

# Global Conference on Measuring New Forms of Employment

Brussels, 4 - 5 July 2024

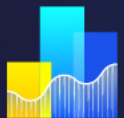


#MeasuringEmployment

# GENERATIVE AI AND THE FUTURE OF EMPLOYMENT: A COLLABORATIVE APPROACH

## Session 6

Professor Aleksandra Przegalinska  
Kozminski University, Harvard, CampusAI



# THE FUTURE OF WORK WITH COLLABORATIVE AI



# WHY COLLABORATIVE AI? (IT'S MORE THAN HUMAN- CETRIC)



# EVOLUTION

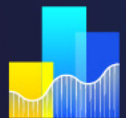
LARGE LANGUAGE  
MODELS +  
DIFFUSION  
MODELS

CONVERSATIONAL  
MODELS

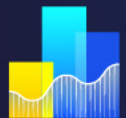
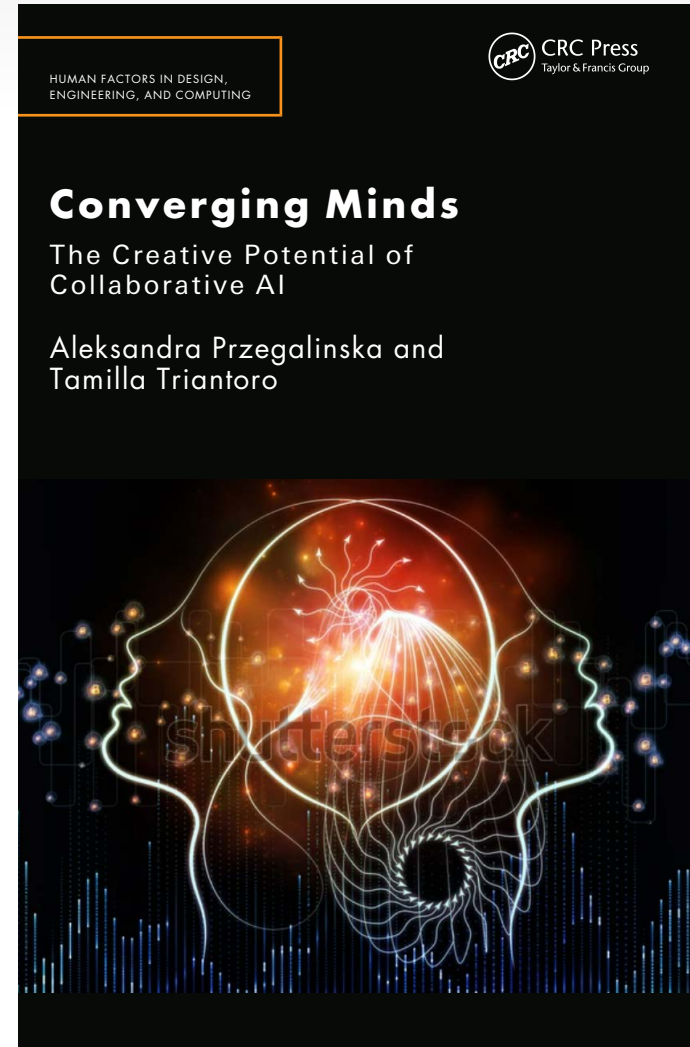
MULTIMODAL  
MODELS

LARGE ACTION  
MODELS

OMNIMODELS

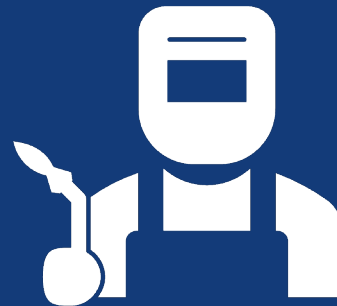
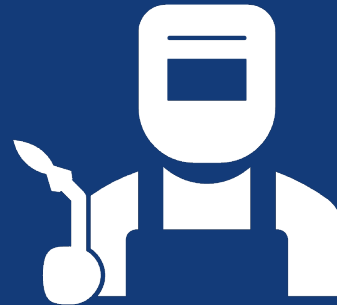


