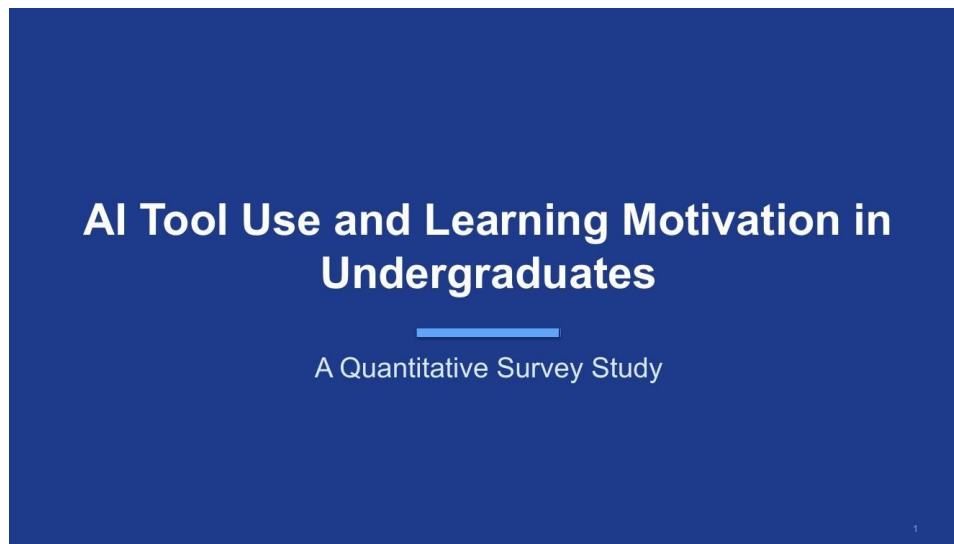


Handout — Use of artificial-intelligence tools and learning motivation among undergraduate students: a quantitative survey study

Slide 1: AI Tool Use and Learning Motivation in Undergraduates



Slide 2: AI in Higher Education: The Motivation Gap

AI in Higher Education: The Motivation Gap

- AI tools are now structurally embedded in undergraduate study.
- Research documents what students do with AI, not what AI does to motivation.
- No study has used a validated instrument to link AI-use frequency to intrinsic motivation.
- Year of study as a moderator is entirely unexplored.

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Slide 3: Two Hypotheses, One Theoretical Anchor

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- SDT: intrinsic motivation flourishes when autonomy, competence, and relatedness are met.
- H1: Higher AI-use frequency → higher intrinsic motivation (positive association).
- H2: Year of study moderates H1 — effect is stronger for first-year students.
- Scaffolding logic: AI reduces cognitive/affective friction most where demands peak.

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Slide 4: Study Design and Sample

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- Cross-sectional online survey; N = 200 undergraduates.
- A-priori power analysis: $\alpha = .05$, power = .80 for moderation model.
- 89 first-year (44.5%) vs. 111 advanced students (55.5%).
- Controls: gender, GPA; mean GPA = 82.4 (SD = 8.3).

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Slide 5: Measures and Analysis Strategy

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- AI-use frequency: researcher-developed 5-point Likert scale ($\alpha = .76$).
- Intrinsic motivation: Academic Motivation Scale composite ($\alpha = .81$).
- H1 tested with Pearson r ; H2 with Hayes PROCESS Model 1.
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Slide 6: H1 Confirmed: Medium Positive Association

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$$r = .34, p < .001$$

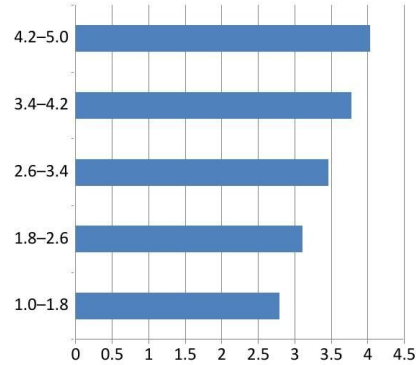
AI-use frequency $\rightarrow$ intrinsic motivation; 95% CI [.21, .46]; $N = 200$

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Slide 7: Intrinsic Motivation Rises with AI-Use Frequency

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- Mean intrinsic motivation increases across frequency bands.
- Lowest-frequency band: M = 2.81; highest: M = 4.05.
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Slide 8: H2 Confirmed: Year of Study Moderates the Effect

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$$\Delta R^2 = .043, p = .015$$

Interaction: AI-use $\times$ year of study; overall model $R^2 = .189$, $F(4,195) = 11.37$, $p < .001$

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Slide 9: First-Year Students Benefit Most

First-Year Students Benefit Most

- First-year slope: $B = 0.41$, $p < .001$ — strong and significant.
- Advanced-year slope: $B = 0.14$, $p = .082$ — smaller, marginal.
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Slide 10: Why First-Year Students Benefit More

Why First-Year Students Benefit More

- Early undergraduates face peak cognitive load and affective uncertainty.
- AI scaffolding lowers the threshold for experiencing competence (SDT).
- Advanced students have internalised strategies — AI adds less motivationally.
- Consistent with Dong (2025): AI motivations vary by year and discipline.
- Cross-sectional design: association, not causation — reverse causality possible.

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Slide 11: Implications and Future Directions

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- Integrate AI tools intentionally into first-year curricula, not uniformly.
- Frame AI as an inquiry scaffold, not a production shortcut.
- Disaggregate evaluation data by year of study — aggregate metrics mislead.
- Future work: longitudinal design, mediation via self-efficacy, multi-institution samples.

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Slide 12: Key Takeaways

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- AI-use frequency positively predicts intrinsic motivation: $r = .34$, medium effect.
- Year of study is a genuine moderator — first-year students benefit most.
- Motivational scaffolding, not digital skill, explains the pattern.
- Policy implication: differentiate AI integration by stage of study.
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