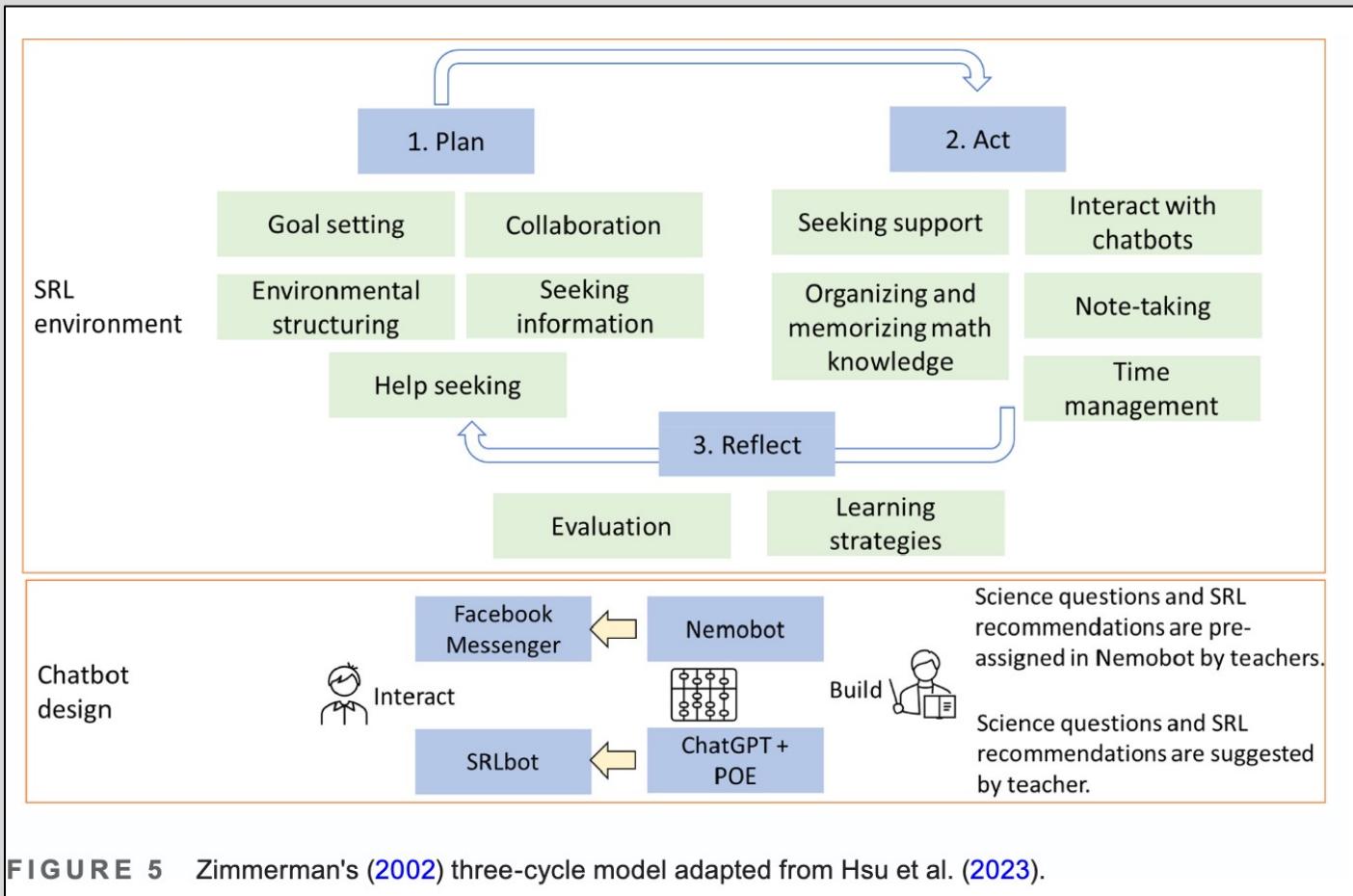
P2: Nein

P3: Nein

P4: Nein

V. Ergebnisse

V.4. Selbstregulation



Bewertung

T1: Ja

T2: Ja

T3: Nein

T4: Nein

E1: Ja

E2: Ja

E3: Nein

E4: Nein

P1: Ja

P2: Nein

P3: Nein

P4: Nein

FIGURE 5 Zimmerman's (2002) three-cycle model adapted from Hsu et al. (2023).

Ng, D.T.K., Tan, C.W., & Leung, J. K. L. (2024). Empowering student self-regulated learning and science education through ChatGPT: A pioneering pilot study. *British Journal of Educational Technology*, 55(4), 1328–1353.

V. Ergebnisse

V.5. Metakognition

TABLE 1 Metacognitive support framework.

