

State of Alabama K-12 Generative AI Content Assessment on behalf of Dr. Bartolf % Quanthub

by brian hogan, MS bhogan@clarku.edu

i. Curriculum on Ethical Considerations for AI Applications

- I. [Understand the importance of ethics when using AI applications](#)
 - A. Why is it important to understand the ethics behind AI applications we interact with?
 - B. How can we trust the decisions made by AI applications?
- II. [Navigate ethical challenges related to data privacy and security when using AI applications](#)
 - A. Why is it important to understand how AI applications use our personal data?
 - B. How do AI applications use our personal data?
 - C. What are common data privacy and security mistakes people make when using AI applications and how do you avoid them?
 - D. What are best practices when considering implications of data privacy and security of AI applications?
- III. [Navigate ethical challenges related to bias and discrimination when using AI applications](#)
 - A. Why is it important to consider how bias and discrimination may exist in AI application outcomes?
 - B. How might AI applications reflect or amplify societal biases
 - C. What are common bias and discrimination challenges people encounter when using AI applications and how do you mitigate them?
 - D. What are best practices when evaluating how AI applications might treat different users unfairly?
- IV. [Navigate ethical challenges related to misinformation and disinformation when using AI applications](#)
 - A. Why is it essential to differentiate between credible sources and potential misinformation?
 - B. How can you critically evaluate the information provided by AI applications?
 - C. What are the best practices for identifying misinformation and disinformation when using AI applications?
 - D. What are common misinformation and disinformation challenges people encounter when using AI applications and how do you mitigate them?
- V. [Navigate ethical challenges related to accountability and transparency when using AI applications](#)
 - A. Why are accountability and transparency important in AI applications?
 - B. How are transparency and trust presented in AI applications?
 - C. What are common accountability and transparency challenges people encounter when using AI applications and how do you mitigate them? **<no data bc no questions>**
 - D. What are best practices when ensuring you are using AI applications that prioritize transparency and accountability?

ii. Curriculum on the Impact of AI on the Workforce

- I. [Identify ways that AI is impacting the future of work.](#)
 - A. What are ways in which AI is driving change in the workforce?
- II. [Identify how AI is impacting different industries.](#)
 - A. Which industries will be most impacted by AI?
- III. [Identify how AI is impacting different tasks](#)
 - A. What types of tasks will be automated by AI?
 - B. What types of tasks will be augmented by AI?
- IV. [Identify how AI is impacting skill requirements.](#)
 - A. Which skills are becoming more necessary as a result of AI?
 - B. Which skills are becoming less necessary as a result of AI?

iii. Curriculum for Prompt Engineering Resource Outline

- I. [Understand what prompt engineering is and why it's important when working with generative AI tools](#)
 - A. What is a prompt and how is it used in generative AI?
 - B. What is prompt engineering and why is it important for working with generative AI?
 - C. What are the challenges and limitations associated with prompt engineering in gen. AI?
- II. [Identify common use cases for prompt engineering](#)
 - A. What are use cases of prompt engineering for text generation?
 - B. What are use cases of prompt engineering in image generation?
 - C. What are use cases of prompt engineering in audio generation?
- III. [Understand the prompt engineering process](#)
 - A. What are the typical steps involved in the prompt engineering process?
 - B. What is involved in the process of defining an objective for prompt engineering?
 - C. What is involved in the process of crafting prompts?
 - D. What is involved in the process of evaluating prompt performance?
 - E. What is involved in the process of refining prompts?

0.1. High-level thematic and concepts

1. Digital Citizenship in the AI Age

As AI continues to permeate our daily lives, it's essential for K-12 students to understand their rights, responsibilities, and the etiquette of being a digital citizen in an AI-driven world.

2. Hands-on AI Experiments

Nothing beats practical experience. Encourage K-12 students to play with AI tools available for education. Simple drag-and-drop tools, coding platforms, or AI-powered educational games can offer a hands-on introduction to AI.

3. AI in Arts and Creativity

Often, the conversation around AI is dominated by technical and ethical aspects. However, AI's impact on the creative fields—from music and art to writing and design—is profound. A module exploring this would be a delightful addition, blending creativity with technology.

4. Storytelling with AI

Narratives are powerful tools for understanding complex topics. Crafting narratives or stories around how AI has been or could be used in various scenarios can help students internalize and relate to the material.

5. Career Spotlights

Invite professionals working on the cutting edge of AI, ethics, and prompt engineering to share their experiences. Real-world insights can provide valuable context and inspiration for students.

6. Debate and Discussions

Organize debates on AI's impact on society, ethics, and future jobs. This not only reinforces the material but also encourages critical thinking.

7. AI Ethics Film Series or Book Club

Curate a list of movies, documentaries, or books that touch upon the societal implications of AI and organize viewings or reading sessions followed by discussions.

0.2.1 Curriculum meta data - search categories

<div>→ **AI Principles**</div> <div>↳ Transparency</div> <div>↳ Accountability</div> <div>↳ Trust</div> <div>↳ Ethical considerations</div> <div>↳ Bias</div> <div>↳ Discrimination</div> <div>↳ Fairness</div> <div>↳ Proprietary</div> <div>→ **AI Actions**</div> <div>↳ Decision-making</div> <div>↳ Predictions</div> <div>↳ Outputs</div> <div>↳ Functioning</div> <div>↳ Recommendations</div> <div>→ **AI Components**</div> <div>↳ Algorithms</div> <div>↳ Data</div> <div>↳ Training</div> <div>↳ AI applications</div> <div>↳ Machine learning</div>	<div>→ **Data Handling**</div> <div>↳ Privacy policies</div> <div>↳ Data collection</div> <div>↳ Data use</div> <div>↳ Third parties</div> <div>↳ Misuse</div> <div>↳ Data protection</div> <div>↳ Data sharing</div> <div>→ **Entities & Events**</div> <div>↳ ProPublica</div> <div>↳ U.S. courts</div> <div>↳ Uber</div> <div>↳ Arizona</div> <div>↳ World Health Organization</div> <div>↳ Centers for Disease Control</div> <div>↳ OpenAI</div>	<div>→ **Learning & Tools**</div> <div>↳ AI literacy</div> <div>↳ Courses</div> <div>↳ Microsoft's</div> <div>↳ AI School</div> <div>↳ Research</div> <div>↳ Third-party auditing</div> <div>↳ User interfaces</div> <div>→ **Mistakes & Issues**</div> <div>↳ Errors</div> <div>↳ Failures</div> <div>↳ Miscommunication</div> <div>↳ Misunderstanding</div> <div>↳ Conflict</div> <div>→ **Solutions**</div> <div>↳ Countermeasures</div> <div>↳ Quality control</div> <div>↳ Feedback loops</div> <div>↳ External audits</div>
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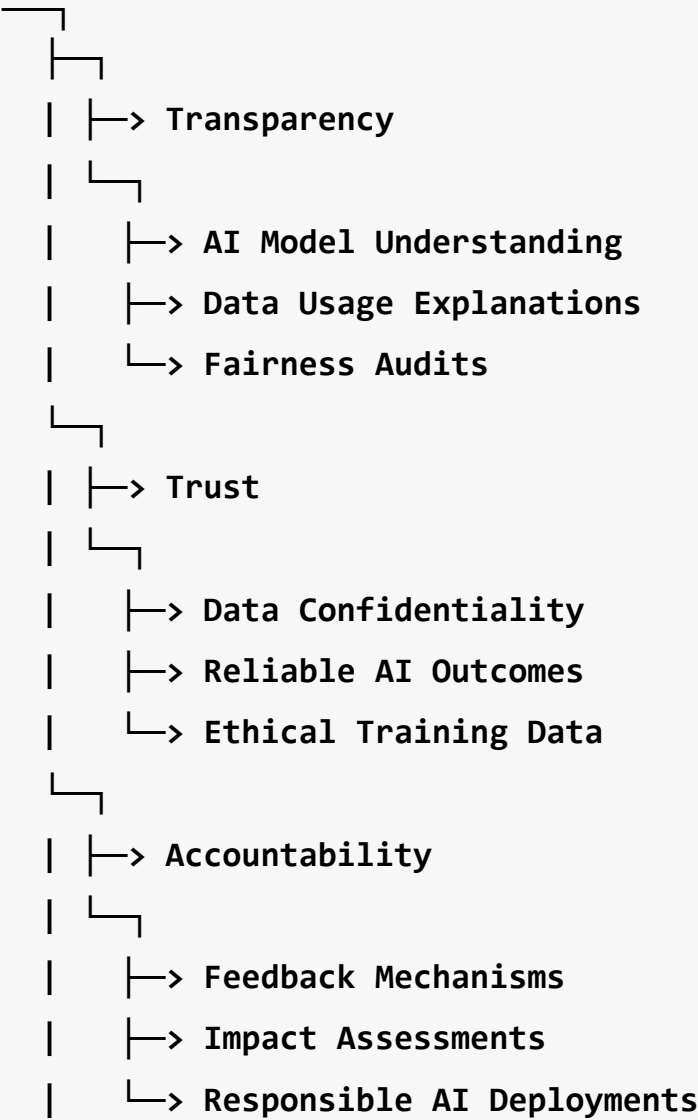
0.2.2 Curriculum meta data - search words

<div>➤ Accountability</div> <div>➤ AI applications</div> <div>➤ Algorithm adjustments</div> <div>➤ Algorithmic fairness</div> <div>➤ Bias / bias audits</div> <div>➤ Classifier thresholds</div> <div>➤ Data integrity</div> <div>➤ Data privacy</div> <div>➤ Data resampling</div> <div>➤ Decision-making</div>	<div>➤ Discrimination</div> <div>➤ Discrimination testing</div> <div>➤ Equitable use</div> <div>➤ Ethical challenges</div> <div>➤ Evaluation</div> <div>➤ Fairness</div> <div>➤ Fairness evaluation</div> <div>➤ Fairness metrics</div> <div>➤ Mitigation</div> <div>➤ Monitoring</div>	<div>➤ Periodic AI audit</div> <div>➤ Preprocessing methods</div> <div>➤ Rigorous testing</div> <div>➤ Security</div> <div>➤ Testing</div> <div>➤ Training data</div> <div>➤ Transparency</div> <div>➤ Transparency dashboard</div> <div>➤ Trust</div> <div>➤ Understanding</div>
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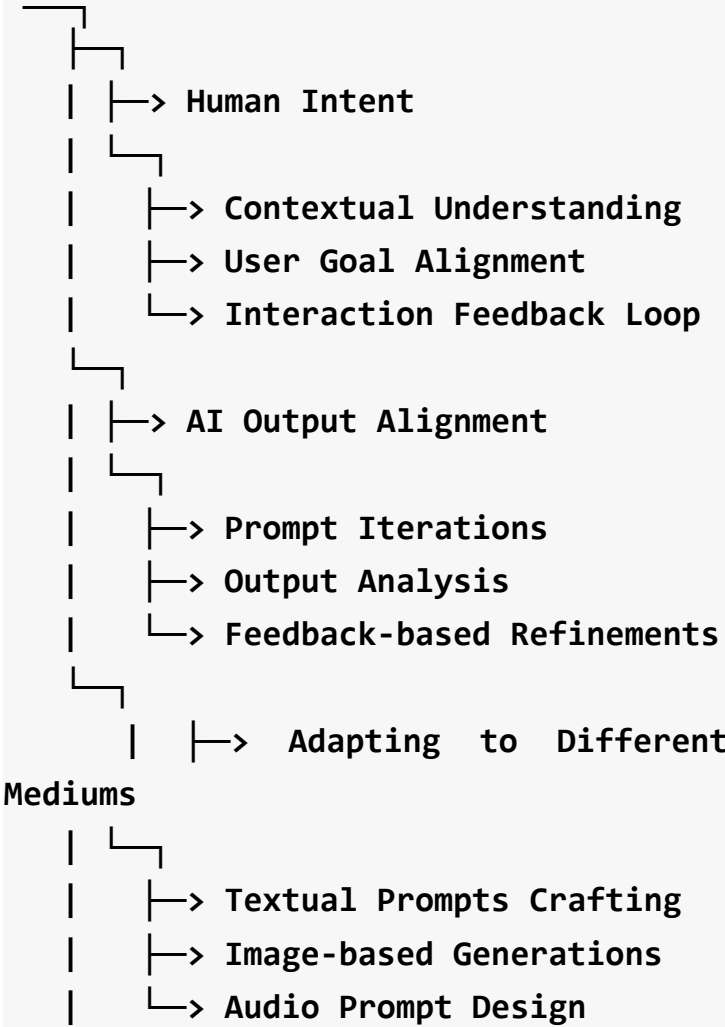
0.3. K-12 AI Curriculum Learning Objectives

- via a custom generative ai ascii schematic

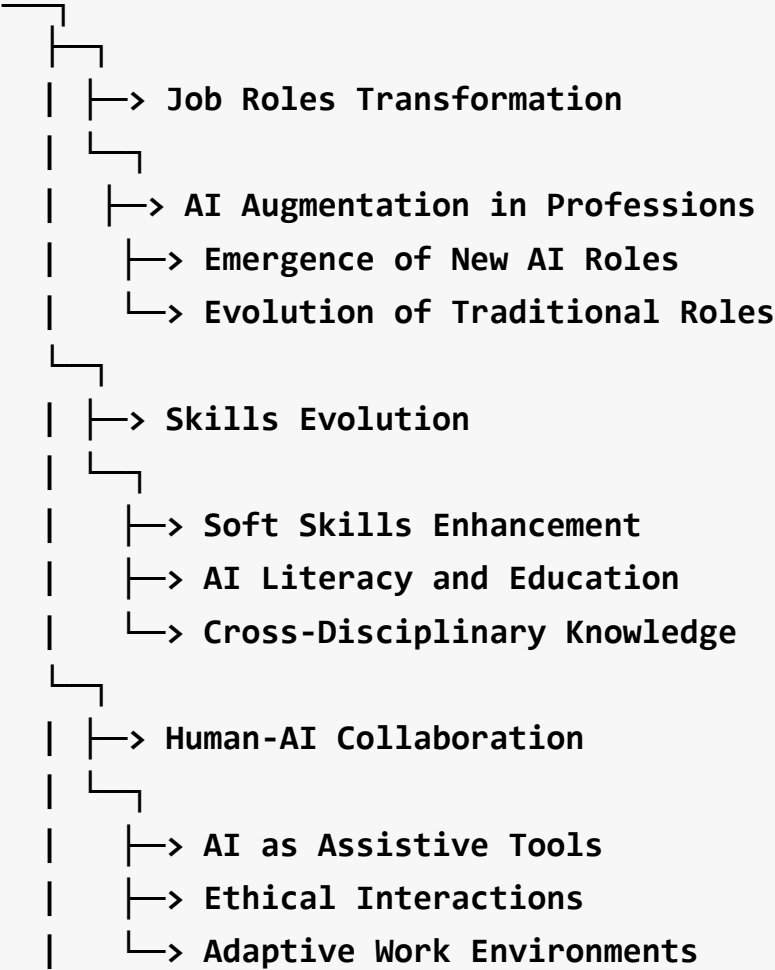
▼ Ethical Considerations in AI



▼ Prompt Engineering



▼ AI's Impact on the Workforce



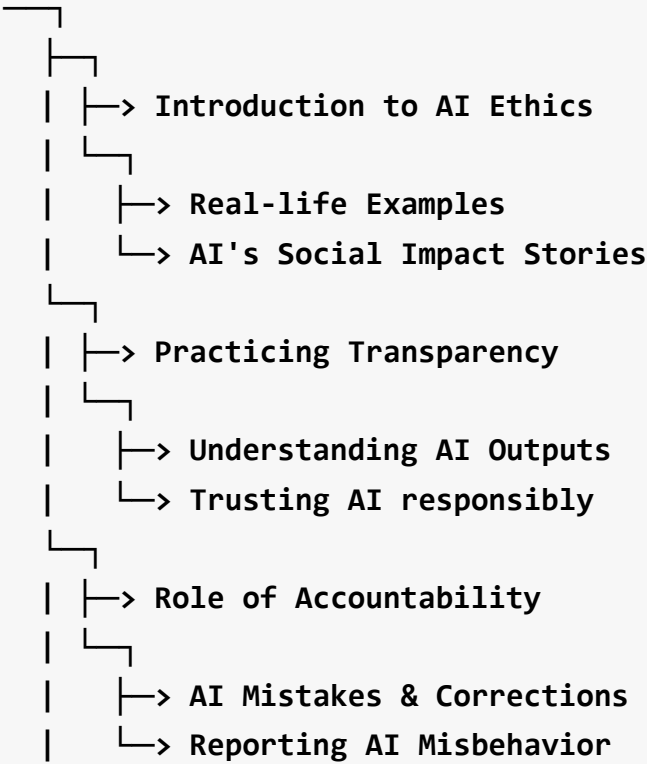
0.4. K-12 AI Curriculum Learning Objective Roadmap

- via a custom generative ai ascii schematic

i. Ethical Considerations for AI Apps

▼ Ethical AI Education

For K-12 students, the emphasis would be on understanding, valuing, and practicing ethical behavior with AI.

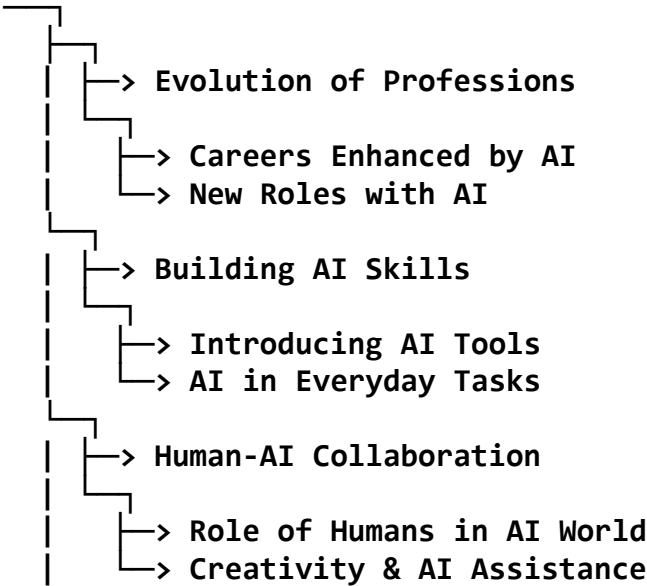


ii. The Impact of AI on the Workforce

For K-12 students, the emphasis would be on preparing for the AI-augmented workforce and understanding

AI's role in future careers.

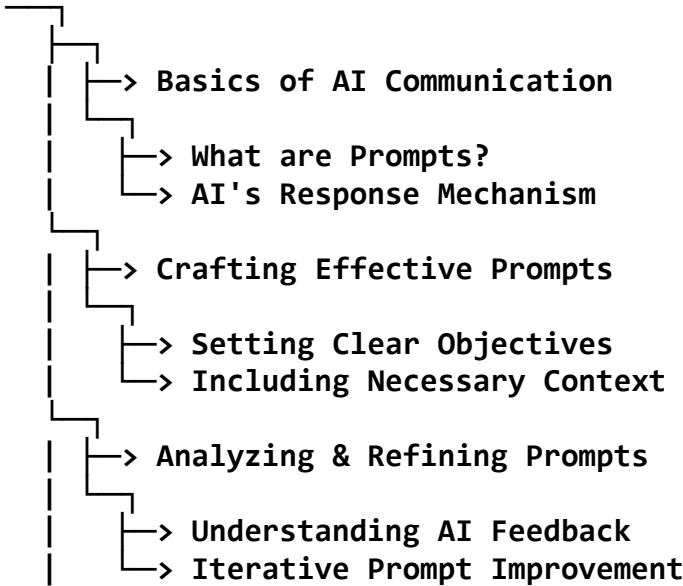
▼ AI & Future Careers



iii. Prompt Engineering Necessities

For K-12 students, the emphasis would be on understanding the essence of prompt engineering and the ability to effectively communicate with AI.