**New book  
written with  
Claude,  
Bard and  
GPT**



## ASSISTANTS (NOT AGENTS):

- CRITIC
- STYLE AMPLIFIER
- IDEA GENERATOR
- FIELD EXPERT



**It greatly elevated  
our academic  
output!**







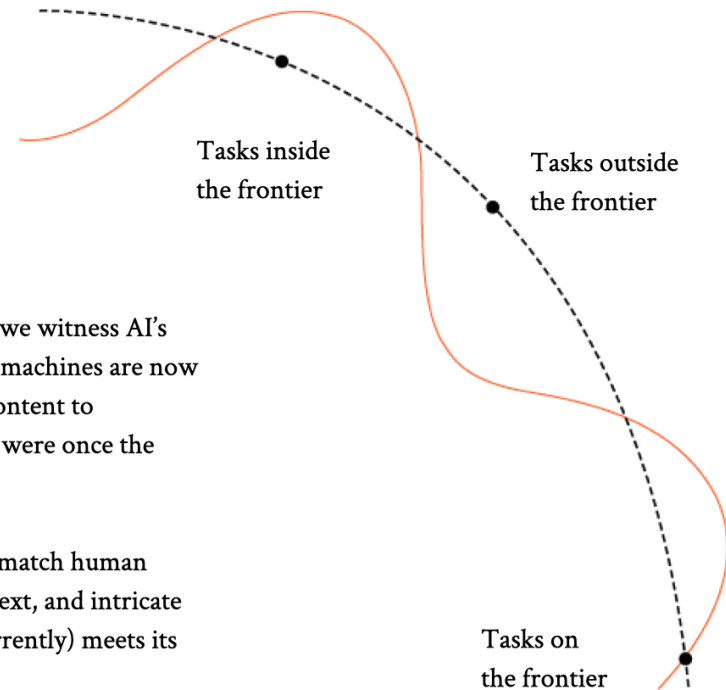
**WE ALREADY  
HAVE EVIDENCE  
CONCERNING AI  
AND  
PRODUCTIVITY**



# THE JAGGED FRONTIER

The jagged frontier works as follows: on one end of the spectrum, we witness AI's remarkable prowess — tasks that once seemed insurmountable for machines are now executed with precision and efficiency (from generating creative content to predicting complex patterns), with AI showcasing capabilities that were once the exclusive domain of human cognition.

Yet, on the flip side, there are tasks where AI falters, struggling to match human intuition and adaptability. These are areas marked by nuance, context, and intricate decision-making — realms where the binary logic of machines (currently) meets its match.



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## OUR FORMER STUDY

### COBOTS IN THE KNOWLEDGE WORK:

**40% - job effectiveness  
increase**

**60% - job satisfaction  
increase**

**<https://www.sciencedirect.com/science/article/pii/S014829632030792X>**



# THEORIES:

- **Resource-Based View** - *dependence of an organization's competitive edge on the unique resources it owns or controls*
- **Task Technology Fit**- *provides a model that details how the characteristics of a task, paired with the attributes of the technology used to perform it, can influence individual performance.*



# HYPOTHESES

- *H1: Firms that possess advanced generative AI resources and capabilities will have a competitive advantage over firms that do not.*
- *H2: Firms with employees more experienced in AI will have a competitive advantage over firms that do not.*
- *H3a: Firms utilizing generative AI for automation tasks will perform these tasks more effectively than firms that do not.*
- *H3b: Firms employing generative AI for decision support tasks will perform these tasks more effectively than firms that do not.*
- *H3c: Firms that integrate generative AI in their creative tasks will perform these tasks more effectively than firms that do not.*
- *H3d: Firms that leverage generative AI for innovation tasks will perform these tasks more effectively than firms that do not.*



Table II. Task classification based on complexity and creativity

<b>Task</b>	<b>Simple</b>	<b>Complex</b>
<b>Routine</b>	Automation task: Persona Ideation	Decision Support Task: Competitive Analysis
<b>Creative</b>	Creation task: Text-based Ad	Innovation task: Product Naming

## Participants n=94

Participants from Executive MBA groups were experienced and had seniority in management related tasks, including marketing.

The participants completed the tasks within a set time frame (45 mins).

**On average participants completed study within 19.5 min (SD = 10.8 min).**

## INDEPENDENT JUDGES QUALITY ASSESSMENT

The quality of the output for each task was evaluated by a panel of independent judges rating participants' responses for the tasks (scale from 1 to 5).