Dimension of prompts	Examples
A. During the task initiation phase (planning)	
Identifying specific learning objectives	What is the main objective of this task?
Planning how to use the GenAI tools	What questions do I need to ask ChatGPT to get useful information?
Clarifying their prior knowledge	What prior knowledge will I rely on when interacting with GenAI?
B. During the task execution phase (monitoring and reflection)	
Assessing the relevance and accuracy of the AI responses	Is this information accurate and relevant to my task?
Summarizing their understanding of the content	Can I explain this concept in my own words?
Adjusting strategies as needed	Do I need to ask more specific questions or improve my methods?
C. Task completion phase (evaluation and reflection)	
Assessing the achievement of initial learning goals	Have I achieved my goals? What else can I improve next time?
Reflecting on GenAI tools' contribution to comprehension	How has ChatGPT helped me understand the topic?
Planning improvements for future learning	What strategies will I use next time to improve my learning outcomes?
[Participants can record their thoughts in the blank space.]	

Xu, X., Qiao, L., Cheng, N., Liu, H., & Zhao, W. (2025). Enhancing self-regulated learning and learning experience in generative AI environments: The critical role of metacognitive support. *British Journal of Educational Technology*, 56(5), 1842–1863.

Bewertung

T1: Ja

T2: Ja

T3: Ja

T4: Nein

E1: Ja

E2: Nein

E3: Nein

E4: Nein

P1: Ja

P2: Nein

P3: Nein

P4: Nein

V. Ergebnisse

V.6. Instruktionale Interaktion

Table 5 Summary of the students' and instructors' perceptions of AI systems in online learning

Factor of learner-instructor interaction	The impact of AI systems	Students' perceptions	Instructors' perceptions
Communication	Quantity & Quality	(+) Students believe that the anonymity afforded by AI would make them less self-conscious and, as a result, allow them to ask more questions	(+) Instructors believe that AI could help answer simple, repetitive questions, which would allow them to focus on more meaningful communication with students
	Responsibility	(-) Students worry that AI could give unreliable answers and negatively impact their grades	(-) Instructors predicted conflicts between students and the instructor due to AI-based misunderstandings or misleadingness
Support	Just-in-time support	(+) Students believe that AI would support personalized learning experiences, particularly with studying and group projects	(+) Instructors believe AI could be effectively leveraged to help students receive just-in-time personalized support
	Agency	(-) Students perceived that canned and standardized support from AI might have a negative influence on their ability to learn effectively	(-) Instructors are wary of the fact that too much support from AI could take away students' opportunities for exploration and discovery
Presence	Connection	(+) Students believe that AI can address privacy concerns and support learner-instructor connections by providing social interaction cues without personal camera information	(+) Instructors believe that the addition of AI would help them become more aware of students' needs
	Surveillance	(-) Students are uncomfortable with the measurement of their unconscious behavior, such as eye tracking or facial expression analysis, because it feels like surveillance	(-) Instructors were negative about relying on AI interpretation to understand students' social interaction cues