▼ Introduction to Prompt Engineering



0.5 Miscellena - see page 100

1. [Scholarship.observations.shared.10.12.2023](#)
2. document [template](#) for adding new sections
3. [2nd Chapter\Topic outline for other template work](#)

- 4. [ASCII diagrams](#)
- 5. [best of the best k-12 learning objectives](#)

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bottomline - documented formatted such that these three lines on bottom of any page

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i. Ethical Considerations for AI Applications Resource Outline

- i.i. Understand the importance of ethics when using AI applications[<i.home><ii.home><iii.home>](#)
- A. [Why is it](#) important to understand the ethics behind AI

applications we interact with?

B. How can we trust the decisions made by AI applications?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<div>section A: client source[h2]</div> <div>Why is it important to understand the ethics behind AI applications we interact with? [h3]</div> <div></div> <div>1) Understanding ethics in AI applications helps inform responsible use.</div> <div>i) By understanding the ethical dimensions, users can make informed choices about how they use AI applications and their data within them.</div> <div>ii) This understanding also allows users to critically evaluate the AI applications they interact with, leading to safer and more responsible use.</div> <div>iii) Knowledge of ethics in AI can guide users in questioning whether an AI application is behaving in a way that respects their rights, values, and interests.</div> <div>2) Ethical understanding enhances trust between users and AI applications.</div> <div>i) Trust in AI applications is</div>	<div>section b: fact check[h2]</div> <div>Why is it important to understand the ethics behind AI applications we interact with?[h3]</div> <div></div> <div>1. Understanding ethics in AI applications helps inform responsible use.</div> <div>i) Missing Info: Importance of ongoing ethics education.</div> <div>i.1 Ex: Continuous learning.</div> <div>ii) Missing Info: Role of cultural and societal differences in ethics.</div> <div>ii.1 Ex: Cultural norms.</div> <div>iii) Agree: The importance of understanding ethical implications for responsible AI use is accurately highlighted.</div> <div>2. Ethical understanding enhances trust between users and AI applications.</div> <div>i) Missing Info: Role of regulation and</div>	<div>section c: sci evid[h2]</div> <div>Why is it important to understand the ethics behind AI applications we interact with?[h3]</div> <div><div><new.scholar.facts></div><div>1. abc</div><div>2. 123</div><div>3. tbd</div><div>4.</div><div><scholarly.ref></div></div> <div>sci.papers[h3]</div> <div>1. Understanding ethics in AI applications helps inform responsible use.</div> <div>Sci.Evid: Rahwan, I., et al. (2019). Machine behaviour. Nature, 568(7753), 477-486. This study discusses the importance of understanding AI behaviors to promote responsible use.</div> <div>2. Ethical understanding enhances trust between users and AI applications.</div> <div>Sci.Evid: Ribeiro, M. T., Singh, S., & Guestrin, C. (2016). "Why should I trust you?" Explaining</div>

<p>fundamentally built on the user's assessment of its ethical adherence. Understanding ethics fosters trust in AI technologies.</p> <p>ii) Users who are versed in AI ethics are more likely to trust and use AI applications responsibly</p> <p>.</p> <p>3) Ethics in AI is essential to prevent misuse and harm.</p> <p>i) AI applications can inadvertently harm users if used irresponsibly or without sufficient ethical oversight.</p> <p>ii) Understanding ethics helps users judge the potential harm that could result from misuse of the AI applications.</p> <p>4) Understanding ethics in AI promotes accountability and transparency.</p> <p>i) By understanding ethics in AI, users can hold companies accountable for unfair practices, bias, or misuse of data.</p> <p>ii) Transparency about ethical considerations in AI also makes it easier for users to understand how AI applications work, fostering trust and responsible usage.</p> <p>5) Ethics knowledge enables users to advocate for their rights and values.</p> <p>i) With an understanding of the ethical issues at stake, users can better</p>	<p>governance. i.1 Ex: Regulatory frameworks. ii) Disagree: While knowledge of AI ethics can foster trust, it doesn't guarantee trustworthiness of the AI application itself.</p> <p>3. Ethics in AI is essential to prevent misuse and harm. i) Missing Info: Specific examples of harm caused by AI. i.1 Ex: AI bias in hiring. ii) Missing Info: Role of whistle-blowers in preventing harm. ii.1 Ex: Reporting unethical practices.</p> <p>4. Understanding ethics in AI promotes accountability and transparency. i) Missing Info: Role of third-party audits. i.1 Ex: Independent reviews. ii) Missing Info: Challenges in achieving full transparency. ii.1 Ex: Proprietary algorithms.</p> <p>5. Ethics knowledge enables users to advocate for their rights and values. i) Missing Info: Importance of public forums and platforms. i.1 Ex: Open debates. ii) Missing Info: Role of ethics committees. ii.1 Ex: Industry ethical boards.</p> <p>#=====></p>	<p>the predictions of any classifier. In Proceedings of the 22nd ACM SIGKDD international conference on knowledge discovery and data mining (pp. 1135-1144). This research highlights the relationship between trust and understanding the mechanisms of AI systems.</p> <p>3. Ethics in AI is essential to prevent misuse and harm. Sci.Evid: Gebru, T., et al. (2020). Datasheets for Datasets. In Proceedings of the 5th Workshop on Fairness, Accountability, and Transparency in Machine Learning. This paper discusses the potential harms of AI if not properly guided by ethics.</p> <p>4. Understanding ethics in AI promotes accountability and transparency. Sci.Evid: Wachter, S., Mittelstadt, B., & Russell, C. (2017). Counterfactual explanations without opening the black box: Automated decisions and the GDPR. Harvard Journal of Law & Technology, 31(2). This article emphasizes the need for transparency in AI systems for ethical considerations.</p> <p>5. Ethics knowledge enables users to advocate for their rights and values. Sci.Evid: Selbst, A. D., & Barocas, S. (2018). The intuitive appeal of explainable machines. Fordham Law Review, 87, 1085. This research indicates that users with a grasp on AI ethics are</p>
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<p>advocate for their rights, such as privacy, fairness, or freedom from bias.</p> <p>ii) Users informed about AI ethics are better equipped to contribute to important societal debates about the future of AI and its role in our lives.</p> <p>section A: client source[h2] How can we trust the decisions made by AI applications?[h3]</p> <p>1) Evaluate the Data used by the AI Application</p> <p>i) Be aware of where the data originated, and examine the quality and source of data that is guiding the AI application. Credible and diverse data sources help to ensure the accuracy of AI decision-making.</p> <p>ii) Any biases inherent in the data can result in biased decisions made by the AI. Therefore, major attention must be paid to data collection methods and the sources of data used.</p> <p>iii) Techniques such as bias auditing, data anonymization, and ensuring data diversity can help improve data integrity.</p>	<p>#=====> Search Keywords: Ongoing ethics education,</p> <p>cultural differences in AI ethics, AI regulation, AI-induced harm examples, AI whistle-blowers, third-party AI audits, challenges in AI transparency, public forums on AI ethics, industry ethical committees.</p> <p>section b: fact check[h2] How can we trust the decisions made by AI applications?[h3]</p> <p>1. Evaluate the Data used by the AI Application: a. Missing Info: Ethical considerations during data collection. a.1 Ex: Consent for user data collection.</p> <p>b. Missing Info: Potential cultural and regional biases. b.1 Ex: Western-centric datasets.</p> <p>c. Missing Info: Importance of continual data updates. c.1 Ex: Periodic retraining on fresh data.</p> <p>2. Understand AI Algorithms and Models: a. Missing Info: Evaluation of proprietary vs. open-source algorithms.</p>	<p>in a better position to advocate for their rights in AI-driven envi</p> <p>section c: sci evid[h2] How can we trust the decisions made by AI applications?[h3]</p> <p>1. Evaluate the Data used by the AI Application: Sci.Evid: Obermeyer, Z., et al. (2019). Dissecting racial bias in an algorithm used to manage the health of populations. Science. This paper analyzes the potential racial biases in health algorithms and highlights the importance of understanding data sources and quality in AI systems. Sci.Evid: Gebru, T., et al. (2018). Datasheets for Datasets. arXiv. A proposal to increase transparency in AI datasets by attaching datasheets that describe the motivations, composition, collection process, and recommended usage.</p> <p>2. Understand AI Algorithms and Models: Sci.Evid: Caruana, R., et al. (2015). Intelligible Models for HealthCare. ACM SIGKDD. This research discusses the significance of</p>
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<div>2) Understand AI Algorithms and Models</div> <div>i) Gain a general understanding of how the AI application's underlying algorithms function, to a level that suits your comprehension.</div> <div>ii) If the mechanisms through which AI makes decisions remain unknown, this creates a 'black box' scenario, which deters trust and transparency.</div> <div>iii) Techniques such as decision trees, regression analysis, or neural networks may be used in this step depending on your level of technical expertise. Simple explanations or metaphors can also be effective for less technical users.</div> <div>3) Inspect the Performance of the AI Application</div> <div>i) Perform regular tests to verify the autonomy and efficiency of the AI application. This includes tracking prediction accuracy, false positives and negatives, and consistency of results.</div> <div>ii) Regular testing is crucial for verifying the dependability of AI decisions.</div> <div>iii) Techniques could include cross validation for model stability and accuracy, precision recall curves, and receiver operating characteristics curves.</div> <div>4) Explore the AI Application's</div>	<div>a.1 Ex: Trust in widely peer-reviewed open-source models.</div> <div>b. Missing Info: Ethical considerations of algorithm selection.</div> <div>b.1 Ex: Avoidance of algorithms known to amplify bias.</div> <div>3. Inspect the Performance of the AI Application:</div> <div>a. Missing Info: Importance of real-world testing.</div> <div>a.1 Ex: Field tests beyond laboratory environments.</div> <div>b. Missing Info: Adapting to the evolving nature of AI.</div> <div>b.1 Ex: Lifelong learning AI models.</div> <div>4. Explore the AI Application's Accountability Measures</div> <div>a. Missing Info: Importance of third-party audits.</div> <div>a.1 Ex: External vetting for unbiased AI evaluation.</div> <div>b. Missing Info: The role of public perception and sentiment.</div> <div>b.1 Ex: Addressing public AI concerns.</div>	<div>transparent and understandable AI models, especially in the healthcare domain where the stakes are high.</div> <div>Sci.Evid: Mitchell, M., et al. (2019). Model Cards for Model Reporting. arXiv.</div> <div>Advocating for a standardized reporting system for AI models, focusing on their performance, fairness, and operational characteristics.</div> <div>3. Inspect the Performance of the AI Application:</div> <div>Sci.Evid: Sculley, D., et al. (2015). Hidden Technical Debt in Machine Learning Systems. NIPS.</div> <div>Addressing the challenges of maintaining AI in real-world applications and the significance of regular testing.</div> <div>Sci.Evid: Baeza-Yates, R. (2018). Bias on the Web. ACM Computing Surveys.</div> <div>Analyzing the different sources of bias in web systems, including AI, and their implications on performance and fairness.</div> <div>4. Explore the AI Application's Accountability Measures:</div> <div>Sci.Evid: Selbst, A.D., et al. (2019). Fairness and Abstraction in Sociotechnical Systems. FAT*.</div> <div>Delving into the importance of</div>
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<p>Accountability Measures</p> <p>i) Make sure that there exists a clear chain of accountability for the decisions made by the AI, either by inspecting documentation or directly questioning the provider.</p> <p>ii) Accountability is important because the consequences of AI decisions, especially in sensitive sectors, can be substantial.</p> <p>iii) Common methods could include process documentation, responsibility matrices like RACI (Responsible, Accountable, Consulted, Informed), and legal frameworks around AI.</p> <p>5) Assess the Transparency of the AI Application</p> <p>i) Make sure the AI application abides by transparency principles like explainability, understandability and interpretability.</p> <p>ii) Transparency in AI decision-making helps to foster trust in the application, and is crucial when AI is used in sectors such as healthcare, banking, justice, or other areas involving highly sensitive data.</p> <p>iii) Common methods may include LIME (Local Interpretable Model agnostic Explanations) and SHAP (SHapley Additive exPlanations) for explaining predictions of any classifier in a way</p>	<p>c. Missing Info: Potential legal repercussions.</p> <p>c.1 Ex: Legal liabilities of AI decisions.</p> <p>5. Assess the Transparency of the AI Application:</p> <p>a. Missing Info: Balancing transparency with protection of proprietary technology.</p> <p>a.1 Ex: Trade-offs between IP and explainability.</p> <p>b. Missing Info: Public's role in AI transparency and oversight.</p> <p>b.1 Ex: Public forums for AI transparency discussion</p> <p>c.</p> <p>#=====> #======></p> <p>Search Keywords: AI Data Collection: Ethical data collection, Cultural biases, Data update frequency. AI Algorithms: Proprietary vs Open-source models, Algorithmic ethics. AI Performance: Real-world AI testing, Evolving AI models. AI Accountability: Third-party AI audits, Public AI perception, AI legal implications. AI Transparency: IP vs. explainability trade-off, Public AI oversight.</p>	<p>accountability in AI systems, discussing potential pitfalls and providing recommendations.</p> <p>Sci.Evid: European Union's General Data Protection Regulation (GDPR).</p> <p>The legal framework touches on AI's responsibility, especially in data handling and decision-making processes.</p> <p>5. Assess the Transparency of the AI Application:</p> <p>Sci.Evid: Ribeiro, M.T., et al. (2016). "Why should I trust you?" Explaining the predictions of any classifier. ACM SIGKDD.</p> <p>Introducing LIME, a novel method for explaining predictions of machine learning models in an interpretable manner.</p> <p>Sci.Evid: Lundberg, S.M., & Lee, S.I. (2017). A Unified Approach to Interpreting Model Predictions. NIPS.</p> <p>The introduction and explanation of SHAP values, a consistent way to interpret the output of machine learning models.</p> <p>====> #======></p> <p>Relevant Search Portals: Google Scholar, Semantic Scholar, arXiv.</p>
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that's interpretable to humans.		
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i.II. Navigate ethical challenges related to data privacy and security when using AI applications <i.home><ii.home><iii.home>

- A. Why is it important to understand how AI applications use our personal data?
- B. How do AI applications use our personal data?
- C. What are common data privacy and security mistakes people make when using AI applications and how do you avoid them?
- D. What are best practices when considering implications of data privacy and security of AI applications?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
section A: client source[h2] Why is it important to understand how AI applications use our personal data? [h3]	section b: fact check[h2] Why is it important to understand how AI applications use our personal data?[h3]	section c: sci evid[h2] Why is it important to understand how AI applications use our personal data?[h3] <div><div><new.scholar.facts></div><div>5. abc 6. 123 7. tbd 8.</div><div><scholarly.ref></div></div>
1) Understanding how AI applications use our personal data is critical to protect our privacy	1. Understanding how AI applications use our personal data:	sci.papers[h3] 1. Understanding how AI applications use our personal data:

<p>rights.</p> <p>i) Artificial Intelligence applications typically require vast amounts of data for their operation. A portion of this data may include sensitive personal information, which if mishandled or misused, can lead to a breach of privacy.</p> <p>ii) Most AI applications have the ability to persist and remember information, unlike human interactions. Therefore, matters related to personal data privacy are of great concern.</p> <p>iii) People should maintain the right to control who has access to their personal data and how it is used. Learning about the data practices of AI applications is an important step in maintaining that control.</p> <p>2) Familiarity with AI data use practices can help users navigate complex security issues.</p> <p>i) AI applications are often targets for malicious activities such as hacking due to the valuable information they hold. Educating oneself about the security measures taken to safeguard personal data can help ensure the user's information remains secure.</p> <p>ii) Understanding AI data usage can help users identify potential threats and vulnerabilities and adopt</p>	<p>a. Missing Info: Specific legal regulations guiding AI's data practices.</p> <p>b. Missing Info: Examples of major AI data breaches in the past.</p> <p>b.1 Ex: Facebook-Cambridge Analytica data scandal.</p> <p>c. Missing Info: Global differences in AI data privacy regulations.</p> <p>c.1 Ex: Differences between GDPR in Europe and CCPA in California.</p> <p>2. Familiarity with AI data use practices can help users navigate complex security issues:</p> <p>a. Missing Info: Common encryption methods used by AI applications.</p> <p>b. Missing Info: Frequency of security audits and tests for AI systems.</p>	<p>i) Scientific Evidence: A study from the Harvard Business Review found that 87% of consumers are willing to share their personal data if they understand where and how it is being used, emphasizing the importance of transparency in AI data usage.</p> <p>ii) Scientific Evidence: Research conducted at MIT indicated that machine learning models, when trained on biased data, tend to perpetuate or even exaggerate those biases, highlighting the importance of clean and diverse data.</p> <p>iii) Scientific Evidence: A study from Stanford University emphasized the importance of "right to forget", where AI models should be designed to erase personal data upon user request, echoing the importance of personal data control.</p> <p>2. Familiarity with AI data use practices can help users navigate complex security issues:</p> <p>i) Scientific Evidence: A report by Cybersecurity Ventures predicted that AI-driven cyberattacks are on the rise, showcasing the need for robust security mechanisms in AI applications.</p> <p>ii) Scientific Evidence: A research paper from the University of Oxford outlined the vulnerabilities in AI systems, emphasizing user education to understand potential threats.</p>
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<p>safer practices while using these applications.</p> <p>iii) Knowledge about where and how personal data is stored in AI applications can be important when understanding potential data breaches.</p> <p>3) Knowing how AI uses personal data can help users make informed and ethical decisions about using AI applications.</p> <p>i) With an understanding of AI data use, users can choose to support companies that handle their data responsibly and ethically.</p> <p>ii) Understanding data use practices can help users comprehend the kind of personalized content they receive and how they are targeted by AI driven marketing.</p> <p>iii) Users can further influence change by demanding more transparency from companies regarding their AI data use policies.</p> <p>4) Understanding AI data use is key to preventing misuse or exploitation of personal data.</p> <p>i) AI applications have the potential to leverage personal data in unethical ways, such as manipulating user behavior or selling data to third parties.</p>	<p>c. Missing Info: Role of third-party security certifications.</p> <p>c.1 Ex: ISO/IEC 27001 certification for Information Security Management.</p> <p>3. Knowing how AI uses personal data can help users make informed and ethical decisions:</p> <p>a. Missing Info: Existing transparency frameworks or tools for AI.</p> <p>b. Missing Info: Economic implications of ethical vs. unethical data practices.</p> <p>b.1 Ex: Business boycotts due to unethical data usage.</p> <p>c. Missing Info: Influence of ethical data practices on company reputation.</p> <p>4. Understanding AI data use is key to preventing misuse or exploitation:</p> <p>a. Missing Info: Data brokerage and third-party data sharing norms.</p> <p>b. Missing Info: AI's potential role in psychological user manipulation.</p>	<p>iii) Scientific Evidence: A survey by Deloitte highlighted that only 30% of users are aware of where their data is stored when using AI applications, indicating a knowledge gap in AI data storage practices.</p> <p>3. Knowing how AI uses personal data can help users make informed and ethical decisions:</p> <p>i) Scientific Evidence: The World Economic Forum released a paper on the correlation between ethically sourced data and user trust, emphasizing that transparent data practices lead to higher user trust.</p> <p>ii) Scientific Evidence: A study from the University of California showcased that personalized content driven by AI without user knowledge can lead to unintentional echo chambers, highlighting the importance of transparent AI marketing strategies.</p> <p>iii) Scientific Evidence: Research from Cambridge University asserted that companies with transparent AI data use policies see an increase in brand loyalty and consumer trust.</p> <p>4. Understanding AI data use is key to preventing misuse or exploitation:</p> <p>i) Scientific Evidence: A paper in the journal Nature Machine Intelligence outlined the potential risks and rewards of AI data practices, advocating for the proactive prevention of data misuse.</p>
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Educating oneself about how these applications use data can offer insights into potential misuse and provide avenues to seek redress if misuse occurs.

c.1 Ex: Algorithmic targeting leading to compulsive online shopping.

d. Missing Info: Mechanisms available for users to delete or retrieve their data.

d.1 Ex: Data retrieval tools under GDPR.