Cronbach's alpha was calculated to measure the **reliability of the created scales**. Cronbach's alpha for the quality of product name scale was 0.84, competitive analysis 0.98, text-based ad 0.97 and persona 0.98

**The methodology also included qualitative review of experiment output by the judges.** The results of the analyses revealed an average inter-judge reliability of 0.464 (Krippendorff Alpha) which indicates a moderate level of agreement among the judges.

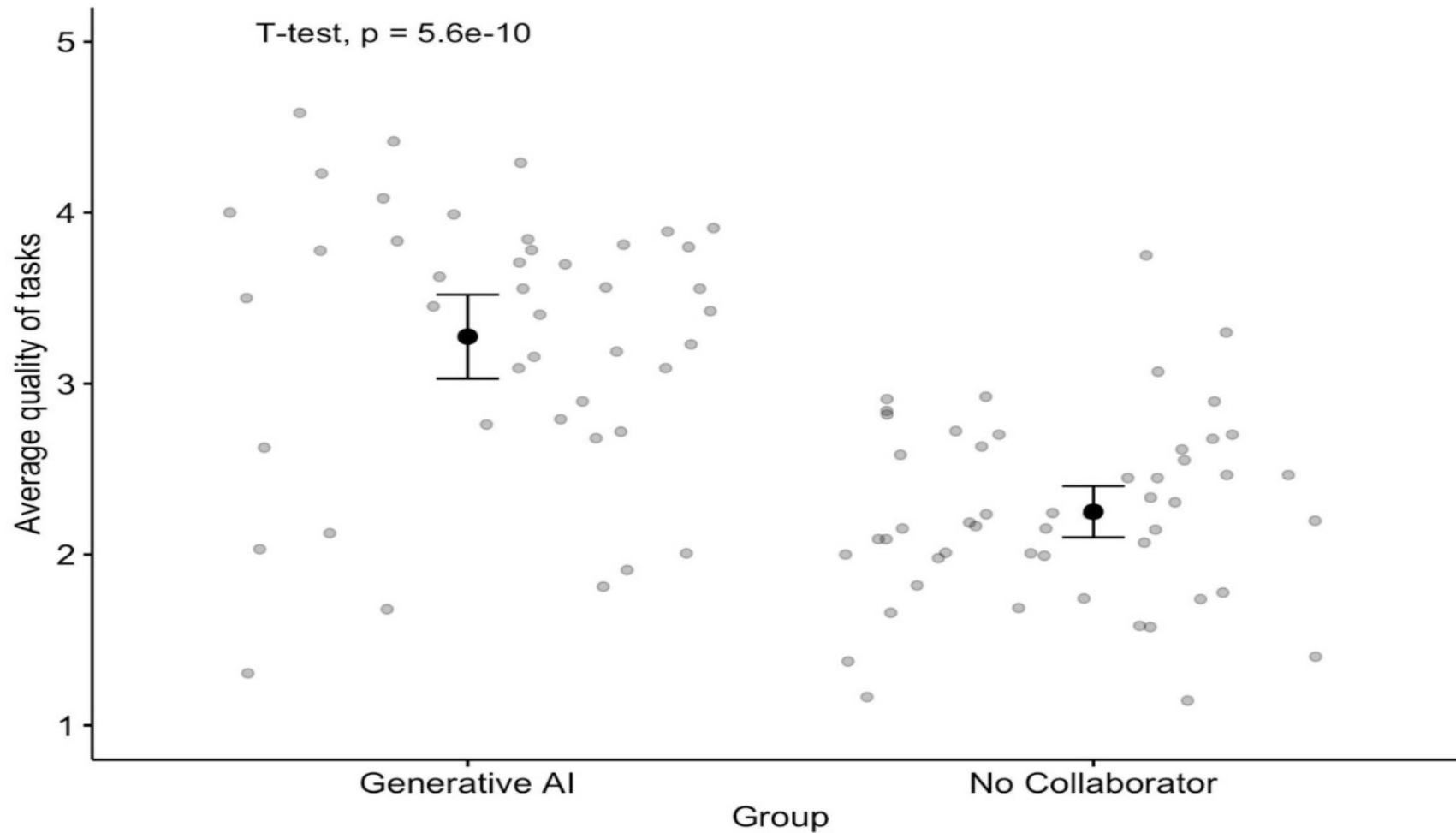


## HYPOTHESES TESTING

### Hypothesis 1:

*Welch Two Sample t-test to examine the difference **in the quality of task output** between participants in the group interacting with Generative AI ( $M = 3.27$ ,  $SD = 0.8$ ,  $n = 43$ ) and the group with No Collaborator ( $M = 2.25$ ,  $SD = 0.54$ ,  $n = 51$ )*