(+) indicates perceived benefit and (–) indicates perceived concern

Bewertung

T1: Ja

T2: Nein

T3: Ja

T4: Nein

E1: Ja

E2: Ja

E3: Nein

E4: Nein

P1: Ja

P2: Nein

P3: Nein

P4: Nein

V. Ergebnisse

V.7. Spielendes Lernen

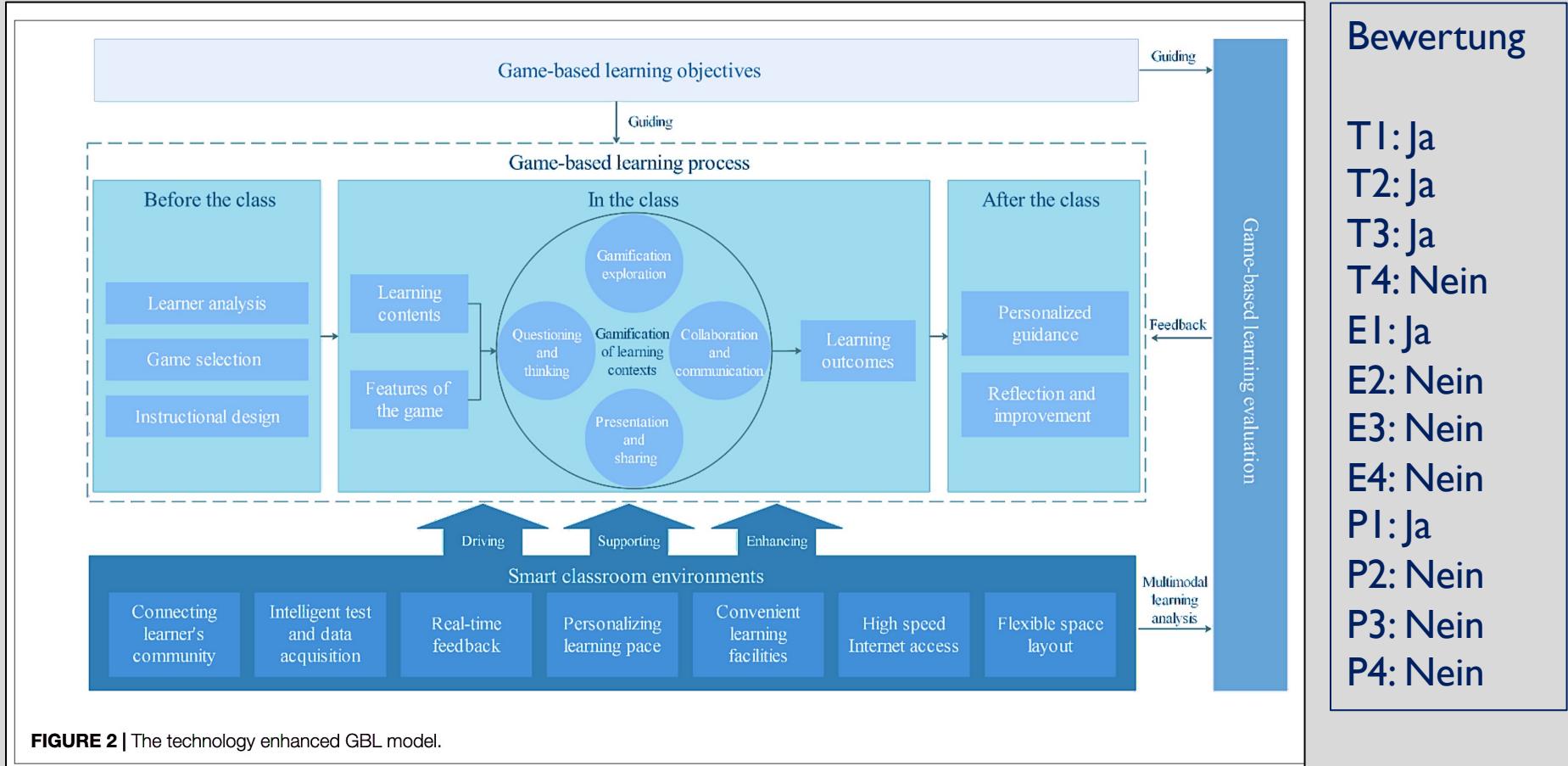


FIGURE 2 | The technology enhanced GBL model.

Pan, L., Tlili, A., Li, J., Jiang, F., Shi, G., Yu, H., & Yang, J. (2021). How to implement game-based learning in a smart classroom? A model based on a systematic literature review and Delphi method. *Frontiers in Psychology*, 12, 749837.

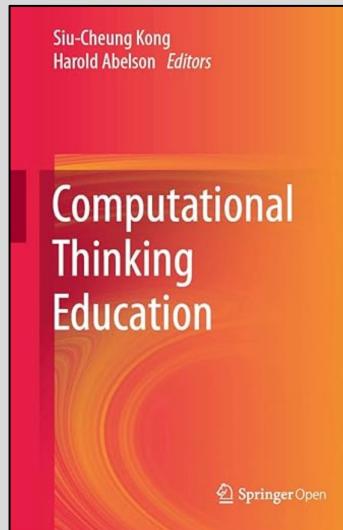
V. Ergebnisse

V.8. Computational Thinking



Figure 3. Diagram of CT practices.

Weng, X., Ye, H., Dai, Y., & Ng, O. L. (2024). Integrating artificial intelligence and computational thinking in educational contexts: A systematic review of instructional design and student learning outcomes. *Journal of Educational Computing Research*, 62(6), 1420–1450.



Bewertung

- T1: Nein
- T2: Ja
- T3: Nein
- T4: Nein
- E1: Ja
- E2: Nein
- E3: Nein
- E4: Nein
- P1: Ja
- P2: Nein
- P3: Nein
- P4: Nein

V. Ergebnisse

V.9. Denkschulung

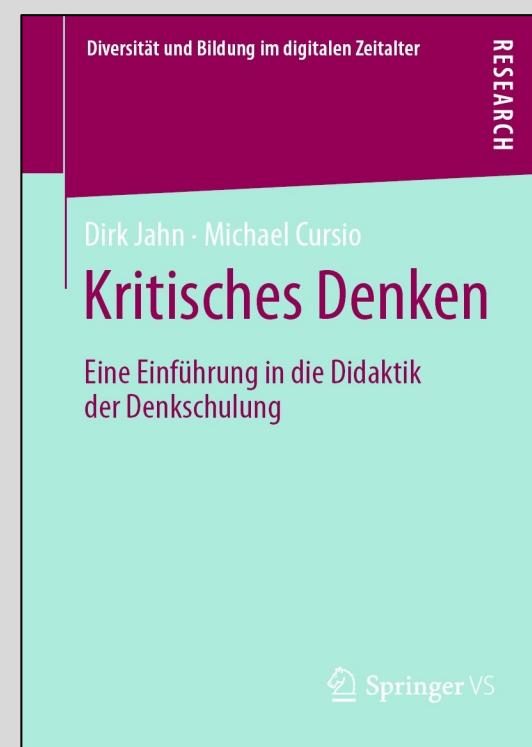
Didaktik der Denkförderung

Prompts zu Ebenen des Kritischen Denkens

Didaktisches Fördermodell
(Practical Inquiry Model von Garrison)