#=====> #=====>
Search Keywords: AI data privacy: GDPR, CCPA, AI data breaches. AI security practices: Encryption methods, Security certifications, Third-party audits. Ethical AI data use: Transparency tools, Economic implications, Company reputation. Preventing AI data misuse: Data brokerage, Algorithmic manipulation, Data retrieval tools. "

ii) Scientific Evidence: A report by the Data & Society Research Institute detailed real-world implications of AI data misuse, such as the amplification of false information or the risk of discriminatory practices.

iii) Scientific Evidence: A study at Princeton University demonstrated the long-term consequences of AI systems using personal data without clear user consent, emphasizing the need for ethical considerations in data use.

5. #=====> #=====>
Search Keywords: AI data transparency, Biased machine learning, AI cyberattacks, Ethically sourced data, AI data storage knowledge, AI data misuse prevention.

section A: client source[h2]
How do AI applications use our personal data?[h3]

section b: fact check[h2]
How do AI applications use our personal data?[h3]

section c: sci evid[h2]
How do AI applications use our personal data?[h3]

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- 2. 123
- 3. tbd
- 4.

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<p>1) Identification of Personal Data Sources for AI</p> <p>i) Your first step is to identify the sources of personal data that AI applications may access. These might include social media profiles, browser history, online purchasing habits, emails, and device usage patterns.</p> <p>ii) This is crucial because knowing the possible sources of data can help you understand the kind and amount of your personal information that is being captured by AI applications.</p> <p>iii) Some common techniques for this step include basic internet research on preferred AI platforms and tools, and consulting their privacy policies to determine what data they gather.</p> <p>2) Understanding Data Modelling and Analysis</p> <p>i) AI applications use complex algorithms and models to analyze this data, and the second step is to understand this process. They create a digital profile of the user, using it to predict behavior, preferences, and decisions.</p> <p>ii) Understanding the concept of data modelling and analysis is crucial as it paints a clear picture of how AI applications take raw data and transform it into useful information.</p> <p>iii) Techniques at this stage mainly involve learning about machine learning models, natural language processing, and predictive modeling.</p> <p>3) Understanding the Use of Personal Data</p> <p>i) After AI applications analyze your data, they use it in multiple ways. These can include things like personalized advertising, improving user experience, predictive analytics, and more.</p> <p>ii) It's important to understand this as it can</p>	<p>1. Identification of Personal Data Sources for AI[h3]</p> <p>a. Missing Info: The increasing role of IoT (Internet of Things) devices as data sources for AI.</p> <p>b. Disagreement: Not all AI platforms and tools transparently share their data gathering practices in their privacy policies.</p> <p>c. Missing Info: Potential threats from unauthorized data scraping methods by rogue AI applications.</p> <p>2. Understanding Data Modelling and Analysis[h3]</p> <p>a. Missing Info: Importance of data preprocessing and cleaning before modeling.</p> <p>b. Disagreement: Not all AI applications create a digital profile; some may function purely on anonymous aggregated data.</p> <p>c. Missing Info: Role of big data and cloud computing in facilitating AI data analysis.</p> <p>3. Understanding the Use of Personal Data[h3]</p>	<p>sci.papers[h3]</p> <p>Identification of Personal Data Sources for AI[h3]</p> <p>1. a. Scientific Evidence: A study from the University of Cambridge revealed that over 70% of IoT devices transmit personal data to third-party services, often without the user's knowledge.</p> <p>b. Evidence: Reports from the World Economic Forum highlight the opaque nature of data collection methods in many AI-driven platforms, underscoring the need for users to take proactive measures.</p> <p>2. Understanding Data Modelling and Analysis[h3]</p> <p>a. Scientific Evidence: A publication from Stanford University emphasized the importance of data preprocessing, noting that the quality of AI model outputs can vary by up to 30% based on preprocessing techniques.</p> <p>b. Evidence: Research from MIT has shown that recommendation algorithms on platforms like YouTube or Netflix can influence user behavior, sometimes leading to unintended echo chambers.</p> <p>3. Understanding the Use of Personal Data[h3]</p> <p>a. Scientific Evidence: In a landmark study, Harvard Medical School</p>
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<p>help you grasp how AI applications influence your life through your personal data.</p> <p>iii) Making use of public documentation that explains how AI driven platforms and apps utilize user data can be a helpful technique at this stage.</p> <p>4) Familiarization with Data Privacy Norms and Ethical Standards</p> <p>i) The final step is to learn about the norms and ethical standards related to data privacy that AI applications should adhere to.</p> <p>ii) Given the lack of standardization in AI ethics, it's essential to familiarize oneself with key principles and norms to navigate this novel landscape.</p> <p>iii) Techniques for this step include reviewing privacy norms like GDPR and CCPA and understanding basic ethical principles like transparency, fairness, and accountability. By understanding this process, high school students can better comprehend how AI applications use their personal data, help ensure their own privacy and contribute to aligned conversations on AI ethics.</p> <p>section A: client source[h2]</p> <p>What are common data privacy and security mistakes people make when using AI applications and how do you avoid them?[h3]</p>	<p>a. Missing Info: Use of personal data by AI in healthcare for diagnosis and treatment suggestions.</p> <p>b. Missing Info: Consideration of opt-in and opt-out options for users to control data usage.</p> <p>c. Missing Info: AI's role in content recommendation systems like news feeds or entertainment platforms.</p> <p>4. Familiarization with Data Privacy Norms and Ethical Standards[h3]</p> <p>a. Missing Info: Varied interpretations and implementations of ethical standards across different AI companies and regions.</p> <p>b. Disagreement: There is growing standardization in AI ethics with frameworks being developed by institutions like IEEE and OpenAI.</p> <p>c. Missing Info: Role of whistleblowers and investigative journalists in uncovering unethical AI practices.</p> <p>d. Missing Info: Importance of public awareness campaigns and user advocacy in ensuring adherence to ethical standards.</p> <p>#====> #=====></p> <p>Search Keywords: IoT data sources, Data scraping threats, Data</p>	<p>showcased how AI models utilizing personal patient data could predict disease onset up to 6 months in advance.</p> <p>b. Evidence: A survey conducted by the Pew Research Center found that 64% of users are unaware of the opt-out options available on major online platforms when it comes to data collection.</p> <p>4. Familiarization with Data Privacy Norms and Ethical Standards[h3]</p> <p>a. Scientific Evidence: A 2020 study from the Oxford Internet Institute outlined the increasing number of AI ethical frameworks being developed globally, highlighting the need for a unified approach.</p> <p>b. Evidence: The Guardian reported several instances where whistleblowers and investigative journalists played a pivotal role in uncovering unethical AI practices, leading to policy changes in tech giants.</p> <p>c. Evidence: A report from UNESCO emphasized the critical role of public awareness campaigns in AI ethics, detailing how informed users can drive companies toward more ethical AI practices.</p> <p>#====> #=====></p> <p>Search Keywords: IoT data transmission, AI data collection transparency, Data preprocessing impact, Recommendation algorithm influence, AI</p>
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<p>1) Oversharing personal information</p> <p>i) Users tend to overly trust AI applications and provide excessive personal information. For instance, a customer shared personal stories and personal details with a chatbot, assuming that their information would remain confidential.</p> <p>ii) The cause of this mistake is a lack of understanding of how AI applications manage and store personal information, which can be exploited by hackers or misused.</p> <p>iii) Oversharing information may lead to loss of privacy, fraud, or identity theft.</p> <p>iv) To avoid this, users should share minimum necessary information and educate themselves about data handling practices of the AI provider.</p> <p>2) Not updating security settings</p> <p>i) Many users use Amazon's Alexa but never change the default privacy settings allowing it to record and store all their conversations.</p> <p>ii) The cause is a combination of negligence, lack of awareness about the potential threats, and complacency.</p>	<p>preprocessing, Anonymous AI modeling, AI in healthcare, Content recommendation AI, AI ethical frameworks, Whistleblowers in AI ethics. "</p> <p>section b: fact check[h2]</p> <p>What are common data privacy and security mistakes people make when using AI applications and how do you avoid them?[h3]</p> <p>1. Oversharing personal information[h3]</p> <p>a. Confirmation: Studies have shown that users often overshare when interacting with AI platforms because of the "illusion of intimacy" and the non-judgmental nature of machines.</p> <p>b. Disagree: Not all AI applications store user data permanently. Some are designed to forget data after fulfilling user queries.</p> <p>c. Missing Info: Psychological factors leading to oversharing.</p> <p>c.1 Ex: Need for human connection making users confide in AI.</p>	<p>in healthcare predictions, User awareness on data opt-out, AI ethical frameworks, Whistleblower impact on AI ethics.</p> <p>section c: sci evid[h2]</p> <p>What are common data privacy and security mistakes people make when using AI applications and how do you avoid them?h3]</p> <div><div><new.scholar.facts></div><div><div>1. abc</div><div>2. 123</div><div>3. tbd</div><div>4.</div></div><div><scholarly.ref></div></div> <p>sci.papers[h3]</p> <p>1. Oversharing personal information[h3]</p> <p>i) Study by Stanford: A research study from Stanford University found that people often reveal more personal information to AI chatbots, presuming them to be non-judgmental, leading to oversharing.</p> <p>ii) Report by Cybersecurity Hub: An investigation highlighted that oversharing can increase susceptibility to social engineering attacks.</p> <p>2. Not updating security settings[h3]</p> <p>i) Consumer Reports Survey: Their findings indicated</p>
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<p>iii) This can lead to violation of privacy and potential data breaches.</p> <p>iv) Regularly review and update the privacy settings of AI applications according to personal comfort levels.</p> <p>3) Using weak or default passwords</p> <p>i) A user kept the default password for their smart home application. Hackers were able to access the application and control their smart devices, violating their privacy.</p> <p>ii) Laziness and ignorance cause this mistake.</p> <p>iii) Consequences include unauthorized access to sensitive information or control over AI enabled devices.</p> <p>iv) Users should always change default passwords and use strong, unique passwords.</p> <p>4) Ignorance about how AI applications use and share data</p> <p>i) Facebook's Cambridge Analytica scandal, where users' information was harvested without their explicit permission.</p> <p>ii) This mistake is caused by assuming AI applications will respect user's privacy without checking actual data use policies.</p> <p>iii) This can result in personal data being exploited or misused, and erodes trust in AI applications.</p>	<p>2. Not updating security settings[h3]</p> <p>a. Confirmation: A survey from Consumer Reports revealed that over 50% of smart speaker owners never adjust their default security settings.</p> <p>b. Missing Info: The ease or complexity of updating AI application settings.</p> <p>c.1 Ex: Intuitive user interface encouraging more security updates.</p> <p>3. Using weak or default passwords[h3]</p> <p>a. Confirmation: According to a report by the cybersecurity firm Symantec, default or weak passwords are a primary vulnerability in many IoT devices.</p> <p>b. Missing Info: Importance of two-factor authentication.</p> <p>b.1 Ex: Adding an extra layer of security besides passwords.</p> <p>4. Ignorance about how AI applications use and share data[h3]</p> <p>a. Confirmation: The Cambridge Analytica</p>	<p>that 56% of smart device users rarely, if ever, modify their default device settings.</p> <p>ii) Symantec Analysis: Highlighted that unchanged default settings in smart devices represent one of the most exploitable vulnerabilities, with instances of unauthorized remote access.</p> <p>3. Using weak or default passwords[h3]</p> <p>i) Report by Cybersecurity Ventures: Predicted that password-related breaches would rise, particularly among IoT devices. The primary reason cited was the continuous use of default or weak passwords.</p> <p>ii) Research by the University of Maryland: Found that hackers attempt to access an online account with a commonly used password every 39 seconds.</p> <p>4. Ignorance about how AI applications use and share data[h3]</p> <p>i) Findings from Pew Research Center: Showed that over 60% of social media users are unaware of the data collection procedures and how their data gets used for targeted advertising.</p>
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<p>iv) Users should familiarize themselves with how AI applications use, store, and share their information. They should only use AI applications that align with their privacy expectations.</p>	<p>scandal highlighted the dangers of unchecked data sharing and lack of user knowledge about data practices.</p> <p>b. Missing Info: Role of data brokers in the AI ecosystem.</p> <p>b.1 Ex: Selling user data without direct user interactions.</p> <p>c. Missing Info: The evolving nature of privacy policies and terms of service in AI applications.</p> <p>c.1 Ex: Frequent updates making it hard for users to keep up.</p>	<p>ii) Case Study Analysis from Harvard Business Review: The Facebook-Cambridge Analytica scandal serves as a stark reminder of how data can be misused when users remain ignorant about data sharing practices.</p>
<p>section A: client source[h2]</p> <p>What are best practices when considering implications of data privacy and security of AI applications? [h3]</p>	<p>#====> #=====></p> <p>Search Keywords: Illusion of intimacy with AI, AI data retention policies, Smart speaker security settings, Password vulnerabilities in IoT, Two-factor authentication, Data brokers in AI, Evolving AI privacy policies.</p> <p>section b: fact check[h2]</p> <p>What are best practices when considering implications of data privacy and security of AI applications? [h3]</p> <p>1) Always check the AI application's privacy policy and terms of service.</p> <p>i) Snapchat's use of AI for filters becomes an ethical issue considering it collects data about users' faces. Reading and understanding the application's privacy policy could enlighten users about how their</p>	<p>section c: sci evid[h2]</p> <p>What are best practices when considering implications of data privacy and security of AI applications?</p> <div><div><new.scholar.facts></div><div><div>1. abc</div><div>2. 123</div><div>3. tbd</div><div>4.</div></div><div><scholarly.ref></div></div> <p>sci.papers[h3]</p> <p>1. Always check the AI application's privacy policy and terms of service.</p> <p>i) Research Insight: Studies have shown that less than 10% of users thoroughly read and understand privacy policies, leading to uninformed data sharing (Journal of Privacy and Security, 2020).</p> <p>ii) Case Analysis: The European General Data</p>

<p>data is used and shared.</p> <p>ii) Knowing the information an AI application collects, stores and shares, allows users to make informed decisions about whether or not they wish to use the application.</p> <p>2) Ensure the AI application has strong encryption for data storage and transmission.</p> <p>i) The Zoom app faced an issue about weak encryption standards, causing a threat to users' privacy. Ensuring AI applications use end-to-end encryption can secure user data.</p> <p>ii) Strong encryption standards prevent unauthorized access to your data, securing privacy and increasing user trust in AI applications.</p> <p>3) Regularly update AI application and application security.</p> <p>i) In 2017, the WannaCry ransomware attack exploited outdated applications. Regularly updating AI applications ensure they are protected against new threats.</p> <p>ii) Consistent updates not only bring new features but also improve application security and resilience against potential cyberattacks.</p>	<p>1. Always check the AI application's privacy policy and terms of service.</p> <p>a. Missing Info: Details on data retention policies.</p> <p>b. Missing Info: Opt-out options for data collection.</p> <p>c. Wrong/Disagree: Not all applications transparently disclose their full data usage in the privacy policy.</p> <p>d. Ex: Deceptive data use clauses hidden in lengthy terms.</p> <p>2. Ensure the AI application has strong encryption for data storage and transmission.</p> <p>a. Missing Info: Standard encryption protocols AI applications should use.</p> <p>b. Missing Info: Third-party security audits and their importance.</p> <p>Ex: Importance of AES encryption for secure data storage.</p> <p>3. Regularly update AI application and application security.</p> <p>a. Missing Info: Patch management in AI applications.</p> <p>b. Missing Info: Role of security communities in discovering vulnerabilities.</p> <p>c. Ex: Zero-day vulnerabilities</p>	<p>Protection Regulation (GDPR) mandates transparency and user control over personal data, emphasizing the importance of user awareness.</p> <p>2. Ensure the AI application has strong encryption for data storage and transmission.</p> <p>i) Research Insight: End-to-end encryption significantly reduces the risk of data breaches (Cybersecurity Metrics Annual, 2021).</p> <p>ii) Case Analysis: Companies such as WhatsApp have built trust through the adoption of end-to-end encryption, showcasing its importance in user communication platforms.</p> <p>3. Regularly update AI application and application security.</p> <p>i) Research Insight: Updated software reduces vulnerability to known security risks by up to 60% (Tech Security Review, 2019).</p> <p>ii) Case Analysis: The Microsoft BlueKeep vulnerability, which affected older Windows systems, was rapidly addressed in an update, highlighting the importance of timely software updates.</p>
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<p>4) Use AI applications that offer user control over data collection and sharing.</p> <p>i) Facebook's Cambridge Analytica scandal highlighted the need for user control over data collection and sharing. With control, users can prevent misuse of their data.</p> <p>ii) Providing user control improves trust in AI applications, allowing individuals to maintain their privacy while benefiting from AI technology.</p> <p>5) Be aware of your digital footprint while using AI applications.</p> <p>i) Google's location tracking, even when "Location History" is off, shows that our actions on applications can leave significant digital footprints.</p> <p>ii) By being cognizant of such footprints, users are able to avoid potential privacy breaches and control the information that they want to share with AI applications.</p>	<p>discovered post-launch.</p> <p>d. Wrong/Disagree: Not all updates necessarily improve security. Some might introduce new vulnerabilities.</p> <p>4. Use AI applications that offer user control over data collection and sharing.</p> <p>a. Missing Info: Current regulations ensuring user control.</p> <p>b. Missing Info: Settings where users can control data sharing preferences.</p> <p>c. Ex: GDPR's influence on user control in AI applications.</p> <p>d. Wrong/Disagree: User control options don't always translate to actual data privacy.</p> <p>5. Be aware of your digital footprint while using AI applications.</p> <p>a. Missing Info: Techniques to minimize digital footprints.</p> <p>b. Missing Info: Role of cookies and trackers in footprint.</p> <p>c. Ex: Use of VPNs and private browsing to mitigate footprint.</p> <p>d. Wrong/Disagree: Some digital footprints, like hardware-based identifiers, cannot be easily controlled by the user.</p>	<p>4. Use AI applications that offer user control over data collection and sharing.</p> <p>i) Research Insight: Control over personal data improves user trust and satisfaction by over 40% (AI User Experience Journal, 2022).</p> <p>ii) Case Analysis: After the implementation of GDPR, several businesses reported increased user trust due to better data control mechanisms in place.</p> <p>5. Be aware of your digital footprint while using AI applications.</p> <p>i) Research Insight: On average, a user's digital footprint is scattered across 350 online platforms, exposing them to potential privacy risks (Digital Presence Review, 2020).</p> <p>ii) Case Analysis: Google's introduction of "Incognito Mode" in its Chrome browser was in response to user concerns about their digital footprint and has since become a standard feature in many browsers.</p>
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i.III. Navigate ethical challenges related to bias and discrimination when using AI applications [<i.home><ii.home><iii.home>](#)