## Hypothesis 2:

*Welch one-way ANOVA to investigate the **differences in the quality of the tasks between the groups with no familiarity, moderate familiarity, significant familiarity with Generative AI.***



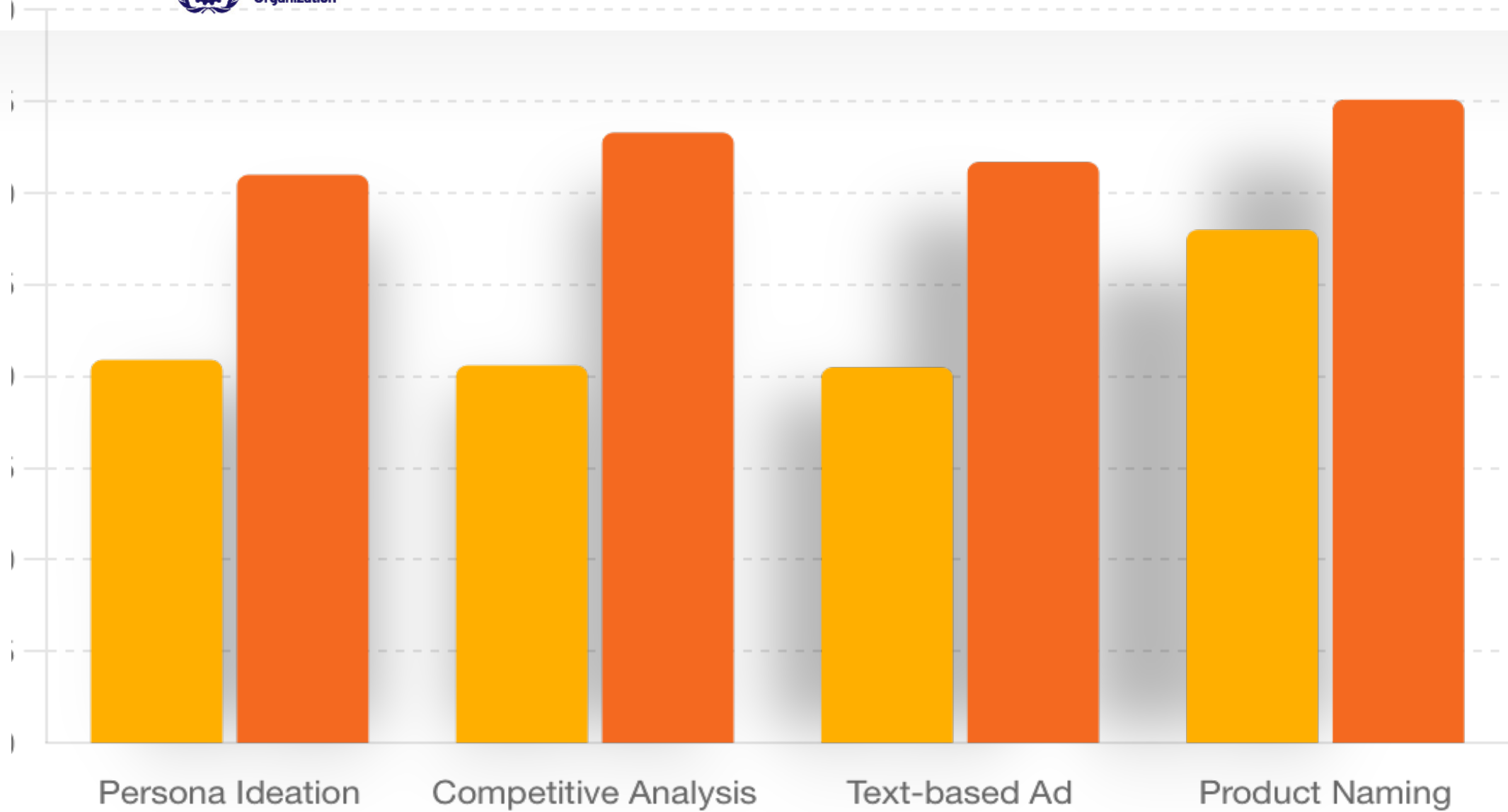
Table III. Descriptive Statistics for Quality of Tasks Completed by Participants with No familiarity  
Moderate familiarity, Significant familiarity with Generative AI.