Schrittweises Vorgehen mit Validierung

Jahn, D. (2024). CT goes ChatGPT - Kritisches Denken mit dem Chatbot fördern und einen souveränen Umgang damit kultivieren. In T. Köhler (Hrsg.), *Handbuch E-Learning. Expertenwissen aus Wissenschaft und Praxis* (106. Ergänzungslieferung Februar 2024. Beitrag 4.88). Fachverlag Deutscher Wirtschaftsdienst.



Bewertung

T1: Ja
T2: Ja
T3: Ja
T4: Ja
E1: Ja
E2: Ja
E3: Nein
E4: Nein
P1: Ja
P2: Ja
P3: Ja
P4: Ja

VI. Diskussion: Toolbox für R&D + Denken

Tabelle 4: Theorien in unterschiedlichen Entwicklungsphasen
(vgl. Astleitner, 2011)

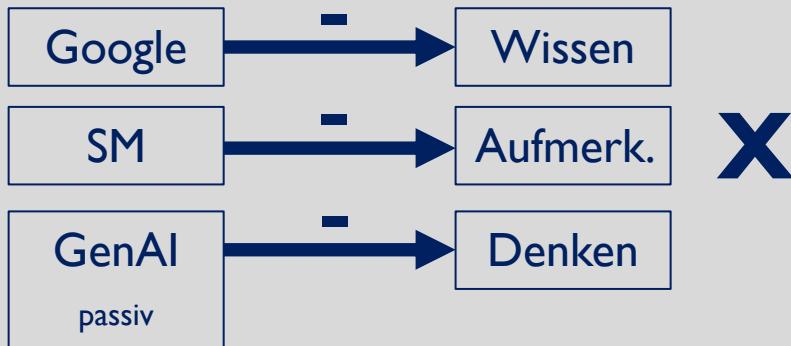
Phase	Phase 1	Phase 2	Phase 3
Zustand	„Naive“ Annahmen Alltagstheorien Einschätzungen	Sammlung von Arbeitshypothesen	Theorie in einer ersten Version
Grundlage	Subjektive Ansichten	Explorationen von wissenschaftlichem Wissen	Literaturbasierte und kriterienorientierte Entwicklung
Phase	Phase 4	Phase 5	Phase 6
Zustand	Wenig geprüfte Theorie	Stark geprüfte Theorie	Weiterentwicklungen einer Theorie
Grundlage	Beschreibende und Zusammenhänge klärende empirische Prüfungen	Kontrollierte Prüfungen von Ursache-Wirkungs- Beziehungen	Reformulierungen aufgrund empirischer und theoretischer Erkenntnisse

NZZ
GASTKOMMENTAR
Klaus Zierer
Im Zeitalter von KI kommt es mehr denn je auf die Tugend des Selberdenkens an

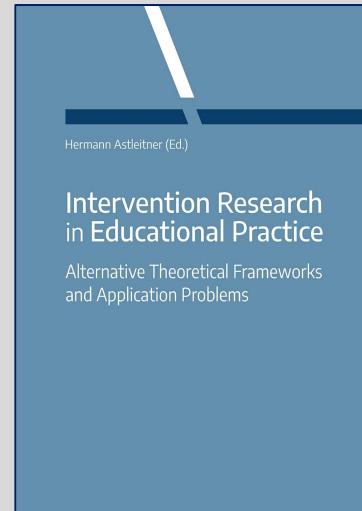
Armitage, R. (2025). Your brain on ChatGPT. *British Journal of General Practice*, 75(758).

The MIT-study suggests that early reliance on LLMs may impair the development of essential cognitive skills.

VII. Implikation: GenAI und Denkregression?



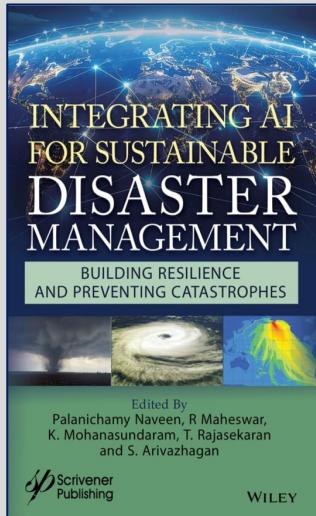
Fehlfunktionen im Unterricht



=
**Denk-
regression**

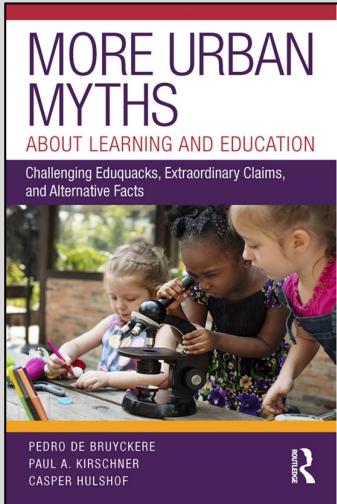
Regression = eine Rückkehr zu einem früheren, niedrigeren Zustand der kognitiven, emotionalen oder verhaltensbezogenen Funktionsfähigkeit.
(<https://dictionary.apa.org/regression>)