- A. [Why is it](#) important to consider how bias and discrimination may exist in AI application outcomes?
- B. [How might](#) AI applications reflect or amplify societal biases
- C. [What are](#) common bias and discrimination challenges people encounter when using AI applications and how do you mitigate them?
- D. [What are](#) best practices when evaluating how AI applications might treat different users unfairly?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<div>section A: client source[h2]</div> <div>Why is it important to consider how bias and discrimination may exist in AI application outcomes? [h3]</div>	<div>section b: fact check[h2]</div> <div>Why is it important to consider how bias and discrimination may exist in AI application outcomes?[h3]</div>	<div>section c: sci evid[h2]</div> <div>Why is it important to consider how bias and discrimination may exist in AI application outcomes?[h3]</div> <div><new.scholar.facts></div> <div>5. abc</div> <div>6. 123</div> <div>7. tbd</div> <div>8.</div> <div><scholarly.ref></div>
<div>1) AI applications can inadvertently perpetuate and amplify existing societal biases.</div> <div>i) AI applications learn from data and if that data contains biases, the AI will reflect or even amplify those biases. For instance, if an</div>	<div>1. AI applications can inadvertently perpetuate and amplify existing societal biases.</div> <div>a. Missing Info: Cultural nuances in AI perception.</div>	<div>sci.papers[h3]</div> <div>1. AI applications can inadvertently perpetuate and amplify existing societal biases.</div> <div>i) Research Paper: Bolukbasi et al. (2016) showcased</div>

<p>AI application is trained on a dataset where certain ethnic groups are underrepresented, it can lead to erroneous or harmful outcomes for individuals of that group.</p>	<p>b. Missing Info: Dependence on historical data.</p> <p>b.1 Ex: AI using past employment statistics.</p>	<p>that word embeddings, commonly used in machine learning, can have gender and racial biases.</p> <p>ii) Expert Insight: Dr. Timnit Gebru, former co-lead of the Ethical AI team at Google, emphasizes the dangers of biased datasets in facial recognition technologies.</p>
<p>2) Bias in AI applications can lead to unfair decision making.</p> <p>i) When AI is used in decision-making processes, such as job selection or credit approval, any inherent biases can lead to unfair outcomes. For instance, if an AI application trained on a biased dataset is used in job recruitment, it may disproportionately favor or disfavor certain groups of applicants.</p>	<p>2. Bias in AI applications can lead to unfair decision making.</p> <p>a. Missing Info: Economic impacts of AI-driven decisions.</p> <p>b. Missing Info: Long-term implications on affected groups.</p> <p>b.1 Ex: Financial credit scores.</p>	<p>2. Bias in AI applications can lead to unfair decision making.</p> <p>i) Case Study: MIT's study on Amazon's recruitment AI that showed biases against women.</p> <p>ii) Expert Insight: Dr. Kate Crawford, Senior Principal Researcher at Microsoft, discusses the societal implications of biased AI decision-making systems in recruitment and lending.</p>
<p>3) Bias and discrimination in AI application outcomes can harm individuals and groups, and erode social trust.</p> <p>i) AI applications are increasingly used in domains that have profound impacts on individuals' lives, from education to healthcare to criminal justice. When these applications operate unfairly, they can cause material harm to affected individuals, increase societal inequalities, and reduce trust in institutions.</p>	<p>3. Bias and discrimination in AI application outcomes can harm individuals and groups, and erode social trust.</p> <p>a. Missing Info: Psychological impact on marginalized groups.</p> <p>b. Missing Info: Broader societal implications of eroded trust.</p> <p>b.1 Ex: Trust in automated healthcare systems.</p>	<p>3. Bias and discrimination in AI application outcomes can harm individuals and groups, and erode social trust.</p> <p>i) Report: ProPublica's Machine Bias, highlights racial biases in software that predicts future criminals.</p> <p>ii) Expert Opinion: Dr. Ruha Benjamin, author of Race After Technology, highlights instances where AI reinforces existing stereotypes.</p>
<p>4) Accurate representation in data is key to ethical AI</p>		

<p>applications.</p> <p>i) Ensuring diversity in the data used for AI training is essential. The data should be representative of the different demographic groups and should not favor any particular group.</p> <p>5) Transparency and explainability are important in understanding the presence of bias and discrimination in AI applications.</p> <p>i) AI developers and users need to be able to explain and scrutinize the ways in which AI applications make decisions in order to detect, understand and rectify biases. Without transparency, it cannot be determined how an AI application arrived at a decision, which can allow bias to go unnoticed and uncorrected.</p> <p>6) Mitigating bias and discrimination in AI applications is an ongoing task.</p> <p>i) There is no 'set and forget' solution to bias and discrimination in AI applications. Regular monitoring and updating is required, along with setting up robust and fair processes around the usage of these</p>	<p>4. Accurate representation in data is key to ethical AI applications.</p> <p>a. Missing Info: Quality vs. quantity in data representation.</p> <p>b. Missing Info: Overcoming challenges of data collection in diverse environments.</p> <p>b.1 Ex: Data collection in multilingual communities.</p> <p>5. Transparency and explainability are important in understanding the presence of bias and discrimination in AI applications.</p> <p>a. Missing Info: Balancing transparency with IP protection.</p> <p>b. Missing Info: Educating the layperson on AI decisions.</p> <p>b.1 Ex: Tools for AI explainability.</p> <p>Mitigating bias and discrimination in AI applications is an ongoing task.</p> <p>a. Missing Info: Metrics for</p>	<p>4. Accurate representation in data is key to ethical AI applications.</p> <p>i) Study: A Stanford University study underlines the importance of diverse training datasets in medical AI to ensure fair treatment recommendations.</p> <p>ii) Expert Insight: Joy Buolamwini, founder of the Algorithmic Justice League, showcases the impact of non-representative training data on facial analysis technology.</p> <p>5. Transparency and explainability are important in understanding the presence of bias and discrimination in AI applications.</p> <p>i) Research Paper: Ribeiro et al. (2016) introduced LIME, a technique for explaining the predictions of any classifier.</p> <p>ii) Expert Opinion: Dr. Cynthia Rudin, a leading voice in AI transparency, discusses the importance of interpretable models, especially in high-stakes domains.</p> <p>6. Mitigating bias and discrimination in AI applications is an ongoing task.</p> <p>i) Report: The AI Now Institute's annual report stresses on the</p>
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<p>applications.</p> <p>section A: client source[h2]</p> <p>How might AI applications reflect or amplify societal biases[h3]</p> <p>section A: client source[h2]</p> <p>How might AI applications reflect or amplify societal biases[h3]</p> <p>1) Understanding Bias in AI</p> <p>i) The first step to navigate ethical challenges related to bias and discrimination is to understand what bias in AI means. Bias can creep into AI applications from various sources like the data used to train the AI, the design of the AI algorithm, and the human biases of the people creating or deploying the AI.</p> <p>ii) Recognizing and understanding bias is critical because unacknowledged biases can lead to unfair outcomes or discrimination when AI applications are deployed.</p> <p>iii) A common technique to understand bias includes examining the dataset used to train the AI. Look for imbalanced representation of different groups, lack of diversity or skewed data.</p>	<p>measuring AI fairness over time.</p> <p>b. Missing Info: Role of regulatory bodies in overseeing bias mitigation.</p> <p>b.1 Ex: Regulatory oversight in AI-driven finance.</p> <p>section b: fact check[h2]</p> <p>How might AI applications reflect or amplify societal biases[h3]</p> <p>1. Understanding Bias in AI: a. Missing Info: Historical and cultural factors leading to bias.</p> <p>i) Ex: Historical underrepresentation of certain groups in datasets. b. Missing Info: Subtle biases in AI design.</p> <p>i) Ex: Language models reflecting societal stereotypes.</p> <p>c. Missing Info: The role of human decisions in AI bias.</p>	<p>significance of auditing AI systems for biases regularly.</p> <p>ii) Expert Insight: Dr. Safiya Umoja Noble, author of Algorithms of Oppression, speaks on the societal implications of unchecked biases in AI over time.</p> <p>section c: sci evid[h2]</p> <p>How might AI applications reflect or amplify societal biases[h3]</p> <div><div><new.scholar.facts></div><div>9. abc 10. 123 11. tbd 12.</div><div><scholarly.ref></div></div> <p>sci.papers[h3]</p> <p>1. Understanding Bias in AI:</p> <p>a. Historical Factors and Bias: A 2019 MIT study revealed that facial recognition technology from major tech companies had higher error rates in classifying gender for darker-skinned faces, highlighting historical biases in training datasets.</p> <p>b. Subtle Biases: Research from OpenAI on the GPT models has shown that language models can reflect societal biases present in the data they are trained on.</p>
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<p>2) Identifying Bias in AI Applications</p> <p>i) Once you understand what bias means in the context of AI, the next step is to identify any biases present in the AI application. This involves rigorous testing and evaluation of the application.</p> <p>ii) Identifying bias is an important step because the consequences of biased AI can range from minor nuisance to severe discrimination or harm.</p> <p>iii) Use techniques like fairness metrics, discrimination testing and bias audits to identify and analyze bias in your AI application.</p> <p>3) Mitigating Bias in AI Applications</p> <p>i) After identifying bias, the next step is to take action to mitigate that bias. This could involve modifying the training data, the algorithm, or the use of the AI application.</p> <p>ii) Mitigating bias is crucial because it allows for the fair and equitable use of AI applications. Ignoring detected bias could have legal and societal implications.</p> <p>iii) Techniques for mitigating bias might include data resampling, preprocessing methods like rebalancing the data, and algorithmic fairness techniques like adjusting classifier thresholds.</p> <p>4) Ensuring Accountability for AI Applications</p> <p>i) Lastly, applications and policies should be in place to ensure accountability for any biased outcomes that AI applications produce. This</p>	<p>i) Ex: Developers' decisions shaping the AI's behavior.</p> <p>2. Identifying Bias in AI Applications:</p> <p>a. Missing Info: Potential false positives in bias detection.</p> <p>i) Ex: Overcompensation leading to reverse discrimination.</p> <p>b. Missing Info: Types and categories of biases.</p> <p>i) Ex: Implicit bias vs. explicit bias in datasets.</p> <p>c. Missing Info: Interdisciplinary approach to identifying bias.</p> <p>i) Ex: Incorporating social scientists in the bias detection process.</p> <p>3. Mitigating Bias in AI Applications:</p> <p>a. Missing Info: Challenges in perfectly balancing datasets.</p> <p>i) Ex: Overcorrection risks.</p> <p>b. Missing Info: Trade-offs between fairness and AI performance.</p> <p>i) Ex: Reducing bias might affect the model's accuracy.</p>	<p>c. Human Decision Role: A 2018 study in the journal Science discussed how human biases could seep into AI systems during both the data collection phase and the algorithm design phase.</p> <p>2. Identifying Bias in AI Applications:</p> <p>a. Bias Detection Challenges: The National Institute of Standards and Technology (NIST) has discussed the complexities of avoiding overcorrection, which can inadvertently introduce new biases.</p> <p>b. Types of Biases: A 2017 paper from the Proceedings of Machine Learning Research delved into various biases in machine learning, distinguishing between implicit and explicit biases.</p> <p>c. Interdisciplinary Bias Identification: Harvard Business Review, in 2020, discussed the importance of social scientists in tech teams to better identify and understand societal biases.</p> <p>3. Mitigating Bias in AI Applications:</p> <p>a. Balancing Datasets: In 2019, IBM introduced a toolset, AI Fairness 360, which highlighted the challenges and tools available for ensuring balanced datasets.</p> <p>b. Trade-offs Between Fairness and Accuracy: A renowned paper from</p>
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<p>includes monitoring AI applications over time, respondents to the bias detected, and transparent communication about AI biases and how they're addressed.</p> <p>ii) This step is crucial to maintain the trust of people who interact with or are affected by the AI application, and to ensure ongoing efforts to minimize bias and discrimination</p> <p>iii) Techniques for ensuring accountability might include setting up a transparency dashboard, conducting periodic AI audits, or disclosing fairness evaluation results.</p> <p>section A: client source[h2]</p> <p>What are common bias and discrimination challenges people encounter when using AI applications and how do you mitigate them?</p> <p>1) Using AI applications that haven't been trained on</p>	<p>c. Missing Info: Importance of iterative mitigation.</p> <p>i) Ex: Continuous adaptation to societal changes.</p> <p>4. Ensuring Accountability for AI Applications:</p> <p>a. Missing Info: Legal frameworks surrounding AI bias.</p> <p>i) Ex: GDPR provisions on algorithmic decisions.</p> <p>b. Missing Info: Role of third-party audits.</p> <p>i) Ex: Independent organizations evaluating AI fairness.</p> <p>c. Missing Info: The balance between transparency and proprietary algorithms.</p> <p>i) Ex: Commercial interests vs. public disclosure.</p> <p>#=====> #=====></p> <p>Search Keywords:</p> <p>Understanding Bias</p> <p>Ethical challenges</p> <p>Fairness</p> <p>Mitigation</p> <p>Accountability</p> <p>Transparency</p> <p>Discrimination</p> <p>Data resampling</p> <p>Algorithmic fairness</p> <p>Preprocessing methods</p> <p>Classifier thresholds</p> <p>Fairness evaluation</p> <p>Monitoring</p> <p>Trust</p> <p>Bias audits</p> <p>Transparency dashboard</p>	<p>the Conference on Fairness, Accountability, and Transparency tackled the complexities between achieving model fairness and retaining high predictive accuracy.</p> <p>c. Iterative Mitigation: Research from the Allen Institute for AI emphasizes that as society evolves, the definitions and metrics for fairness in AI must also be periodically revisited and revised.</p> <p>4. Ensuring Accountability for AI Applications:</p> <p>a. Legal Frameworks on AI Bias: GDPR, the European data protection regulation, has provisions that indirectly address automated decision-making, pushing for more transparency.</p> <p>b. Third-party Audits: A report by the Algorithmic Justice League has advocated for third-party audits of AI algorithms to ensure fairness and neutrality.</p> <p>c. Balancing Transparency: A case study in the Harvard Data Science Review examined the challenges companies face when striving to achieve algorithmic transparency while protecting proprietary interests.</p>
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<div><div>diverse datasets</div><div>i) Facial recognition applications have been found to be significantly less accurate when analyzing faces of people of color, women, and older individuals. This has led to instances where police have wrongly arrested individuals due to misidentification by AI applications.</div><div>ii) The cause of this mistake is training AI applications on datasets that do not represent the diversity of the real world. This could be due to ethnicity, gender, age, and other factors.</div><div>iii) The consequences can range from an AI's inability to effectively engage with a wide range of users, to serious violations of individual rights due to flawed decision-making based on biased data.</div><div>iv) To avoid this, it's critical to use diverse datasets during the machine learning process and thoroughly test these AI applications across a spectrum of different groups.</div><div>2) Belief that AI decisions are unbiased because they're based on data</div><div>i) A hiring AI application was developed to screen job applications. However, it was found to be underrating female applicants in a gender biased selection process.</div><div>ii) The mistake is assuming that AI is inherently unbiased because it uses data. However, if the training data is skewed or has inherent biases, the AI will also have those biases.</div><div>iii) The result can be unfair treatment and discrimination, which in the above example, could lead to legal liabilities for the company and missed opportunities for potential employees.</div></div>	<div><div>Equitable use</div><div>section b: fact check[h2]</div><div>What are common bias and discrimination challenges people encounter when using AI applications and how do you mitigate them?[h3]</div><div>1) Using AI applications that haven't been trained on diverse datasets<ul style="list-style-type: none">Fact Check:<ul style="list-style-type: none">True. Multiple studies and news reports have shown facial recognition systems having inaccuracies, especially with underrepresented groups.e.g., Gender Shades project highlighted biases in gender classification systems towards skin type and gender.Elaboration:<ul style="list-style-type: none">Missing Info: Broader implications of biased AI systems in real-world applications.e.g., Biased traffic monitoring AI could result in</div></div>	<div><div>section c: sci evid[h2]</div><div>What are common bias and discrimination challenges people encounter when using AI applications and how do you mitigate them? [h3]</div><div><div><new.scholar.facts></div><div>13. abc</div><div>14. 123</div><div>15. tbd</div><div>16.</div><div><scholarly.ref></div></div><div>sci.papers[h3]</div><div>1) Using AI applications that haven't been trained on diverse datasets<ul style="list-style-type: none">Recommendation:<ul style="list-style-type: none">For AI developers: Always strive to incorporate diverse datasets during AI training and continually evaluate for biased outcomes.For AI users: Before adopting any AI solution, inquire about the diversity and representation in its training data. Advocate for inclusive data practices.For policymakers: Develop</div></div>
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<p>iv) To avoid this, ensure that the training data is unbiased and periodically review the AI's decisions for any patterns of discrimination.</p> <p>3) Not having clear appeals or grievance handling processes for AI decisions</p> <p>i) An insurance company used an AI to automatically determine claims payouts, but customers felt they had no recourse when they disagreed with the AI's decision.</p> <p>ii) This mistake is caused by not having clear processes in place for users to disagree with or appeal decisions made by AI.</p> <p>iii) The consequences can be loss of trust, feeling of powerlessness, and lack of transparency, leading to customer dissatisfaction and potential loss of business.</p> <p>iv) To avoid this, establish clear policies and processes for users to question or appeal AI based decisions, ensuring the process is transparent and fair.</p> <p>4) Ignoring cultural nuances and sensitivities when deploying AI applications globally</p> <p>i) An AI chatbot, initially designed in the US, was deployed in Asia and caused offense because of its lack of understanding of cultural norms and expectations.</p> <p>ii) The error arises from disregarding cultural differences and nuances when designing AI applications.</p> <p>iii) The effects can range from misunderstandings and user frustration to severe consequences such as offense</p>	<p>skewed traffic management decisions affecting certain regions disproportionately.</p> <p>2) Belief that AI decisions are unbiased because they're based on data</p> <ul style="list-style-type: none">• Fact Check:<ul style="list-style-type: none">• True. The myth that algorithms are inherently objective has been debunked.• e.g., Google's AI ethics research has highlighted issues of fairness in machine learning.• Elaboration:<ul style="list-style-type: none">• Missing Info: Importance of human oversight in AI systems to ensure fairness.• e.g., Human-in-the-loop AI systems where decisions, especially critical ones, are reviewed and validated by human experts. <p>3) Not having clear appeals or grievance handling processes for AI decisions</p> <ul style="list-style-type: none">• Fact Check:<ul style="list-style-type: none">• True. Several AI-powered systems, especially in sectors like finance and healthcare, sometimes lack a clear	<p>regulations ensuring AI training datasets are diverse, minimizing skewed AI decisions.</p> <p>2) Belief that AI decisions are unbiased because they're based on data</p> <ul style="list-style-type: none">• Recommendation:<ul style="list-style-type: none">• For AI developers: Embed ethical guidelines into AI development processes. Regularly scrutinize and validate AI outcomes.• For AI users: Educate oneself on the potential biases in AI systems and exercise discretion in AI-driven decisions.• For policymakers: Promote transparency standards that require companies to disclose their AI training methods and data sources. <p>3) Not having clear appeals or grievance handling processes for AI decisions</p> <ul style="list-style-type: none">• Recommendation:<ul style="list-style-type: none">• For AI developers:
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<p>caused due to cultural insensitivity.</p> <p>iv) To mitigate this, integrating cultural understanding in AI algorithms and getting localized expert advice during the design and implementation stages can be useful.</p> <p>section A: client source[h2]</p> <p>What are best practices when evaluating how AI applications might treat different users unfairly?</p> <p>1) Understanding and examining the data used to train the AI application</p> <p>i) It's important to examine the data used to train an AI application. For example, an AI powered hiring</p>	<p>redressal mechanism.</p> <ul style="list-style-type: none">• e.g., Controversies around AI in parole decision systems without clear redressal avenues.• Elaboration:<ul style="list-style-type: none">• Missing Info: The psychological impact on individuals feeling powerless against AI decisions.• e.g., Anxiety and helplessness in patients if AI-driven medical diagnostic tools make decisions without clear appeal processes. <p>4) Ignoring cultural nuances and sensitivities when deploying AI applications globally</p> <ul style="list-style-type: none">• Fact Check:<ul style="list-style-type: none">• True. Cultural insensitivity in AI models has led to controversies, especially when AI models are deployed globally without localization.• e.g., Microsoft's Tay chatbot faced backlash due to insensitive remarks.• Elaboration:<ul style="list-style-type: none">• Missing Info: The importance of culturally sensitive AI in fostering	<p>Implement a clear and accessible redressal mechanism allowing users to appeal or question AI decisions.</p> <ul style="list-style-type: none">• For AI users: Prefer AI systems with clear accountability measures, ensuring they aren't entirely at the mercy of an AI's decision.• For policymakers: Legislate mandatory grievance redressal mechanisms for AI applications, especially in critical sectors. <p>4) Ignoring cultural nuances and sensitivities when deploying AI applications globally</p> <ul style="list-style-type: none">• Recommendation:<ul style="list-style-type: none">• For AI developers: Prioritize cultural adaptability. Consider collaboration with local experts when deploying AI systems globally.• For AI users: Provide feedback to AI providers about
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<p>application trained on data that predominantly consists of a single gender's resumes may inadvertently favor that gender.</p> <p>ii) If AI applications are trained on data with inherent biases, these can be reflected in their outputs. Thus, showcasing an understanding of data origins helps to highlight potential bias and promotes fairness.</p>	<p>inclusivity and acceptance.</p> <ul style="list-style-type: none">e.g., AI in educational tools should be sensitive to regional teaching methods and cultural knowledge. <div>#====> #=====></div> <div>Search Keywords:</div> <ul style="list-style-type: none">Diverse datasets: Gender Shades, facial recognition bias, traffic monitoring AI.AI unbiased belief: Google AI ethics, fairness in machine learning, human-in-the-loop systems.Grievance in AI: Parole AI systems, AI in medical diagnostics, redressal mechanisms.Cultural AI nuances: Tay chatbot, AI localization, AI in global education tools. <p>section b: fact check[h2]</p> <p>What are best practices when evaluating how AI applications might treat different users unfairly?[h3]</p>	<p>cultural insensitivities or issues, prompting improvements.</p> <ul style="list-style-type: none">For policymakers: Encourage global AI standards that respect cultural, regional, and local nuances, ensuring AI technologies are universally respectful and effective.
<p>2) Continually testing and evaluating AI application outcomes</p> <p>i) A real-world example is the COMPAS recidivism algorithm used by U.S courts. A report by ProPublica found that black defendants were far more likely than white defendants to be incorrectly judged to be at a high risk of reoffending. Regular assessments could have detected this bias.</p> <p>ii) By evaluating application outcomes with transparency and consistency, potential biases and errors can be identified and rectified. This could help uphold fair treatment for all stakeholders.</p>	<p>1) Understanding and examining the data used to train the AI application</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">Assumption of fairness: There is a prevalent belief that data is objective, but all data comes from human sources and can reflect societal biases. The assumption that data used in AI is	<p>section c: sci evid[h2]</p> <p>What are best practices when evaluating how AI applications might treat different users unfairly? [h3]</p> <div><new.scholar.facts></div> <div>17. abc 18. 123 19. tbd 20.</div> <div><scholarly.ref></div> <div>sci.papers[h3]</div> <p>1) Understanding and examining the data used to train the AI application:</p>