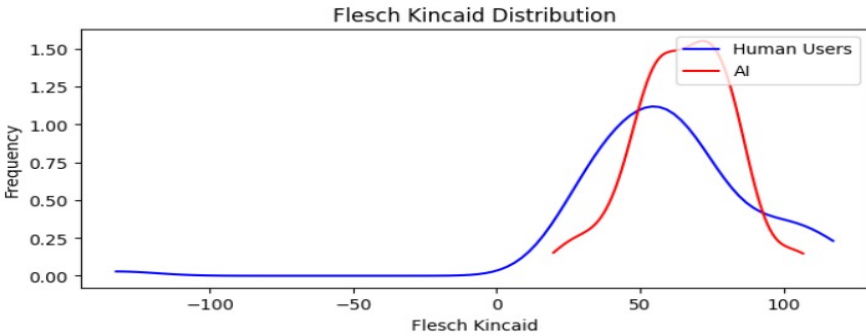
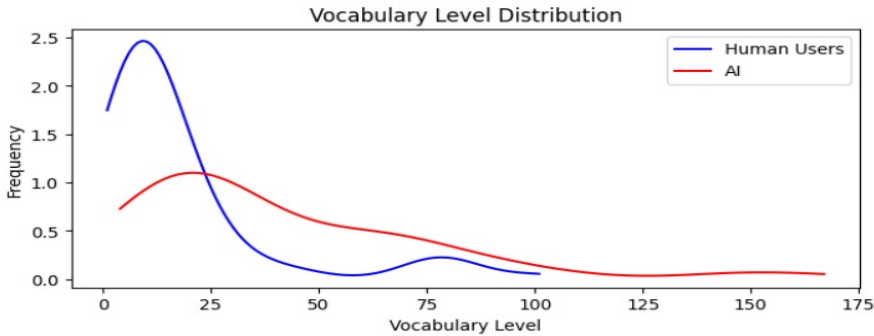
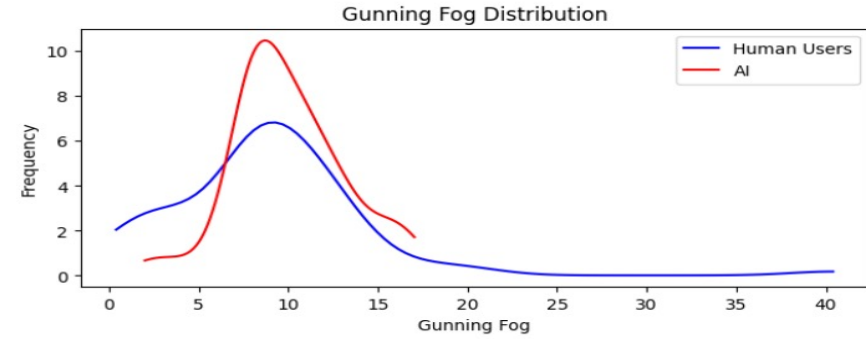
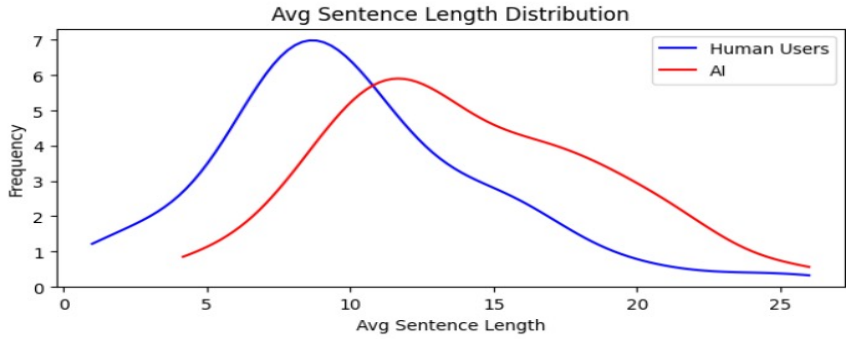
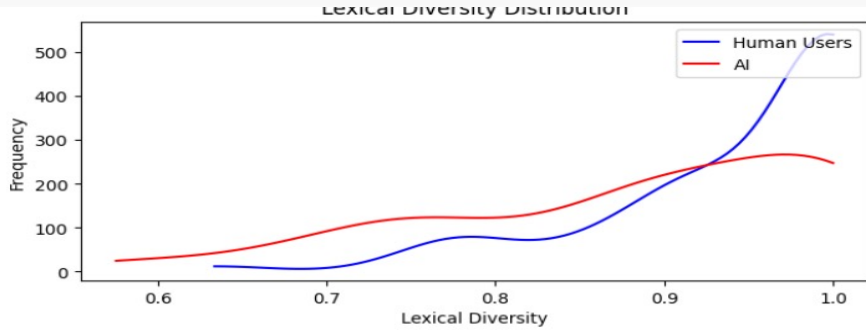
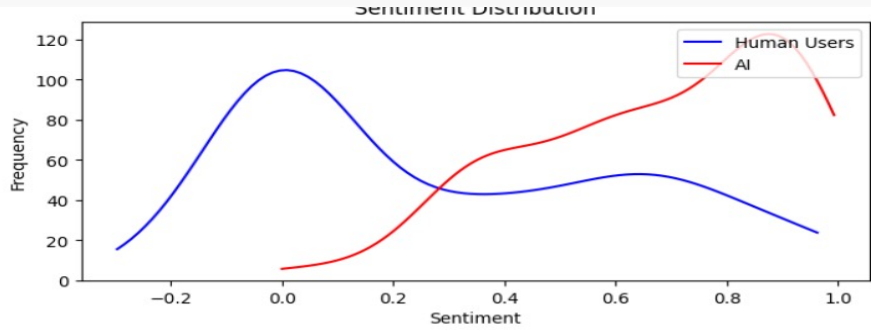
Familiarity	with	Persona Ideation				Competitive Analysis			Text-based Ad			Product Naming		
		<i>N</i>	<i>M</i>	<i>SD</i>	<i>F</i>	<i>M</i>	<i>SD</i>	<i>F</i>	<i>M</i>	<i>SD</i>	<i>F</i>	<i>M</i>	<i>SD</i>	<i>F</i>
No familiarity		12	1.95	0.75	5.18**	1.99	0.84	4.77*	2.10	0.79	3.77*	2.89	0.56	0.44
Moderate familiarity		64	2.54	1.01		2.65	1.08		2.53	0.96		3.14	0.79	
Significant familiarity		13	3.23	1.13		3.31	1.18		3.12	0.96		3.31	0.63	