Ilgun Dibek, M.,
Sahin Kursad, M., &
Erdogan, T. (2024).
Influence of
artificial
intelligence tools
on higher order
thinking skills: a
meta-analysis.
*Interactive Learning
Environments*, 1-23



VII. Implikation: Denkregression in der Bildung möglich?

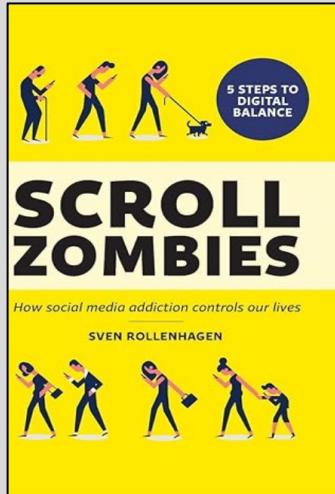
Unwirksames



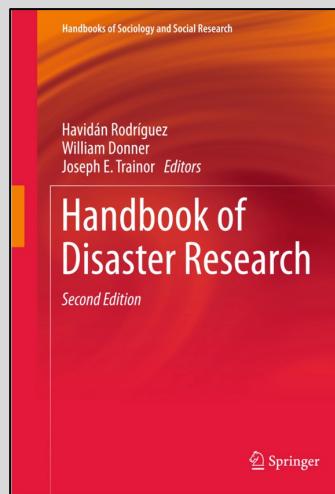
Verstecktes



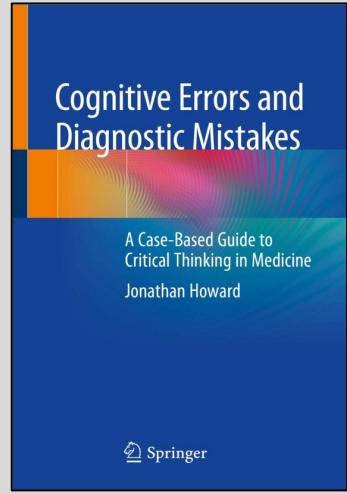
Dunkle Seiten



Nebenwirkung



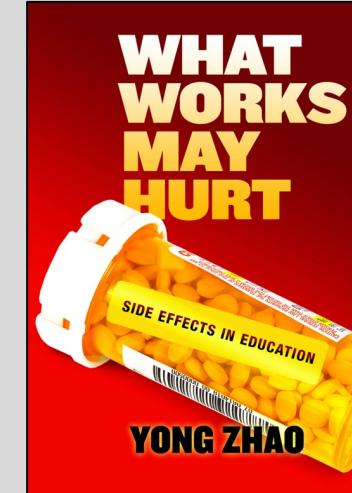
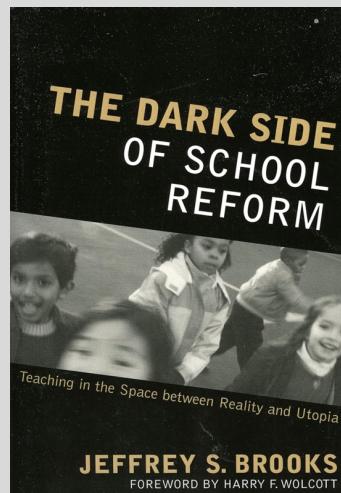
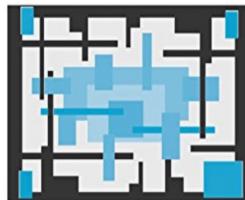
Fehler



Jaap Scheerens
Educational Effectiveness and Ineffectiveness
A Critical Review of the Knowledge Base

Springer

Hans Füchtner
Individuelle und gesellschaftliche Verwahrlosung
Psychoanalytische und sozialpsychologische Diagnosen



Gideon J. Mellenbergh
Counteracting Methodological Errors in Behavioral Research

Springer

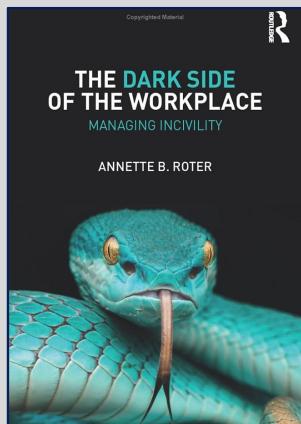
VII. Implikation: Denkgression in der Bildung möglich?

Schönreden Toxische Positivität

Seidel, T. (2014). Angebots-Nutzungs-Modelle in der Unterrichtspsychologie. Zeitschrift für Pädagogik, 60(6), 850-866.