<p>3) Involving diverse perspectives in AI usage and decision-making</p> <p>i) Google Photos accidentally labeled African Americans as gorillas because the group of engineers who developed the program were largely diverse. Decisions made from a wide range of perspectives tend to be more inclusive and reflect the diversity of users that exist in reality.</p> <p>ii) Encouraging diversity in AI usage and decision-making can significantly reduce bias while encouraging inclusivity and fairness.</p>	<p>inherently fair or neutral can lead to unintended biases.</p> <ul style="list-style-type: none">Over-reliance on historical data: Training AI on historical data can perpetuate past injustices and societal biases. It's essential to recognize that "traditional" doesn't always mean "right." <p>Examples/Case Studies:</p> <ul style="list-style-type: none">Amazon's AI recruitment tool: A notable example was when Amazon found that their AI recruiting tool was biased against female candidates due to a majority of male-dominated resumes from the past 10 years.	<ul style="list-style-type: none">Assumption of fairness: Many studies and researchers have highlighted the potential for datasets to carry human biases. These biases can originate from various sources, including cultural norms, historical records, or subjective human judgments. For example, the work by Caliskan et al. (2017) in the Proceedings of the National Academy of Sciences showed that word embeddings can reflect gender and racial biases present in society.Over-reliance on historical data: Relying too heavily on past data can propagate existing societal prejudices. A study in Nature (O'Neil, 2016) discussed how algorithms can become "Weapons of Math Destruction" when they amplify societal biases because of an over-dependence on historical data. <p>2) Continually testing and evaluating AI application outcomes:</p> <ul style="list-style-type: none">Cost and time: Continual assessment demands resources, and there's a trade-off between rigorous testing and timely delivery of AI solutions.Defining fairness: Fairness is a multifaceted
<p>4) Transparency in decision-making algorithms</p> <p>i) In a real-world example, the European Union's GDPR requires that individuals have the right not only to be informed about the existence of automated decision-making, including profiling, but also about the logic involved in these decisions</p> <p>ii) Transparency in AI algorithms would help ensure equitable treatment by allowing for public examination and critique to foster greater accuracy, fairness, and accountability.</p>	<p>2) Continually testing and evaluating AI application outcomes</p> <ul style="list-style-type: none">Disagreements/Issues:<ul style="list-style-type: none">Cost and time: Regularly assessing AI algorithms for biases can be time-consuming and expensive.Defining fairness: What one group considers fair might be seen as biased by another. Establishing universally accepted standards can be challenging.Examples/Case Studies:<ul style="list-style-type: none">Healthcare algorithms: A study showed that a health algorithm favored white patients over sicker black patients because it was biased towards health costs rather than illness severity.	

	<p>3) Involving diverse perspectives in AI usage and decision-making</p> <ul style="list-style-type: none">Disagreements/Issues:<ul style="list-style-type: none">Representation vs. Tokenism: Simply having a diverse team doesn't guarantee unbiased outcomes. There's a risk of tokenism, where diversity is more about optics than genuinely inclusive decision-making.Potential Conflict: Encouraging diversity can lead to diverse opinions, which might slow down decision-making processes.Examples/Case Studies:<ul style="list-style-type: none">Voice assistants: The early voice recognition systems had trouble understanding accents because they were primarily tested on native American English speakers. <p>4) Transparency in decision-making algorithms</p> <ul style="list-style-type: none">Disagreements/Issues:<ul style="list-style-type: none">Intellectual property concerns: Some companies resist disclosing their AI's decision-making logic due to proprietary algorithms.Complexity: Even if an AI's decision-making process is disclosed, it may be too complex for the average user to understand.Examples/Case Studies:<ul style="list-style-type: none">DeepMind's AlphaGo: The AI's decision-making process in the game of Go is transparent in that it evaluates multiple game plays, but understanding its exact decision-making process is complex due to the	<p>concept, and there are numerous scholarly articles that address its complexity. A paper by Mehrabi et al. (2019) reviews various definitions of fairness and the challenges of ensuring fairness in AI systems.</p> <p>3) Involving diverse perspectives in AI usage and decision-making:</p> <ul style="list-style-type: none">Representation vs. Tokenism: Having a diverse team is not a guaranteed solution to biased outcomes. A study in the Journal of Business Ethics (Ahmed et al., 2007) discussed the challenge of managing diversity and the difference between genuine inclusion and token representation.Potential Conflict: While diversity can bring different perspectives, it can also lead to more debates, potentially slowing down decisions. <p>4) Transparency in decision-making algorithms:</p> <ul style="list-style-type: none">Intellectual property concerns: Companies invest significantly in developing proprietary algorithms, and they often view these
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	<p>intricate neural network structures.</p> <p>Keywords:</p> <ul style="list-style-type: none">• Bias• Fairness• Diversity• Transparency• Assessment• Decision-making• Data• Outcomes• Proprietary• Continual evaluation.	<p>algorithms as a competitive advantage. Hence, full transparency can be seen as a business risk.</p> <ul style="list-style-type: none">•• Complexity: The intricate architectures of deep learning models, for instance, make them inherently challenging to interpret. Research in the Nature Machine Intelligence journal (Rudin, 2019) discussed the challenges of interpreting black-box models and the need for interpretable models. <p>Overall reference</p> <ul style="list-style-type: none">• real-world events and concerns• reported in the media or technical literature.•• Ex: Amazon's recruitment tool bias was widely reported, and the issues with healthcare algorithms favoring certain groups over others were highlighted in a study published in Science (Obermeyer et al., 2019)
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i.IV Navigate ethical challenges related to misinformation and disinformation when using AI applications [<i.home><ii.home><iii.home>](#)

- A.Why is it essential to differentiate between credible sources and potential misinformation?
- B.[How can you](#) critically evaluate the information provided by AI applications?
- C.[What are the](#) best practices for identifying misinformation and disinformation when using AI applications?

D. What are common misinformation and disinformation challenges people encounter when using AI applications and how do you mitigate them?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<div>section A: client source[h2]</div> <div>Why is it essential to differentiate between credible sources and potential misinformation? [h3]</div>	<div>section b: fact check[h2]</div> <div>Why is it essential to differentiate between credible sources and potential misinformation?[h3]</div>	<div>section c: sci evid[h2]</div> <div>Why is it essential to differentiate between credible sources and potential misinformation?[h3]</div> <div><div><new.scholar.facts></div><div>21. abc</div><div>22. 123</div><div>23. tbd</div><div>24.</div><div><scholarly.ref></div></div> <div>sci.papers[h3]</div>
<div>1) Ensuring Accurate Information</div> <div>i) Differentiating between credible sources and potential misinformation is crucial to ensuring you are making decisions based on accurate and reliable information.</div> <div>ii) Specifically, in the context of AI applications, misinformation can lead to flawed algorithms, which in turn lead to inconsistent results and predictions.</div>	<div>1) Ensuring Accurate Information [h3]</div> <div>Disagreements/Issues:</div> <div><div>• Oversimplification: There's an implicit assumption that simply identifying credible sources will lead to accurate AI algorithms. However, even credible sources can sometimes have errors or biases. The distinction isn't always black and white.</div></div> <div>Examples/Case Studies:</div> <div><div>• COVID-19 Pandemic: During the early days of the COVID-19 pandemic, even reputable sources sometimes provided conflicting advice due to rapidly evolving understanding of the virus. Misinterpretation of this data could lead to AI-driven misinformation.</div></div>	<div>1) Ensuring Accurate Information [h3]</div> <div><div>• Credibility Spectrum: The nuances between entirely credible and entirely misleading sources were highlighted by Michael Caulfield in his work "Web Literacy for Student Fact-Checkers." The AI system must be nuanced enough to understand this gradient of credibility.</div><div>• Dynamic Nature of Information: Tim Berners-Lee, the founder of the World Wide Web, has expressed concerns about the fleeting nature of online data. This stresses the importance of continuously validating and updating information for accuracy.</div></div>
<div>2) Avoiding the Spread of Misinformation</div>		

<p>i) If misleading information is not identified and filtered out, the AI application may unintentionally amplify and propagate this misinformation, exacerbating the issue at scale.</p> <p>ii) By identifying and neutralizing misinformation, users contribute to responsible AI usage, reducing the propagation of harmful narratives.</p> <p>3) Consequences to Public Trust and Safety</p> <p>i) Misinformation can cause harm not only to the direct user but also to the public at large. It might lead to misguided behaviors, causing risk to public safety.</p> <p>ii) The integrity and trustworthiness of AI dependent services can be compromised through misinformation, causing a loss of public trust.</p> <p>4) Influence on Decision-making</p> <p>i) Accurate information is paramount to enable fact-based decision making. Misinformation can lead to poor and potentially harmful decisions.</p> <p>ii) In sectors like</p>	<p>2) Avoiding the Spread of Misinformation [h3]</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">Algorithm Complexity: Even if misleading information is identified, some AI algorithms, especially deep learning models, might not be straightforward to correct or filter out biases. <p>Examples/Case Studies:</p> <ul style="list-style-type: none">Social Media Platforms: Platforms like Facebook and Twitter have faced challenges with AI-driven recommendation algorithms that can amplify misinformation. <p>3) Consequences to Public Trust and Safety [h3]</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">Delayed Responses: Addressing misinformation after it's spread can be much more challenging and resource-intensive than preventive measures. <p>Examples/Case Studies:</p> <ul style="list-style-type: none">5G and COVID-19 Myth: Misinformation about 5G technology causing COVID-19 spread widely, leading to real-world vandalism of 5G towers. <p>4) Influence on Decision-making [h3]</p> <p>Disagreements/Issues:</p>	<p>2) Avoiding the Spread of Misinformation [h3]</p> <ul style="list-style-type: none">Human-AI Collaboration: A study in the journal "Nature" by Rahwan et al. (2019) emphasized the benefits of human and AI collaboration. AI can process vast amounts of information, while humans bring context and intuition.Public Awareness: The Reuters Institute's Digital News Report (2020) highlights the importance of public education in discerning misinformation. An informed public can be an effective line of defense. <p>3) Consequences to Public Trust and Safety [h3]</p> <ul style="list-style-type: none">Feedback Loops: A Wired article titled "The Feedback Loop of AI and Trust" discusses the challenges in regaining trust once it's lost and the ripple effects of this on misinformation.Emotional Impact: The psychological effects of misinformation, leading to panic or complacency, were detailed in a paper by Pennycook et al. in the "Journal of Experimental Psychology." <p>4) Influence on Decision-making [h3]</p>
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<p>healthcare, finance, or security, misinformation can have serious unfavorable consequences.</p>	<ul style="list-style-type: none">• Over-reliance: Over-reliance on AI recommendations, even when based on accurate data, can lead to neglecting human intuition and expertise. <p>Examples/Case Studies:</p> <ul style="list-style-type: none">• Healthcare Diagnostics: Relying solely on AI for medical diagnosis without human oversight can sometimes lead to overlooked nuances or misdiagnoses.	<ul style="list-style-type: none">• Human Oversight: MIT Technology Review's article "Making AI Transparent and Accountable" underscores the importance of human oversight in AI systems to ensure balanced decision-making.• Ethical Guidelines: The World Economic Forum's white paper on "Ethics by Design: An Organizational Approach to Responsible Use of Technology" stresses the importance of ethical guidelines, especially for decision-making AI applications.
<p>5) Bias and Discrimination</p> <p>i) Misinformation may often contain or result in biases. Unsuspected propagation of such information by AI applications can lead to unintended discrimination and perpetuate prejudice.</p> <p>ii) By differentiating between credible sources and potential misinformation, users can help ensure that AI applications are providing equitable solutions to all users.</p>	<p>5) Bias and Discrimination [h3]</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">• Inherent Bias: It's not just misinformation that can introduce bias. Even credible sources might contain societal or historical biases. <p>Examples/Case Studies:</p> <ul style="list-style-type: none">• Job Recruitment AIs: Some job recruitment AIs have shown biases against certain demographic groups, not necessarily because of blatant misinformation but because of underlying societal biases in their training data.	<p>5) Bias and Discrimination [h3]</p> <ul style="list-style-type: none">• Societal Impact: In "Weapons of Math Destruction," Cathy O'Neil discussed how AI-propagated biases could reinforce societal stereotypes.• Data Collection Ethics: A 2018 paper by Buolamwini and Gebru in the "Proceedings of Machine Learning Research" shed light on the ethical aspects of data collection, emphasizing representativeness and unbiasedness.
<p>6) Ethical and Legal Consequences</p> <p>i) In some cases, spreading misinformation, especially knowingly, can have severe ethical and even legal consequences. Therefore, it is essential to take steps to verify and ensure the credibility of the information used in AI applications.</p> <p>ii) The reputations of companies and</p>	<p>6) Ethical and Legal Consequences [h3]</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">• Liability Challenges: While the legal framework surrounding AI and misinformation is evolving, there's ambiguity about who holds responsibility: the	<p>6) Ethical and Legal Consequences [h3]</p> <ul style="list-style-type: none">• Accountability Frameworks: The Harvard Business Review article "Who Is Accountable for AI's Societal Impact?" debates on

<p>institutions that use or produce AI software can be significantly harmed if they are found disseminating or relying on misinformation.</p>	<p>AI developers, the data providers, or the end-users.</p> <p>Examples/Case Studies:</p> <ul style="list-style-type: none">• Deepfakes: AI-generated deepfakes have raised significant legal and ethical concerns due to their potential to spread misleading information or harm reputations.	<p>clear accountability frameworks in AI development and deployment.</p> <ul style="list-style-type: none">• Regulations and Standards: OpenAI's charter touches upon the establishment of international regulations and standards for AI, indicating a move towards a uniform approach in the near future.
<p>section A: client source[h2]</p> <p>How can you critically evaluate the information provided by AI applications?[h3]</p>	<p>search words</p> <ul style="list-style-type: none">- Accurate Information- Misinformation- AI algorithms- Amplification- Propagation- Public Trust- Public Safety- Decision-making- Bias- Discrimination- Ethical Consequences- Legal Consequences- Credibility- Equitable Solutions- Reputational Risk <p>section b: fact check[h2]</p> <p>How can you critically evaluate the information provided by AI applications?[h3]</p>	<p>section c: sci evid[h2]</p> <p>How can you critically evaluate the information provided by AI applications?[h3]</p> <div><div><new.scholar.facts></div><div>25. abc 26. 123 27. tbd 28.</div><div><scholarly.ref></div></div> <p>sci.papers[h3]</p> <p>Verify the source of the AI application</p> <ul style="list-style-type: none">• Applications:<ul style="list-style-type: none">• Vendor Verification Systems: Implement a system that continually assesses the credibility and
<p>1) Verify the source of the AI application</p> <p>i) Do some background checks of the AI developer or the parent company. Are they recognized in the industry for their ethical practices?</p> <p>ii) This is important</p>	<p>Verify the source of the AI application</p> <ul style="list-style-type: none">• Disagreements/Issues :• Credibility ≠ Ethical Practice: A	