*The effect size was medium for **product name** and large for **competitive analysis** and **persona ideation tasks**.*

*The medium effect size indicates that there is a moderate, but still noticeable, relationship between the familiarity with Generative AI technology and the quality of the **text-based ad**.*







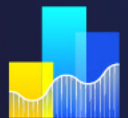


## UNDERSTANDING HUMAN-AI PATTERNS





# AI VS. PERSONALITY



# Methodology

- **Participants:** Varied in age, gender, and employment status.
- **Tasks:** Marketing campaign tasks similar to Study 1
- **Metrics Evaluated:** Task quality, engagement duration, and future AI usage intentions.

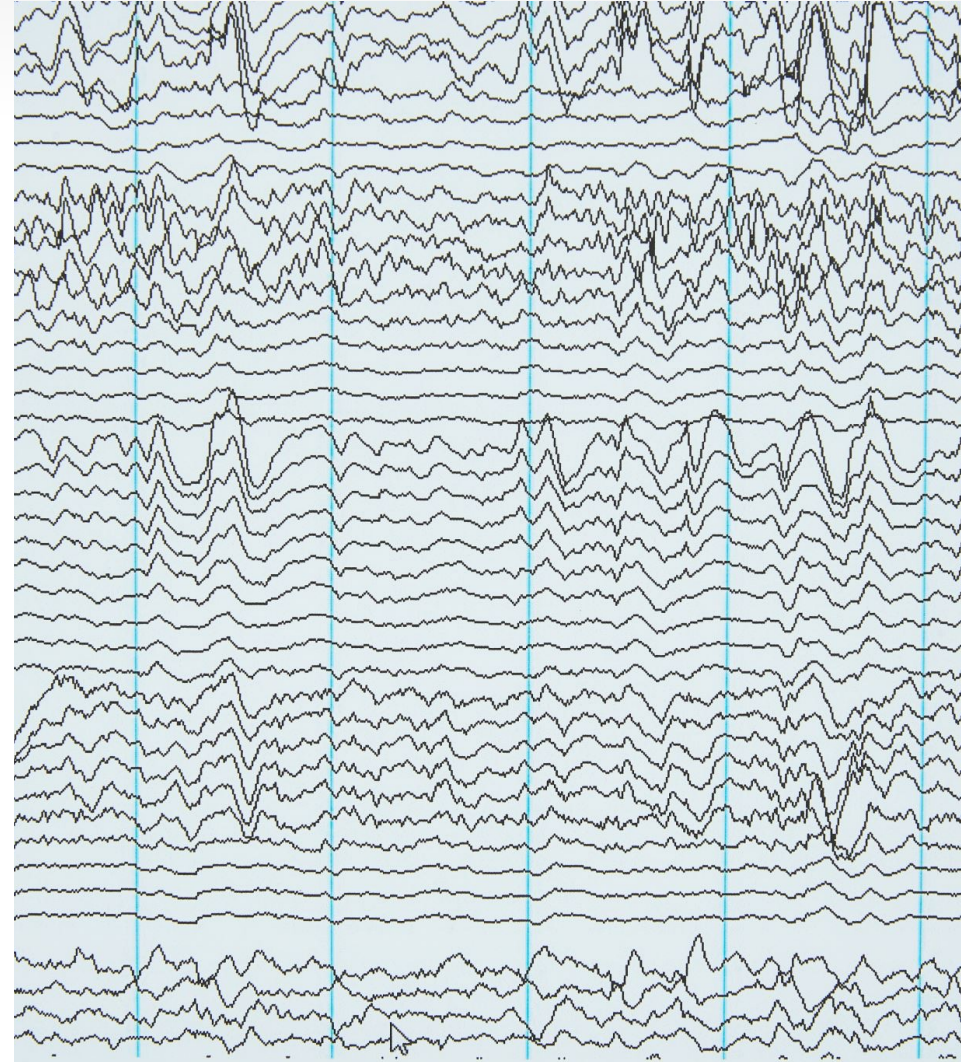


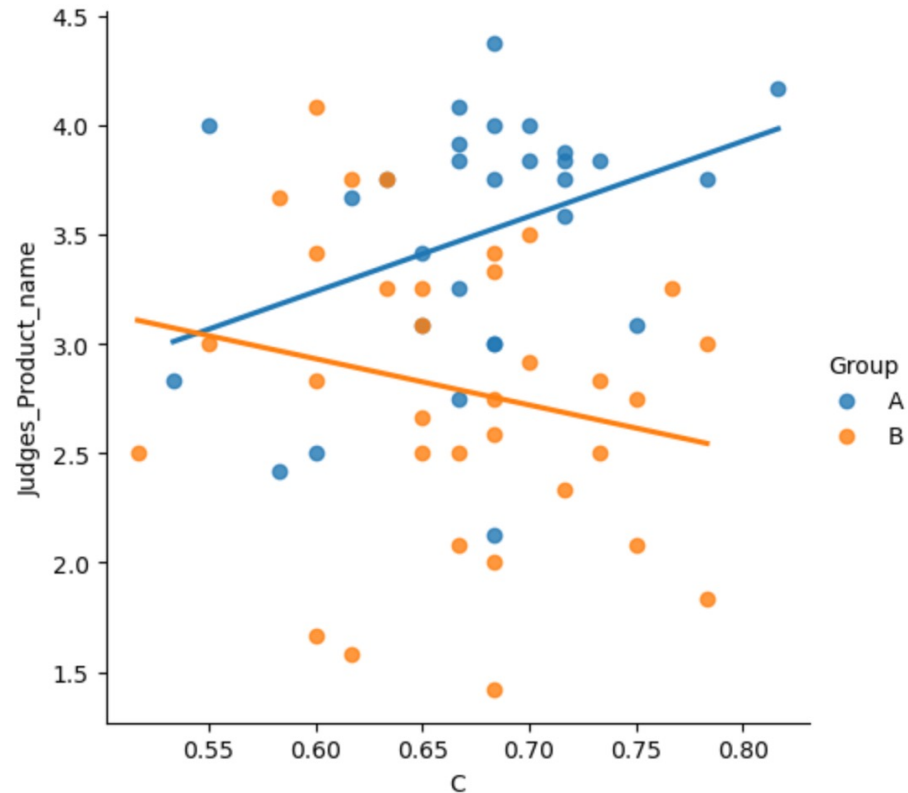
# Methodology

- We used *lsr* library and eta squared function for the calculation of effect size ( $\eta^2$ ) and partial effect size ( $p^2$ ).
- Effect sizes were interpreted based on established guidelines:
  - $p^2$  around 0.01 - small,
  - $p^2$  around 0.06 - medium,
  - $p^2$  around 0.14 and higher - large
- [\(Cohen 2013; Miles and Shevlin 2001\)](#)