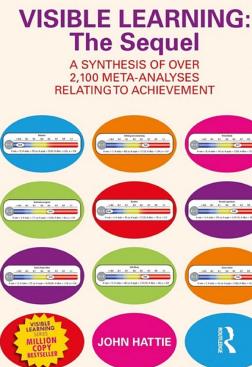
Negativity towards negative results

Matosin, N., Frank, E., Engel, M., Lum, J. S., & Newell, K. A. (2014). Negativity towards negative results: a discussion of the disconnect between scientific worth and scientific culture. *Disease models & mechanisms*, 7(2), 171-173



Ineffektivität Irrelevanz

$$ES = 0.34$$
$$R^2 = 2.8 \%$$



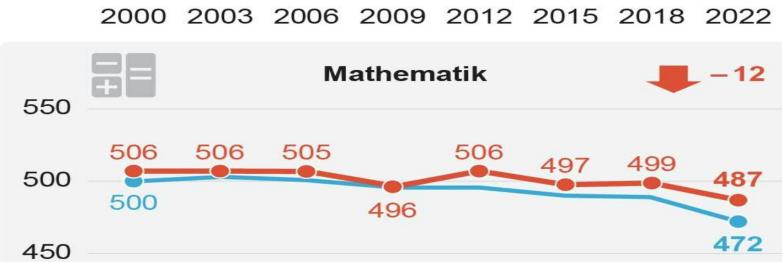
Negative Flynn-Effect

Stillstand Anomie

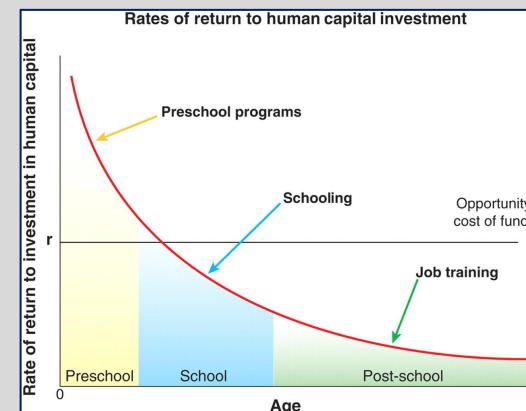
PISA-Test – Österreichs Ergebnisse

15- bis 16-jährige Schülerinnen und Schüler, Punkteschnitt

Österreich Trend 2018–22 OECD



<https://www.vienna.at/pisa-studie-2022-so-haben-osterreichs-schuler-abgeschnitten/8445679>



https://www.unawe.org/static/archives/images/screen/roi_education.jpg

VII. Implikation: Definition Denkregression

Funktionale Stupidität = "Unfähigkeit und/oder der Unwille, kognitive und reflektierende Fähigkeiten über einen engen und vorsichtigen Rahmen hinaus einzusetzen. Sie beinhaltet mangelnde Reflexionsfähigkeit, eine Abneigung gegen die Forderung oder Bereitstellung von Begründungen sowie die Vermeidung substanzialer Argumentation".

Alvesson, M. and A. Spicer (2012) 'A stupidity-based theory of organizations', *Journal of Management Studies*, 49(7): 1194-1220.

Anomie = "Zusammenbruch sozialer Normen, Werte und Erwartungen innerhalb einer Gesellschaft. Sie tritt auf, wenn sich Individuen von gesellschaftlichen Normen entfremdet oder abgeschnitten fühlen, was zu Gefühlen der Ziellosigkeit, Desorientierung und sogar moralischer Verwirrung führt".

<https://oxford-review.com/the-oxford-review-dei-diversity-equity-and-inclusion-dictionary/anomie-definition-and-explanation/>