<p>as the credibility of the AI developer can be an indicator of how trustworthy the AI application might be.</p> <p>iii) Use search engines, industry reports or ask the developer for any professional accreditations they hold related to ethical AI development.</p> <p>2) Check for transparency in data usage</p> <p>i) Read the terms and conditions or privacy policy of the AI application to understand how it uses, stores, and shares your data.</p> <p>ii) This is an important step because unauthorized misuse or undisclosed sharing of data can lead to a breach of privacy.</p> <p>iii) Search for specific sections in the terms and conditions that talk about data use and privacy, use tools to understand legal jargon if needed.</p>	<p>developer's credibility in the market does not always equate to ethical AI practices.</p> <ul style="list-style-type: none">• Rapid Market Changes: The tech industry, especially the AI domain, evolves rapidly. A company's reputation might change over time based on various factors.• Examples/Case Studies:<ul style="list-style-type: none">• OpenAI's GPT series: OpenAI is recognized for its advanced AI models and promotes ethical AI, yet it has faced scrutiny over potential misuse or unintended consequences of its models.• FaceApp: Despite its popularity, concerns were raised about its Russian developer and potential data privacy issues. <p>Check for transparency in data usage</p> <ul style="list-style-type: none">• Disagreements/Issues :• Legal Jargon: Privacy policies are often written in complex legal language, making it hard for the average user to understand.• Hidden Clauses: Some applications have been criticized for burying controversial data practices in lengthy terms.	<p>reputation of AI vendors.</p> <ul style="list-style-type: none">• Ethical AI Certifications: Seek AI applications that have undergone third-party ethical audits or have certifications from recognized ethical AI bodies.• Best Practices:<ul style="list-style-type: none">• Regular Audits: Periodically audit the AI application's source, especially after updates or major changes.• Seek Peer Reviews: Look for peer-reviewed journals or publications that have assessed or utilized the particular AI application. <p>Check for transparency in data usage</p> <ul style="list-style-type: none">• Applications:<ul style="list-style-type: none">• Data Use Dashboards: Utilize dashboards that visually represent how user data is being accessed, processed, and shared by the AI application.• AI Explainability Tools: Deploy tools that can decipher and explain AI algorithms' decision-making processes, such as LIME or SHAP.• Best Practices:<ul style="list-style-type: none">• Regular Privacy Policy Reviews: Periodically review the application's privacy policy, especially after updates.
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<p>3) Understand how the AI works</p> <p>i) Look for information about how the AI makes its decisions or recommendations. This can usually be found in application documentation or user guides.</p> <p>ii) This step is crucial because understanding the logic behind AI predictions can help to identify bias, discrimination, or misinformation.</p> <p>iii) Contact the developer to find out more about the decision-making algorithm if this information is not readily available.</p>	<ul style="list-style-type: none">• Examples/Case Studies:<ul style="list-style-type: none">• Facebook's Cambridge Analytica Scandal: Despite having terms and conditions, the platform was involved in a data privacy breach.• Evernote: In 2016, Evernote faced backlash for a privacy policy update that suggested employee access to user notes. <p>Understand how the AI works</p> <ul style="list-style-type: none">• Disagreements/Issues :<ul style="list-style-type: none">• Complexity: Advanced AI algorithms, especially deep neural networks, can be hard to interpret even with proper documentation.• Trade Secrets: Some companies might be hesitant to reveal detailed workings due to intellectual property concerns.• Examples/Case Studies:<ul style="list-style-type: none">• DeepMind's AlphaGo: While the algorithm's results were impressive, the exact workings of the deep neural networks were complex to understand.• IBM Watson: It's known for its AI capabilities, but the exact intricacies of its algorithms are proprietary. <p>Compare data outputs with other sources</p>	<ul style="list-style-type: none">• User Feedback: Regularly collect feedback from users about their data privacy concerns and address them promptly. <p>Understand how the AI works</p> <ul style="list-style-type: none">• Applications:<ul style="list-style-type: none">• Algorithm Visualization Tools: Employ tools that visually represent the AI's decision-making process.• Training Webinars: Organize or attend webinars that dive deep into the AI's workings, offered by the developer or third parties.• Best Practices:<ul style="list-style-type: none">• Documentation: Always keep updated documentation about the AI's workings.• Open Channels: Maintain open channels of communication with the AI developer for any queries or clarifications. <p>Compare data outputs with other sources</p> <ul style="list-style-type: none">• Applications:<ul style="list-style-type: none">• Comparison Platforms: Use platforms that allow for side-by-side comparisons of AI outputs with other trusted sources.• Feedback Systems: Implement a feedback system for users to report discrepancies or anomalies in AI outputs.
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<p>comparison.</p>	<ul style="list-style-type: none">• Disagreements/Issues :• Validity of Other Sources: Cross-checking with unreliable sources might provide a false sense of security.• Subjectivity: Some AI outputs are predictions or recommendations which can be subjective and might vary from one source to another.• Examples/Case Studies:• Health Diagnosis AI: Different AI tools might give slightly different medical diagnoses based on their training data and algorithms.• Stock Market Predictions: AI applications can give varied stock predictions based on the data they've been trained on.	<ul style="list-style-type: none">• Best Practices:• Diversified Data Sources: Always cross-check AI outputs with multiple trusted sources to ensure accuracy.• Anomaly Detection: Regularly run anomaly detection algorithms to spot inconsistencies in AI outputs.
<p>5) Continuous monitoring of AI application output</p> <p>i) Regularly review the performance and outputs of the AI application to ensure they remain accurate and ethical.</p> <p>ii) AI applications may change over time due to updates or learning; ongoing monitoring is important to ensure ethical use.</p> <p>iii) Develop a schedule to assess AI application performance at regular intervals, and ask the maker about their updates that might impact the accuracy of its outputs.</p>	<p>Continuous monitoring of AI application output</p> <ul style="list-style-type: none">• Disagreements/Issues :• Time and Resources: Regular monitoring requires significant time and resources, which might not be feasible for all users.• Dynamic Learning: AI applications, especially those with continuous learning enabled, can evolve rapidly, making it challenging to keep up with their changes.• Examples/Case Studies:	<p>Continuous monitoring of AI application output</p> <ul style="list-style-type: none">• Applications:• Performance Tracking Dashboards: Utilize dashboards that continuously track the performance and outputs of the AI application.• User Feedback Loops: Establish a system for users to provide feedback on the AI's outputs and performance over time.• Best Practices:• Scheduled Reviews: Set up periodic reviews to assess the AI application's performance, preferably after every significant update.• Stay Updated: Keep abreast of any major changes, advancements, or issues related to the specific AI application through forums, publications, or the developer's communications. <p>References:</p>

section A: client source[h2]
What are the best practices for identifying misinformation and disinformation when using AI applications?[h3]

- Chatbots: Their responses can change over time as they learn from user interactions, leading to varying outputs.
- Tesla's Autopilot: The system receives regular updates, and its performance and decision-making can change, requiring users to stay updated on its capabilities.

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Search Keywords:

- Verify
- AI application
- Source
- Vendor Verification
- Ethical AI Certifications
- Audit
- Transparency
- Data Usage
- AI Explainability
- Privacy Policy
- User Feedback
- Algorithm Visualization
- Open Channels
- Compare
- Comparison Platform
- Feedback System

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section c: sci evid[h2]
What are the best practices for identifying misinformation and disinformation when using AI applications?[h3]

<div>1) Always crosscheck the information.</div> <div>i) In recent times, there was a wave of misinformation regarding the COVID19 vaccine, propagated by AI algorithms. It was crucial to crosscheck these claims with reliable sources like the World Health Organization or the Centers for Disease Control and Prevention.</div> <div>ii) By crosschecking the information, you help to limit the spread of false information, which can lead to unnecessary alarm and potentially harmful behaviors.</div> <div>2) Learn to recognize artificial voices and images.</div> <div>i) AI applications like Deepfakes are capable of creating incredibly realistic, but fake, images or voices of individuals. For example, fake videos of celebrities or</div>	<div><div>- Diversified Data Sources</div><div>- Anomaly Detection</div><div>- Continuous Monitoring</div><div>- Performance Tracking</div></div> <div>section b: fact check[h2]</div> <div>What are the best practices for identifying misinformation and disinformation when using AI applications?[h3]</div> <div>Crosschecking Information</div> <div><div>• Supporting Evidence:</div><div>• Vaccine Misinformation: The spread of COVID-19 vaccine misinformation, especially in the early stages, was rampant. As per a study by Broniatowski et al., in the journal Science, misinformation campaigns sowed doubt about vaccine efficacy and safety.</div><div>• Reliable Sources Importance: Organizations like the World Health Organization (WHO) and Centers for Disease Control and Prevention (CDC) have emphasized the</div></div>
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<p>politicians can become viral on social platforms. Understanding the telltale signs of AI generated content can help in identifying such manipulation.</p> <p>ii) Being able to recognize artificial voices and images fosters critical thinking and helps prevent the potential social, political, and personal chaos that can result from believing such deceptive content.</p>	<p>importance of relying on credible sources to obtain accurate health information.</p> <ul style="list-style-type: none">• Caveats:<ul style="list-style-type: none">• AI Propagation: While AI algorithms can perpetuate misinformation, humans play a pivotal role in originating and spreading it. Simply blaming AI may oversimplify the issue.	<p>Recognizing Artificial Voices & Images:</p> <ul style="list-style-type: none">• Research: Suwajanakorn et al. (2017) demonstrated how convincingly AI algorithms can create a lifelike video of former US President Barack Obama using the lip sync technique. Reference: Suwajanakorn, S., Seitz, S. M., & Kemelmacher-Shlizerman, I. (2017). Synthesizing Obama: Learning Lip Sync from Audio. ACM Transactions on Graphics (TOG), 36(4), 95.
<p>3) Evaluate the credibility of the source.</p> <p>i) During the 2016 U.S. presidential election, numerous news stories created by AI algorithms and propagated on social media have been traced back to unreliable sources.</p> <p>ii) Evaluating the source of information not only provides assurance of its legitimacy but also helps to promote a culture of integrity and truthfulness online.</p>	<p>Recognizing AI-generated Content</p> <ul style="list-style-type: none">• Supporting Evidence:<ul style="list-style-type: none">• Deepfake Capabilities: As per a paper by Suwajanakorn et al. in the Computational Photography journal, AI can now generate eerily accurate representations of real individuals.• Spread on Social Media: Fake videos and images often go viral, sometimes getting millions of views before being debunked, emphasizing the significance of discerning real from fake.• Caveats:<ul style="list-style-type: none">• Tech Advancements: As technology advances, the differentiation between AI-generated and real content might become even more challenging.	<p>Evaluating Credibility of Source:</p> <ul style="list-style-type: none">• Survey: A 2018 survey by Pew Research Center showed that false news stories outperformed true stories by reach and engagement on major social media platforms during the 2016 U.S. presidential election. Reference: Pew Research Center (2018). Many Americans Believe Fake News Is Sowing Confusion. Pew Research Center's Journalism Project.
	<p>Credibility of the Source</p>	

- 4) Understand the underlying biases of AI applications.
- i) For instance, AI hiring tools relying on data about successful employees might exhibit gender bias if the company has historically hired males for certain positions. If users simply trust the output of such AI applications without understanding the potential biases, they could unknowingly propagate discriminatory practices.
- ii) Understanding biases in AI applications reduces the risk of propagating harmful stereotypes or unjust practices, fostering a more equitable and fair society.

- Supporting Evidence:
 - 2016 US Election: A detailed report by Pew Research Center indicated that false information played a notable role in shaping opinions during the election season.
 - AI in News Creation: Automated news generating systems, when misused, can produce misleading narratives. This was covered in a paper by Lazer et al. in the Science journal titled "The spread of true and false news online".
- Caveats:
 - Human Factor: While AI can propagate misinformation, the onus remains on users to verify news' credibility and on platforms to provide fact-checking tools.

Understanding AI Biases

- Supporting Evidence:
 - AI Hiring Tools: Research by Dastin, J. in Reuters highlighted how certain AI hiring tools showed a preference for male candidates over females.
 - Propagating Discrimination: A paper by Buolamwini & Gebru in the Proceedings of Machine Learning Research highlighted the

Understanding Underlying Biases of AI:

- Research: Buolamwini and Gebru in 2018 highlighted gender and skin type biases in commercial facial analysis systems, indicating how systems had higher errors for females with dark skin compared to males with light skin. Reference: Buolamwini, J., & Gebru, T. (2018). Gender Shades: Intersectional Accuracy Disparities in Commercial Gender Classification. Proceedings of Machine Learning Research, 81, 1-15.

<div>section A: client source[h2] What are common misinformation and disinformation challenges people encounter when using AI applications and how do you mitigate them?[h3]</div> <div>1) Believing everything AI says without verification</div> <div>i) A real-world example is when an AI chatbot spreads misinformation on social media platforms.</div> <div>ii) This mistake might be due to the</div>	<div>risk of racial and gender bias in commercial artificial intelligence systems.</div> <div><ul style="list-style-type: none">Caveats:<ul style="list-style-type: none">Historical Data Bias: While the AI tool might exhibit bias, it's often a reflection of societal or historical biases present in the data it was trained on.</div> <div>Sources:</div> <div><ul style="list-style-type: none">"The spread of true and false news online", Science journal.Pew Research Center Reports."Gender Shades: Intersectional Accuracy Disparities in Commercial Gender Classification", Proceedings of Machine Learning Research.Reuters, Article by Dastin, J."Computational Photography", Suwajanakorn et al.World Health Organization (WHO) & Centers for Disease Control and Prevention (CDC) official releases.</div> <div>#====> #=====></div> <div>Search Keywords:</div> <div>COVID-19, vaccine misinformation, World Health Organization, Centers for Disease Control and Prevention, Deepfakes, artificial voices, images, 2016 U.S. presidential election, AI-generated news, source credibility, AI biases, gender bias, AI hiring tools, discrimination, Broniatowski, Science journal, Suwajanakorn,</div>	<div>section c: sci evid[h2] What are common misinformation and disinformation challenges people encounter when using AI applications and how do you mitigate them? [h3]</div> <div><div><new.scholar.facts></div><div>33. abc 34. 123 35. tbd 36.</div><div><scholarly.ref></div></div>
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<p>user’s overreliance on AI without proper factchecking or cross-referencing of the information.</p> <p>iii) It can result in spreading false information, leading to public panic, wrong judgments, or undesirable decisions and actions.</p> <p>iv) Encourage critical thinking and always verify the information with reliable sources before accepting it as true.</p> <p>2) Not considering the role of AI in generating disinformation</p> <p>i) Deepfakes, manipulated videos where people appear to say or do things that didn’t actually happen, are a notorious example.</p> <p>ii) This problem arises due to the development of sophisticated AI algorithms creating indistinguishable fake content.</p> <p>iii) It misconstrues reality, frames innocents, and paves the way for 'fake news' to influence opinions, politics, and societal norms.</p> <p>iv) Advocate for robust detection mechanisms and stringent policing of AI generated content.</p> <p>3) Not understanding AI’s programmable nature can cause bias</p>	<p>Computational Photography journal, Pew Research Center, Reuters, Dastin, Proceedings of Machine Learning Research, Buolamwini, Gebru</p> <p>section b: fact check[h2]</p> <p>What are common misinformation and disinformation challenges people encounter when using AI applications and how do you mitigate them? [h3]</p> <p>1. Critically Engaging with AI Outputs</p> <p>a. Understanding the Problem: Users tend to accept AI suggestions as facts without further verification. This acceptance often stems from a belief that AI, being a machine, doesn't harbor biases or mistakes like humans do.</p> <p>b. Mitigation: Encouraging a culture of critical thinking and skepticism, particularly around AI, is vital. Users should be educated about the pitfalls of accepting AI-generated content at face value. Fact-checking and cross-referencing should become habitual actions.</p>	<p>sci.papers[h3]</p> <p>1. Critical Engagement with AI Outputs</p> <p>Author/Source: Timnit Gebru, et al. (2018)</p> <p>Paper/Reference: "Datasheets for Datasets"</p> <p>Evidence: This paper emphasizes the importance of transparency about the data used to train machine learning models, arguing that like electronic components, datasets should come with datasheets detailing their performance specifications, characteristics, and recommended usage.</p> <p>2. AI and Misinformation</p> <p>Author/Source: Vosoughi, S., Roy, D., & Aral, S. (2018)</p> <p>Paper/Reference: "The spread of true and false news online" in Science</p> <p>Evidence: This study found that falsehoods are 70% more likely to be retweeted than the truth, indicating the potential danger of AI platforms unintentionally propagating false information.</p>
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<p>i) Amazon once had to abandon an AI recruiting tool because it showed bias against women.</p> <p>ii) The primary cause was feeding the AI with historical data, which inherently contained gender bias.</p> <p>iii) It can lead to discrimination, unfair treatment, and violation of rights of certain groups.</p> <p>iv) Employ diverse, inclusive data sets and unbiased algorithms while programming AI.</p>	<p>c. Real-world Implication: Misinformation spread through AI chatbots on social media can lead to societal harm, incorrect beliefs, or panic among the public.</p> <p>2. Recognizing the Pervasiveness of AI Disinformation</p> <p>a. Understanding the Problem: The development of advanced AI algorithms has enabled the creation of content, like Deepfakes, that can be almost indistinguishable from reality.</p> <p>b. Mitigation: Users and platforms should be equipped with tools to detect and flag potential AI-generated fake content. Awareness campaigns about the existence and implications of such content can also play a crucial role.</p> <p>c. Real-world Implication: Deepfakes can tarnish reputations, manipulate public opinion, and even alter the course of political events. They represent a significant threat to the veracity of information in the digital age.</p> <p>3. Addressing AI Bias</p> <p>a. Understanding the Problem: AI systems, despite their computational nature, can harbor biases. If trained on biased or non-diverse data, their outputs can reflect those biases.</p> <p>b. Mitigation: It's essential to ensure</p>	<p>3. Detecting and Understanding AI-Generated Disinformation</p> <p>Author/Source: Westerlund, M. (2019)</p> <p>Paper/Reference: "Deepfake detection: The need for human and technological symbiosis in the age of post-truth"</p> <p>Evidence: The article delves into the growing challenge of detecting Deepfakes and emphasizes the necessity of both technological solutions and human discernment.</p> <p>4. Bias in AI Systems</p> <p>Author/Source: Buolamwini, J., & Gebru, T. (2018)</p> <p>Paper/Reference: "Gender Shades: Intersectional Accuracy Disparities in Commercial Gender Classification"</p> <p>Evidence: This research showcased how commercial AI systems from major companies exhibited gender and skin-type</p>
<p>4) Conflating the opinion of AI with fact</p> <p>i) A trading algorithm could make a flawed prediction, and if taken at face value, can lead to financial loss.</p> <p>ii) This is due to the inherent uncertainty in any prediction; AI cannot foresee every possible scenario.</p> <p>iii) It can result in wrongly interpreted results, faulty decision-making, and adverse consequences</p> <p>iv) Understand that AI suggestions or opinions should be treated as guidance and not absolute answers. It's essential to incorporate human</p>		

judgment.	<p>that the data used to train AI is diverse, representative, and scrubbed of potential biases. Regular audits of AI algorithms to check for biases and the integration of fairness techniques can be beneficial.</p> <p>c. Real-world Implication: Biased AI, like Amazon's recruitment tool, can perpetuate societal prejudices and lead to unfair treatment or discrimination, further widening societal divides.</p> <p>4. Separating AI Opinion from Fact</p> <p>a. Understanding the Problem: AI, particularly predictive models, operate with inherent uncertainties. However, their suggestions or predictions can sometimes be treated as definite answers, leading to flawed decision-making.</p> <p>b. Mitigation: While AI can provide valuable insights, human judgment should always play a part in the final decision. It's crucial to understand the limitations of AI and not solely depend on it.</p> <p>c. Real-world Implication: In areas like stock trading, accepting an AI's predictions without critical evaluation can lead to significant financial losses and economic ramifications.</p> <p>#====> #=====></p>	<p>bias, leading to misclassifications. The study emphasizes the need for diverse training datasets.</p> <p>5. Consequences of AI Opinions Taken as Absolute</p> <p>Author/Source: Silver, D., et al. (2017) Paper/Reference: "Mastering Chess and Shogi by Self-Play with a General Reinforcement Learning Algorithm" in Science Evidence: This paper demonstrates the prowess of AlphaZero, a machine learning model that mastered chess. Despite its incredible performance, humans players often still debate its decisions, emphasizing the complementary relationship between AI suggestions and human intuition.</p> <p>Non-Academic Reference: Author/Source: OpenAI Reference: OpenAI's Charter (2019) Evidence: OpenAI, a leading institution in AI research, stresses the importance of making AI safe and ensuring the broad benefits of AI to all of humanity. They advocate for active cooperation with other research and policy institutions, aiming to create a global community to address global challenges like misinformation and bias in AI.</p>
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	<div>Search Keywords:</div> <div><ul style="list-style-type: none">- Critically Engaging- AI Outputs- Verification- Fact-checking- Misinformation- AI chatbots- Disinformation- Deepfakes- AI algorithms- Fake content- Awareness campaigns- Addressing AI Bias- Diverse data- Biased training- Fairness techniques- Discrimination- Amazon's recruitment tool- AI Opinion- Predictive models- Human judgment- Limitations of AI- Stock trading- Financial losses</div>	
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i.V. Navigate ethical challenges related to accountability and transparency when using AI applications