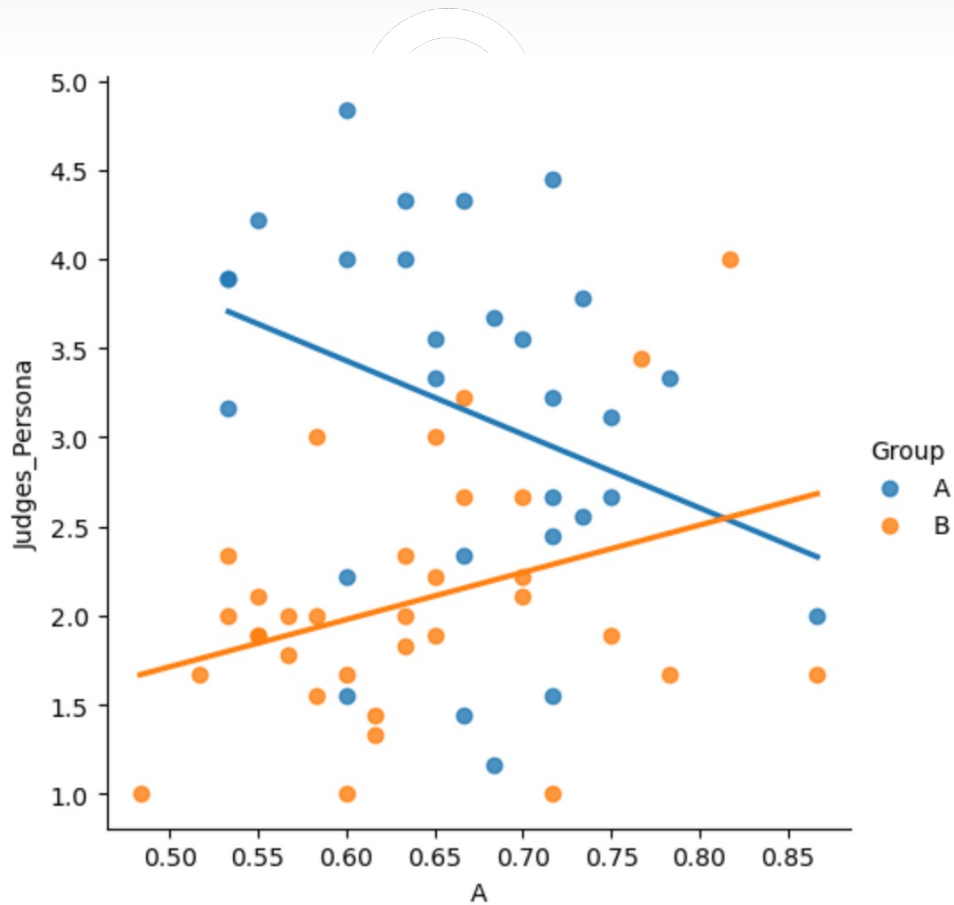
- Following the ANOVA, post hoc analyses including simple slopes and pairwise comparisons were performed using *interactions* and *emmeans* libraries to investigate specific group differences.
- In cases when interaction with the group was insignificant, linear regression analysis using ordinary least squares method (OLS) was conducted to predict performance based on traits.
- Additionally such libraries as *dplyr* and *ggplot2* were used for data exploration and visualization.





## BIG FIVE RESULTS:

"People who are more conscientious and collaborate with a chatbot achieved better quality in persona creation, while less conscientious people who do not collaborate with the bot had better results for product name quality."



## BIG FIVE RESULTS:

"Less agreeable individuals collaborating with a chatbot achieved better quality in all tasks"



## Implications for AI Design

- Incorporate psychological principles into AI design for better user alignment.
- Tailored AI solutions can enhance user satisfaction and productivity.

# www.campus.ai



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# Thank you!



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