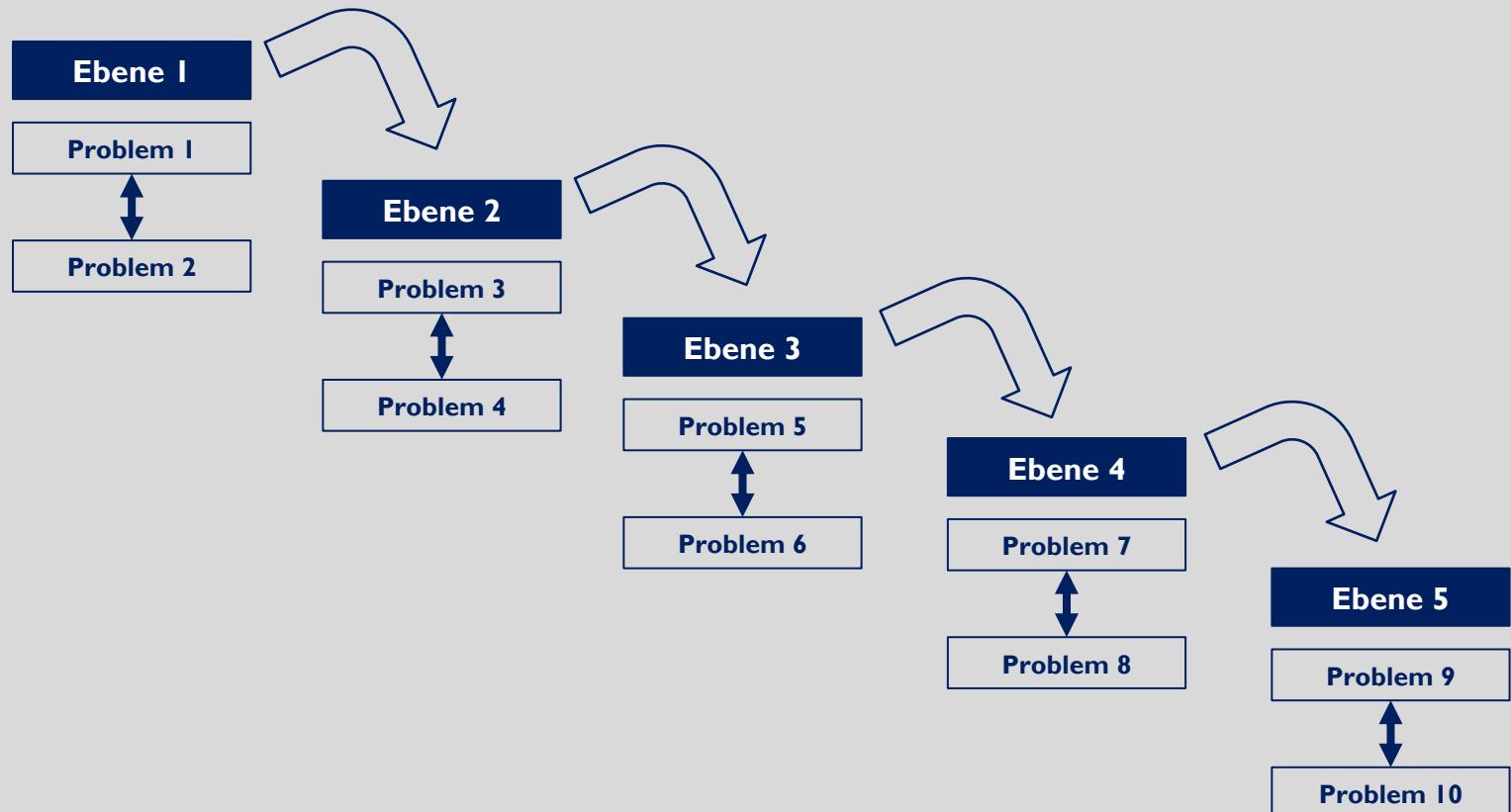
<https://dorsch.hogrefe.com/stichwort/anomie>

= Einschränkung des Denkens und seiner Standards
(= Funktionale Stupidität + Anomie)