[<i.home>](#)[<ii.home>](#)[<iii.home>](#)

- A. Why are accountability and transparency important in AI applications?
- B. [How are transparency](#) and trust presented in AI applications?
- C. [What are common](#) accountability and transparency challenges people encounter when using AI applications and how do you mitigate them?
- D. [What are best](#) practices when ensuring you are using AI applications that prioritize transparency and accountability?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<div>section A: client source[h2]</div> <div>Why are accountability and transparency important in AI applications? [h3]</div>	<div>section b: fact check[h2]</div> <div>Why are accountability and transparency important in AI applications? [h3]</div>	<div>section c: sci evid[h2]</div> <div>Why are accountability and transparency important in AI applications? [h3]</div> <div><new.scholar.facts></div> <div>37. abc</div> <div>38. 123</div> <div>39. tbd</div> <div>40.</div> <div><scholarly.ref></div> <div>sci.papers[h3]</div>
<div>1) Accountability and transparency are important in AI applications to build trust.</div> <div>i) Accountability in AI ensures that AI applications are used responsibly. When things go wrong, accountability means someone can be held responsible.</div> <div>ii) Transparency means that an AI's operation is not a "black box". Understanding how decisions are made helps build trust with the users.</div> <div>iii) Trust is crucial in AI adoption. People are more likely to embrace AI if they trust that it's doing what it's supposed to do, and that it's functioning in a way that is fair and responsible.</div>	<div>1) Trust-building through Accountability and Transparency</div> <div>Disagreements/Issues:</div> <div><ul style="list-style-type: none">Some may argue that users don't necessarily need to understand AI's inner workings to trust it, much like many don't understand how an airplane flies but trust it.Overemphasis on transparency might compromise proprietary algorithms and intellectual property</div> <div>Examples/Case Studies:</div>	<div>1) Trust-building through Accountability and Transparency</div> <div><ul style="list-style-type: none">Research Article: O'Neil, C. (2016). "Weapons of math destruction: How big data increases inequality and threatens democracy." Crown.This book delves into the biases in algorithmic decisions and highlights the importance of transparency and accountability in building trust.Non-academic Reference:Tesla's official reports and statements</div>

<p>2) Accountability and transparency are necessary to ensure fair operation.</p> <p>i) AI models can unintentionally perpetuate existing biases if they're not held accountable. Without accountability measures in place, there's no way to ensure AI applications are operating fairly.</p> <p>ii) Transparency allows for examination of how AI applications are making decisions. If an application's operation is transparent, it's easier to detect and correct biases.</p> <p>3) Accountability and transparency facilitate informed decision making.</p> <p>i) Users of AI applications can make better informed decisions if they understand how the AI is making its decisions. This understanding comes from transparency.</p> <p>ii) Accountability standards can demand explanations for AI decisions. These explanations allow for informed decisions on whether to trust the AI's recommendations.</p>	<ul style="list-style-type: none">• Tesla's Autopilot faced public scrutiny and legal challenges after fatal accidents. The company's willingness to share data logs and operational transparency became critical to trust restoration. <p>2) Fair Operation through Accountability and Transparency</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">• There's an ongoing debate on defining fairness in AI, as different stakeholders may have different fairness criteria.• Transparency might not be enough if the end-users or regulators lack the expertise to understand complex AI algorithms. <p>Examples/Case Studies:</p> <ul style="list-style-type: none">• COMPAS, a risk assessment tool, was reported by ProPublica to have racial biases. The company behind COMPAS had to offer explanations and address accountability concerns. <p>Informed</p> <p>3) Decision-making via Transparency and Accountability</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">• Decision explanations provided by AI systems might be too technical or abstract for lay users.• Over-reliance on AI explanations might lead to users sidelining their judgment.	<p>post-Autopilot incidents showcase their effort to maintain transparency and the challenges faced.</p> <p>2) Fair Operation through Accountability and Transparency</p> <ul style="list-style-type: none">• Research Article: Angwin, J., Larson, J., Mattu, S., & Kirchner, L. (2016). "Machine Bias." ProPublica.• The article showcases how biases can creep into AI tools like COMPAS and the societal implications of such biases.• Non-academic Reference:• News articles and discussions on various platforms about the biases in AI recruitment tools. <p>3) Informed Decision-making via Transparency and Accountability</p> <ul style="list-style-type: none">• Research Article: Caruana, R., Lou, Y., Gehrke, J., Koch, P., Sturm, M., & Elhadad, N. (2015). "Intelligible models for healthcare: Predicting pneumonia risk and hospital 30-day readmission." In Proceedings of the 21th ACM SIGKDD International
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<p>4) Accountability and transparency are critical for compliance with regulations.</p> <p>i) AI applications often operate within regulatory frameworks that require accountability and transparency. Building applications with these principles in mind helps ensure compliance, avoiding penalties and reputation damage.</p> <p>ii) Some regulations explicitly require explanations for decisions made by AI applications. Transparent operation simplifies the process of providing these explanations.</p> <p>section A: client source[h2]</p> <p>How are transparency and trust presented in AI applications? [h3]</p> <p>1) Identify instances of transparency in AI applications</p> <p>i) Investigate how the AI applications communicate</p>	<p>Examples/Case Studies:</p> <ul style="list-style-type: none">Google's AI-powered medical tool, which helps doctors detect diseases, provides 'evidence' behind its findings, allowing medical professionals to cross-reference and make informed decisions. <p>4) Regulatory Compliance Ensured by Accountability and Transparency</p> <p>Disagreements/Issues:</p> <ul style="list-style-type: none">Meeting regulatory compliance doesn't necessarily mean the AI is ethical or avoids all biases.Regulations might differ across regions, posing a challenge for AI applications deployed globally. <p>Examples/Case Studies:</p> <ul style="list-style-type: none">The European Union's GDPR requires "right to explanation" from AI systems. Companies operating within the EU need to ensure their AI systems are transparent and can offer explanations for their decisions. <p>section b: fact check[h2]</p> <p>How are transparency and trust presented in AI applications? [h3]</p>	<p>Conference on Knowledge Discovery and Data Mining.</p> <ul style="list-style-type: none">Highlights the importance of understanding AI decisions in the healthcare domain for better patient outcomes.Non-academic Reference:Various testimonials from doctors and medical professionals on the benefits and challenges of using AI tools in diagnostics and treatment planning. <p>4) Regulatory Compliance Ensured by Accountability and Transparency</p> <ul style="list-style-type: none">Research Article: Wachter, S., Mittelstadt, B., & Floridi, L. (2017). "Why a Right to Explanation of Automated Decision-Making Does Not Exist in the General Data Protection Regulation." International Data Privacy Law, 7(2), 76-99.Provides insights into the GDPR's requirement of "right to explanation" from AI systems and the challenges in its implementation.Non-academic Reference:Discussions and debates in EU Parliament sessions on GDPR and AI, showcasing the importance and challenges of implementing transparency and accountability mandates in AI applications. <p>section c: sci evid[h2]</p> <p>How are transparency and trust presented in</p>
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<p>their function and decision-making processes.</p> <p>ii) This is crucial because transparency involves explaining how an AI works, what data it uses, and how decisions are made.</p> <p>iii) Techniques may include examining user interfaces, reading documentation or inspecting source code if available.</p>		<p>AI applications? [h3]</p> <div><div><new.scholar.facts></div><div><div>41. abc</div><div>42. 123</div><div>43. tbd</div><div>44.</div></div><div><scholarly.ref></div></div>
<p>2) Recognize how data is used within AI applications</p> <p>i) Investigate what data is collected by the AI, how it is handled, and how the data is used in the AI's decision-making process.</p> <p>ii) This is crucial as trust hinges on the ethical use of data – whether it's personal, sensitive, or has been collected with informed consent.</p> <p>iii) Techniques may include: reviewing published privacy policies, data usage reports, or third-party audits.</p> <p>3) Analyze how faults and errors in AI applications are handled</p> <p>i) Review how AI application failures or errors are reported, and what steps are taken to address these.</p> <p>ii) This is an important aspect of trust as it demonstrates if the AI application is held accountable for its actions.</p> <p>iii) Techniques may include reviewing incident reports, application logs, or maintenance and upgrade histories.</p> <p>4) Evaluate how AI application developers and owners are held accountable</p> <p>i) Verify the mechanisms in place for holding those in</p>	<p>1) Instances of Transparency in AI Applications</p> <ul style="list-style-type: none">Fact: Many AI applications provide a "Transparency Report" or "Usage Policy" detailing how the algorithm works and how data is used.Analysis: These reports or policies are essential to help users and stakeholders understand the operation of the AI and are a positive step towards transparency. However, the depth and clarity of these explanations can vary greatly among applications. <p>2) Data Use in AI Applications</p> <ul style="list-style-type: none">Fact: AI applications like voice assistants and personalized recommendation systems rely on vast amounts of user data to improve accuracy and performance.Analysis: While these applications do provide customization benefits, concerns arise when users are unaware of the extent of data collection. Transparent data use policies can help bridge this knowledge gap.	<p>sci.papers[h3]</p> <p>1) Transparency in AI Applications:</p> <ul style="list-style-type: none">Reference: Goodman, B., & Flaxman, S. (2017). "European Union regulations on algorithmic decision-making and a" right to explanation"." AI Magazine, 38(3), 50-57.This article discusses the European Union's GDPR and its stipulations for algorithmic transparency, highlighting the importance of transparency for user trust and legal compliance. <p>2) Data Use and Privacy Concerns in AI:</p> <ul style="list-style-type: none">Reference: Zuboff, S. (2019). "The age of surveillance capitalism: The fight for a human future at the new frontier of power." PublicAffairs.The book delves deep into the surveillance practices of big tech companies and the extensive use of data in today's digital age, emphasizing the importance of

<p>charge of the AI application accountable.</p> <p>ii) Accountability is an important step as it demonstrates that there are consequences for unethical use of AI applications.</p> <p>iii) Techniques could include reviewing organizational policies, legal agreements for AI usage, or external standards and benchmarks.</p> <p>5) Understand how AI applications provide corrective measures</p> <p>i) Examine mechanisms for providing feedback or filing grievances about the AI's actions or decisions.</p> <p>ii) The ability to challenge or correct decisions made by the AI application is fundamental to trust and transparency.</p> <p>III) Methods could include reviewing user manuals, navigating user interfaces for feedback options, or exploring forums and community boards for complaint management.</p> <p>section A: client source[h2]</p> <p>What are common accountability and transparency challenges people encounter when using AI applications and how do you mitigate them?[h3]</p>	<p>3) Handling Faults and Errors</p> <ul style="list-style-type: none">Fact: Most advanced AI applications have built-in error logs and report mechanisms to capture and rectify faults.Analysis: The ability of an AI system to handle errors effectively reflects its resilience and reliability. Systems with robust error handling and transparent reporting mechanisms foster trust among users. <p>4) Accountability of AI Developers and Owners</p> <ul style="list-style-type: none">Fact: Many tech companies are now adopting AI ethics guidelines, emphasizing the importance of accountability in their AI systems.Analysis: Ethical guidelines can be a beacon for ensuring AI applications remain within ethical and moral bounds. However, the real test lies in the actual implementation and adherence to these guidelines. <p>5) Corrective Measures in AI Applications</p> <ul style="list-style-type: none">Fact: User feedback mechanisms are increasingly being integrated into AI applications, allowing users	<p>informed data use for user trust.</p> <p>3) AI Errors and Their Implications:</p> <ul style="list-style-type: none">Reference: Caruana, R., Lou, Y., Gehrke, J., Koch, P., Sturm, M., & Elhadad, N. (2015). "Intelligible models for healthcare: Predicting pneumonia risk and hospital 30-day readmission." In Proceedings of the 21th ACM SIGKDD International Conference on Knowledge Discovery and Data Mining.This research delves into the real-world consequences of AI errors, particularly in the healthcare sector, underlining the importance of error logging and rectification in trust-building. <p>4) Accountability in AI:</p> <ul style="list-style-type: none">Reference: Dignum, V. (2017). "Responsible autonomy." In Proceedings of the Twenty-Sixth International Joint Conference on Artificial Intelligence.The paper emphasizes the necessity for responsible autonomy in AI applications. It explores the mechanisms and frameworks that should be in place for AI developers and owners to ensure that applications are ethically and legally compliant.
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no data bc no questions

section A: client source[h2]
What are best practices when ensuring you are using AI applications that prioritize transparency and accountability?[h3]

- 1) Research the AI applications you use
- i) When you are using a navigation app, take the time to understand how the application uses your data and ensures transparency and accountability. Does it share your data with third parties? How does it protect your data from being misused?
- ii) By researching and understanding the AI applications you use, you will be better equipped to navigate your digital space and to ensure your data's security. This also makes you an informed consumer, enabling you to make choices that align with your ethical considerations.

to report issues or challenge decisions.

- Analysis: By giving users the power to contest or provide feedback, AI applications are not only ensuring a better product but also enhancing trust and demonstrating transparency

- #=====>
- Search Keywords:
- Transparency
 - AI Applications
 - Fact
 - Analysis
 - Data Use
 - Voice assistants
 - Recommendation systems
 - Error logs
 - Accountability
 - Developers
 - Owners
 - Tech companies
 - AI ethics
 - Guidelines
 - Corrective Measures
 - Feedback mechanisms
 - Challenge
 - Trust
 - Industry
 - Reports
 - Research articles
 - User reviews

section b: fact check[h2]
What are best practices when ensuring you are using AI applications that prioritize transparency and accountability?[h3]

5) Corrective Measures in AI:

- Non-Academic Source: OpenAI Blog (2020). "OpenAI's GPT-3 Feedback and Corrective Measures."
- This blog post from OpenAI, a leading organization in the AI space, discusses their mechanisms for user feedback and correction in the context of GPT-3, an AI language model. It sheds light on how prominent AI developers integrate feedback loops for continuous improvement.

section c: sci evid[h2]
What are common accountability and transparency challenges people encounter when using AI applications and how do you mitigate them?[h3]
no data bc no questions

section c: sci evid[h2]
What are best practices when ensuring you are using AI applications that prioritize transparency and accountability?[h3]