VII. Implikation: Wirkmodell einer Denkregression

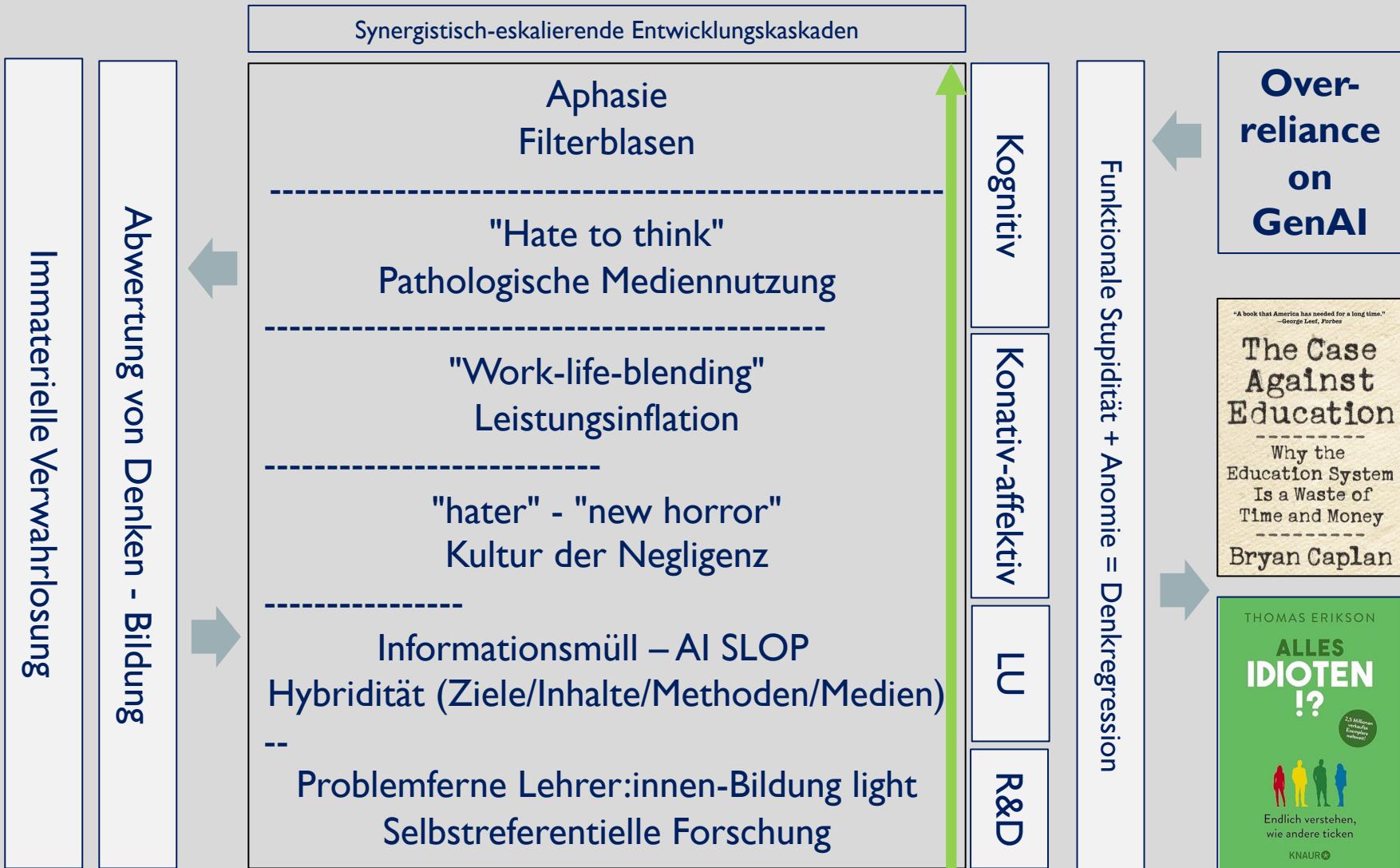
Synergistisch-eskalierende Entwicklungskaskaden

1. Forschung
2. Entwicklung
3. Unterricht
4. Schüler/in:
Konativ-
affektiv
5. Schüler/in:
Kognitiv



Masten, A. S., & Cicchetti, D. (2010). Developmental cascades. *Development and Psychopathology*, 22(3), 491-495.
Dong, H. (2025). A study of the synergistic effects of teacher support and learning engagement on academic achievement in EFL learning—Based on path model and fsQCA analysis. *Psychology in the Schools*.

VII. Implikation: Prototyp einer Theorie der Denkregression



VIII. Literaturhinweise

- Aczel, B., Palfi, B., & Kekecs, Z. (2015). What is stupid?: People's conception of unintelligent behavior. *Intelligence*, 53, 51-58.
- Astleitner, H. (Ed.). (2020). Intervention Research in Educational Practice: Alternative Theoretical Frameworks and Application Problems. Waxmann.
- Astleitner, H. (2000). Designing emotionally sound instruction: The FEASP-approach. *Instructional science*, 28(3), 169-198.
- Astleitner, H. (2007). Theory: Designing task-based learning sequences. A categorical model of task attributes. In H. Astleitner & H.-J. Herber (Eds.), Task- and standard-based learning: An instructional psychology perspective (pp. 9–34). Peter Lang.
- Astleitner, H. (2022). Classroom Assignments for Fostering Resilience--An Instructional Design Model on Affective Personality Development. *Journal of Instructional Research*, 11, 26–43.
- Astleitner, H. (2018). Multidimensional Engagement in Learning--An Integrated Instructional Design Approach. *Journal of Instructional Research*, 7, 6–32.
- Astleitner, H., & Schlick, S. (2025). The social media use of college students: Exploring identity development, learning support, and parallel use. *Active Learning in Higher Education*, 26(1), 231–254.
- Astleitner, H. (2024). We have big data, but do we need big theory? Review-based remarks on an emerging problem in the social sciences. *Philosophy of the Social Sciences*, 54(1), 69–92.
- Karimi-Ghartemani, S., Khani, N., & Nasr Isfahani, A. (2022). Designing a conceptual model for organizational stupidity using a combination of grounded theory and interpretive structural modeling. *Interdisciplinary Journal of Management Studies*, 15(3), 549-568.
- Gong, C., & Yang, Y. (2024). Google effects on memory: a meta-analytical review of the media effects of intensive Internet search behavior. *Frontiers in Public Health*, 12, 1332030.
- Augner, C., Vlasak, T., & Barth, A. (2023). The relationship between problematic internet use and attention deficit, hyperactivity and impulsivity: A meta-analysis. *Journal of Psychiatric Research*, 168, 1-12.
- Ilgun Dibek, M., Sahin Kursad, M., & Erdogan, T. (2025). Influence of artificial intelligence tools on higher order thinking skills: a meta-analysis. *Interactive Learning Environments*, 33(3), 2216-2238.