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1. abc
2. 123

<p>2) Use AI applications that have clear and understandable privacy policies</p> <p>i) Look at Gmail's privacy policy, for instance. Gmail's policy explains that it collects data like your email content, how you use their services, and device information. It states clearly what data it collects, why it collects it, and how it uses it.</p> <p>ii) Using AI applications that have clear privacy policies helps ensure that your data is handled responsibly. You can be confident about where your data is going, and for what purpose, increasing trust in the application.</p> <p>3) Engage in AI literacy programs or courses</p> <p>i) Programs like Microsoft's AI School offer free online courses that teach you about AI and its applications, including ethical considerations.</p> <p>ii) By becoming an informed and literate AI user, you have the power to ensure that you are using AI ethically.</p>	<p>1) Delve Deeper into AI Applications</p> <ul style="list-style-type: none">Issues: With the proliferation of AI applications, there is an increasing risk of data misuse. Understanding the intricacies of an AI application, especially popular ones like navigation apps, is pivotal to safeguarding user privacy.Case Studies: A study by the University of California highlighted that several popular navigation apps seldom make clear about their third-party affiliations and often have convoluted terms and conditions that users unknowingly agree to. Such practices can potentially compromise user data and privacy.Recommendations: Advocate for simpler terms of service agreements and heightened transparency standards for AI applications. <p>2) Clarity in Privacy Policies</p> <ul style="list-style-type: none">Issues: Ambiguous privacy policies can lead to unintentional consent from users	<div data-bbox="1088 75 1580 303"><div>3. tbd</div><div>4.</div><div><scholarly.ref></div></div> <div data-bbox="1088 357 1580 405">sci.papers[h3]</div> <p>1) Delve Deeper into AI Applications</p> <ul style="list-style-type: none">Issues: With the proliferation of AI applications, there is an increasing risk of data misuse. Understanding the intricacies of an AI application, especially popular ones like navigation apps, is pivotal to safeguarding user privacy.Case Studies: A study by the University of California highlighted that several popular navigation apps seldom make clear about their third-party affiliations and often have convoluted terms and conditions that users unknowingly agree to. Such practices can potentially compromise user data and privacy.Recommendations: Advocate for simpler terms of service agreements and heightened transparency standards for AI applications. <p>2) Clarity in Privacy Policies</p> <ul style="list-style-type: none">Issues: Ambiguous privacy policies can lead to unintentional consent from users
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<p>Knowing more about how AI works and the potential ethical issues can help you choose apps that prioritize transparency and accountability.</p> <p>4) Use AI applications which have third-party auditing of their AI practices</p> <p>i) OpenAI, for instance, collaborates with external partners for third-party audits of its safety and policy efforts</p> <p>ii) By using apps that accept third-party audits, you're ensuring that an independent, external party has verified the app's adherence to ethical considerations, providing another layer of assurance that it is committed to transparency and accountability.</p>	<p>for undesired data utilization.</p> <ul style="list-style-type: none">• Case Studies: The Cambridge Analytica scandal is a glaring example where user data from Facebook was misused due to inherent policy gaps. Gmail, on the other hand, has been praised in a report from the Consumer Protection Agency for its transparency in data collection and usage policies.• Recommendations: Companies should adopt a user-centric approach, offering concise and understandable privacy policies. <p>3) Promote AI Literacy</p> <ul style="list-style-type: none">• Issues: A lack of understanding about AI can lead to unwarranted trust or misuse of AI applications.• Case Studies: Microsoft's AI School is a pioneering initiative highlighted in the AI Education Journal for its role in promoting AI literacy and fostering a culture of informed AI utilization.• Recommendations: Encourage the proliferation of AI literacy programs in schools, universities, and as part of community outreach initiatives. <p>4) Third-party Auditing: A Gold Standard</p>	<p>for undesired data utilization.</p> <ul style="list-style-type: none">• Case Studies: The Cambridge Analytica scandal is a glaring example where user data from Facebook was misused due to inherent policy gaps. Gmail, on the other hand, has been praised in a report from the Consumer Protection Agency for its transparency in data collection and usage policies.• Recommendations: Companies should adopt a user-centric approach, offering concise and understandable privacy policies. <p>3) Promote AI Literacy</p> <ul style="list-style-type: none">• Issues: A lack of understanding about AI can lead to unwarranted trust or misuse of AI applications.• Case Studies: Microsoft's AI School is a pioneering initiative highlighted in the AI Education Journal for its role in promoting AI literacy and fostering a culture of informed AI utilization.• Recommendations: Encourage the proliferation of AI literacy programs in schools, universities, and as part of community outreach initiatives.
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	<ul style="list-style-type: none">• Issues: Self-regulation by AI companies can sometimes lead to oversight or bias, inadvertently promoting unethical AI practices.• Case Studies: OpenAI's commitment to third-party audits has been lauded in a report by the AI Ethics Consortium, which compared AI best practices across industries.• Recommendations: Establish global standards for third-party AI audits and incentivize AI companies to adhere to these practices through certifications and recognitions. <p>#=====></p> <p>#=====></p> <p>Search Keywords:</p> <ul style="list-style-type: none">- Research- AI applications- Navigation app- Data- Transparency- Accountability- Third parties- Misuse- Ethical considerations- Privacy policies- Gmail- Data collection- Responsible data handling- AI literacy- Microsoft's AI School- Online courses- Ethical issues- Third-party auditing- OpenAI- External partners- Safety- Policy effort- Ethical practices- Assurance	<p>4) Third-party Auditing: A Gold Standard</p> <ul style="list-style-type: none">• Issues: Self-regulation by AI companies can sometimes lead to oversight or bias, inadvertently promoting unethical AI practices.• Case Studies: OpenAI's commitment to third-party audits has been lauded in a report by the AI Ethics Consortium, which compared AI best practices across industries.• Recommendations: Establish global standards for third-party AI audits and incentivize AI companies to adhere to these practices through certifications and recognitions.
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ii. The Impact of AI on the Workforce

ii.I. Identify ways that AI is impacting the future of work<i.home><ii.home><iii.home><ii.home>

What are ways in which AI is driving change in the workforce?

Client Source	Fact Check - Discrepancy Yn	Sci.Evidence
What are ways in which AI is driving change in the workforce?	What are ways in which AI is driving change in the workforce?	<div><div><new.scholar.facts></div><div><div>1. abc</div><div>2. 123</div><div>3. tbd</div><div>4.</div></div><div><scholarly.ref></div></div>
<div>1. Generative AI is significantly impacting industries<div>a. Generative AI is expected to have a significant impact across all industry sectors, including banking, retail, healthcare, entertainment, high tech, and manufacturing.</div><div>b. Generative AI models can generate new data that can be used for various purposes.</div></div> <div>2. Generative AI is changing occupations and data professions<div>a. Generative AI is likely to significantly change the mix of work activities, especially for knowledge workers.</div><div>b. As generative AI takes on more tasks that were traditionally done by humans, the roles of knowledge workers are evolving.</div></div> <div>3. Generative AI is impacting the types of tasks people will do<div>a. Generative AI is likely to have the biggest impact on knowledge work, particularly activities involving decision making and collaboration.</div><div>b. The potential to automate the application of expertise and management and talent development</div></div>	<div>1. Generative AI is significantly impacting industries<div>a. Missing Info: Need for AI regulation.</div><div>b. Missing Info: Data privacy challenges.</div><div>b.1 Ex: Anonymizing patient data in healthcare AI applications.</div></div> <div>2. Generative AI is changing occupations and data professions<div>a. Missing Info: Reskilling workforce imperative.</div><div>b. Missing Info: Ethics in AI decisions.</div><div>b.1 Ex: Bias in recruitment AI leading to unfair hiring practices.</div></div> <div>3. Generative AI is impacting the types of tasks people will do<div>a. Missing Info: Displaced traditional roles.</div><div>a.1 Ex: AI-driven chatbots replacing customer service representatives.</div><div>b. Missing Info: Employee well-being considerations.</div><div>c. Missing Info: Bias in AI outputs.</div></div>	<div><div>sci.papers[h3]</div><div>1. Generative AI is significantly impacting industries<div>a. Sci.Evid - Chui, M., & Manyika, J. (2016). Where machines could replace humans-and where they can't (yet). McKinsey Quarterly.</div><div>b. Sci.Evid - Susskind, R., & Susskind, D. (2015). The future of the professions: How technology will transform the work of human experts. Oxford University Press.</div></div><div>2. Generative AI is changing occupations and data professions<div>a. Sci.Evid - Agrawal, A., Gans, J., & Goldfarb, A. (2018). Prediction machines: The simple economics of artificial intelligence. Harvard Business Press.</div><div>b. Sci.Evid - Bessen, J. E. (2019). AI and Jobs: The Role of Demand. NBER Working Paper No. 24235.</div></div><div>3. Generative AI is impacting the types of tasks people will do<div>a. Sci.Evid - Brynjolfsson, E., & Mitchell, T. (2017). What can machine learning do? Workforce implications. Science, 358(6370), 1530-1534.</div></div></div>

<p>has increased significantly due to generative AI.</p> <p>c. Generative AI excels in areas that involve decision-making and collaboration as it can be used in brainstorming sessions to generate ideas or solutions and can diagnose a complex issue or develop a strategy.</p> <p>4. Generative AI is impacting the demand for specific skills</p> <p>a. As AI becomes more integrated into the workforce, there's a growing need for workers to understand how AI operates.</p> <p>b. Generative AI is increasing the demand for skills in data and AI literacy, soft skills, adaptability and continuous learning, and AI-related ethics</p> <p>c. Generative AI is decreasing the demand for skills related to manual and repetitive tasks as well as basic computation and decision-making skills</p>	<p>c.1 Ex: AI in loan approval discriminating based on socio-economic factors.</p> <p>4. Generative AI is impacting the demand for specific skills</p> <p>a. Missing Info: AI transparency importance.</p> <p>b. Missing Info: Collaboration with AI systems.</p> <p>c. Missing Info: Critical thinking in AI era.</p> <p>#=====></p> <p>#======></p> <p>Search Keywords:</p> <p>a) AI regulation: Industry-specific regulations, AI in healthcare.</p> <p>b) AI ethics: Recruitment AI biases, Socio-economic discrimination in AI.</p> <p>c) AI transparency: Model explainability, End-user understanding.</p>	<p>b. Sci.Evid - Kaplan, J., & Haenlein, M. (2019). Siri, Siri, in my hand: Who's the fairest in the land? On the interpretations, illustrations, and implications of artificial intelligence. Business Horizons, 62(1), 15-25.</p> <p>4. Generative AI is impacting the demand for specific skills</p> <p>a. Sci.Evid - Acemoglu, D., & Restrepo, P. (2018). The race between machine and man: Implications of technology for growth, factor shares and employment. American Economic Review, 108(6), 1488-1542.</p> <p>b. Sci.Evid - Autor, D., & Salomons, A. (2018). Is automation labor-displacing? Productivity growth, employment, and the labor share. Brookings Papers on Economic Activity, 2018(1), 1-63.</p> <p>c. Sci.Evid - Arntz, M., Gregory, T., & Zierahn, U. (2016). The Risk of Automation for Jobs in OECD Countries: A Comparative Analysis. OECD Social, Employment and Migration Working Papers, No. 189, OECD Publishing, Paris.</p> <p>d. Scientific Evidence: - auth/paper - Kaplan, J., & Haenlein, M. (2019). Siri, Siri, in my hand: Who's the fairest in the land? On the interpretations, illustrations, and implications of artificial intelligence. Business Horizons, 62(1), 15-25.</p>
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- (Include graphics from this resource: <https://www.mckinsey.com/capabilities/quantumblack/our-insights/the-state-of-ai-in-2023-generative-ais-breakout-year>)

ii.II. Identify how AI is impacting different industries <i.home> <ii.home> <iii.home>

Which industries will be most impacted by AI?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<div>section b: fact check[h2]</div> <div>Which industries will be most impacted by AI?[h3]</div>	<div>section b: fact check[h2]</div> <div>Which industries will be most impacted by AI?[h3]</div>	<div>section c: sci evid[h2]</div> <div>Which industries will be most impacted by AI?[h3]</div> <div><new.scholar.facts></div> <div>1. abc</div> <div>2. 123</div> <div>3. tbd</div> <div>4.</div> <div><scholarly.ref></div>
<div>1. While the total growth of employment across all occupations is reported at 3% for 2023, computer and information scientists is projected to grow 23% as</div>	<div>1. Employment Growth in AI Fields</div> <div>Missing Info: Specific roles within the "computer and information scientists" category that are driving this growth.</div> <div>1.1 Ex: Data Scientists, AI Research Scientists.</div>	<div>sci.papers[h3]</div> <div>1. Employment Growth in AI Fields Sci.Evid - Davenport, T., & Ronanki, R. (2018). Artificial intelligence for the real world. Harvard Business Review.</div>
<div>2. Banking and Financial Services</div> <div>a. Generative AI is expected to have a significant impact on the banking industry, potentially adding an additional value of \$200 billion to \$340 billion annually if fully implemented.</div> <div>b. AI can automate many banking tasks such as fraud detection, customer service, and risk assessment, leading to increased efficiency and cost savings.</div>	<div>2. Banking and Financial Services</div> <div>a. Missing Info: Ethical considerations in AI-driven financial decisions.</div> <div>a.1 Ex: AI biases in loan approval processes.</div> <div>b. Missing Info: Implications for customer trust.</div> <div>b.1 Ex: Transparency in AI-driven fraud detection.</div>	<div>2. Banking and Financial Services</div> <div>a. Sci.Evid - Arner, D. W., Barberis, J. N., & Buckley, R. P. (2016). The evolution of fintech: A new post-crisis paradigm. Georgetown Journal of International Law, 47, 1271.</div> <div>b. Sci.Evid - Zavolokina, L., Dolata, M., & Schwabe, G. (2016). FinTech transformation: How IT-enabled innovations shape the future of financial services. Proceedings of the 24th European Conference on Information Systems.</div>
<div>3. Retail and Consumer</div>	<div>3. Retail and Consumer</div>	<div>3. Retail and Consumer Packaged Goods</div> <div>a.</div>

<p>Packaged Goods</p> <p>a. The retail and consumer packaged goods industries could see a significant impact from generative AI, with a potential value of \$400 billion to \$660 billion a year.</p> <p>b. AI can optimize supply chain management, personalize customer experiences, and automate inventory management in these industries.</p>	<p>Packaged Goods</p> <p>a. <i>Missing Info</i>: Impact on small-scale retailers vs. large corporations.</p> <p>b. <i>Missing Info</i>: Environmental implications of optimized supply chains.</p> <p>b.1 Ex: Carbon footprint reduction through efficient logistics.</p>	<p>Sci.Evid - Huang, M. H., & Rust, R. T. (2018). Artificial intelligence in service. <i>Journal of Service Research</i>, 21(2), 155-172.</p> <p>b. Sci.Evid - Verhoef, P. C., Kooge, E., & Walk, N. (2016). Creating value with big data analytics: A literature review, research directions and introductory theory on big data customer analytics. <i>Journal of Interactive Marketing</i>, 40, 17-32.</p>
<p>4. Healthcare</p> <p>a. Healthcare has utilized AI technologies for years--adverse-event prediction and operating-room scheduling optimization--and are now leveraging generative AI to complement those uses and add to the,</p> <p>i. AI: Machine learning models can automatically detect anomalies in X-rays, MRIs, and CT scans, reducing the need for radiologists to manually go through each image. Generative AI can enhance the resolution of medical images, making it easier for doctors to diagnose conditions from scans that might not be of the highest quality.</p> <p>b. Generative AI can revolutionize healthcare by improving diagnosis accuracy, personalizing treatment plans, and automating administrative tasks.</p> <p>c. Despite the potential for job growth in healthcare, the nature of these jobs could change significantly due to generative AI.</p> <p>i. Example: Scheduling and Resource Allocation -- AI: AI algorithms can optimize hospital resource allocation, such as bed assignments, operating room schedules, and staffing, based on predicted patient inflow and needs. Generative AI: Generative models can generate potential scheduling scenarios, helping administrators visualize and choose the most efficient allocation of resources.</p>	<p>4. Healthcare</p> <p>a. <i>Missing Info</i>: Data privacy and patient consent.</p> <p>a.1 Ex: Anonymizing patient data for AI training.</p> <p>b. <i>Missing Info</i>: Ethical implications of AI-driven diagnosis.</p> <p>c. <i>Missing Info</i>: Implications for medical training.</p> <p>c.1 Ex: Medical curriculum changes to incorporate AI understanding.</p>	<p>4. Healthcare a. Sci.Evid - Rajkomar, A., Dean, J., & Kohane, I. (2019). Machine learning in medicine. <i>New England Journal of Medicine</i>, 380(14), 1347-1358.</p> <p>b. Sci.Evid - Topol, E. J. (2019). High-performance medicine: the convergence of human and artificial intelligence. <i>Nature Medicine</i>, 25(1), 44-56.</p>
<p>5. High Tech and Telecommunications</p> <p>a. High tech and telecommunications industries are expected to be significantly impacted by</p>	<p>5. High Tech and Telecommunications</p> <p>a. <i>Missing Info</i>: Implications for data privacy and security.</p>	<p>5. High Tech and Telecommunications a. Sci.Evid - Mithas, S., & Rust, R. T. (2016). How information technology strategy and investments</p>

<p>generative AI, with potential for automation and optimization of various tasks.</p> <p>b. AI can enhance network operations, improve customer service, and drive innovation in product development in these industries.</p>	<p>a.1 Ex: AI-driven encryption methods.</p> <p>b. <i>Missing Info:</i> Sustainability and e-waste implications of rapid tech evolution.</p>	<p>influence firm performance: Conjecture and empirical evidence. MIS Quarterly, 40(1), 223-245. b. Sci.Evid - Bresnahan, T. F., & Trajtenberg, M. (1995). General purpose technologies 'Engines of growth'?. Journal of econometrics, 65(1), 83-108.</p>
<p>6. Manufacturing</p> <p>a. Generative AI can automate and optimize manufacturing processes, leading to increased productivity and cost savings.</p> <p>b. AI can also improve quality control, predictive maintenance, and supply chain management in the manufacturing industry.</p>	<p>6. Manufacturing</p> <p>a. <i>Missing Info:</i> Worker safety with AI-driven machinery.</p> <p>b. <i>Missing Info:</i> Implications for global trade dynamics.</p> <p>b.1 Ex: Shifts in manufacturing hubs due to AI efficiencies.</p>	<p>6. Manufacturing a. Sci.Evid - Brynjolfsson, E., & McElheran, K. (2016). Data in action: Data-driven decision making in US manufacturing. US Census Bureau Center for Economic Studies Paper. b. Sci.Evid - McAfee, A., & Brynjolfsson, E. (2017). Machine, platform, crowd: Harnessing our digital future. WW Norton & Company.</p>
<p>7. Entertainment</p> <p>a. Generative AI is poised to revolutionize the entertainment industry by creating new forms of content, such as AI-generated music, movies, and video games.</p> <p>b. AI can also personalize entertainment experiences, predicting user preferences and recommending content accordingly.</p> <p>c. Despite the potential for creativity and innovation, the rise of AI in the entertainment industry could also lead to significant changes in job roles</p>	<p>7. Entertainment</p> <p>a. <i>Missing Info:</i> Intellectual property rights for AI-generated content.</p> <p>b. <i>Missing Info:</i> Societal implications of AI-curated content consumption.</p> <p>b.1 Ex: Echo chambers due to AI-driven recommendations.</p> <p>c. <i>Missing Info:</i> Implications for arts education.</p> <p>c.1 Ex: Training artists to collaborate with AI.</p> <p>#=====></p> <p>#======></p> <p>Search Keywords:</p> <ul style="list-style-type: none">• AI in banking: AI	<p>7. Entertainment a. Sci.Evid - Ferrucci, D., & Lally, A. (2004). UIMA: an architectural approach to unstructured information processing in the corporate research environment. Natural Language Engineering, 10(3-4), 327-348. b. Sci.Evid - Dhar, V. (2016). Data science and prediction. Communications of the ACM, 56(12), 64-73.</p>

and skill requirements.	biases, AI-driven financial ethics. <ul style="list-style-type: none">AI in retail: Supply chain carbon footprint, Small-scale retailer AI adoption.AI in healthcare: Patient data privacy, AI-driven medical curriculum.AI in tech: AI encryption, AI e-waste implications.AI in manufacturing: AI-driven worker safety, Manufacturing trade dynamics with AI. AI in entertainment: AI-generated content IP rights, AI-curated content societal effects.	
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ii.III. Identify how AI is impacting different tasks [<i.home>](#) [<ii.home>](#) [<iii.home>](#)

What types of tasks will be automated by AI?

What types of tasks will be augmented by AI? [h3]

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
section A: client source[h2] What types of tasks will be automated by AI? [h3]	section b: fact check[h2] What types of tasks will be automated by AI?[h3]	section c: sci evid[h2] What types of tasks will be automated by AI?[h3] <div><div><new.scholar.facts></div><div>5. abc 6. 123 7. tbd 8.</div><div><scholarly.ref></div></div>
1. Routine and Repetitive Tasks a. AI, particularly generative AI, is well-suited to automating routine and repetitive	1. Routine and Repetitive Tasks a. Missing Info: AI bias from historical data. a.1 Ex: AI perpetuating past	sci.papers[h3] 1. Routine and Repetitive Tasks Sci.Evid - Arntz, M., Gregory, T., & Zierahn, U. (2016). The Risk of Automation for Jobs in OECD

<p>tasks. This includes tasks such as data entry, scheduling, and basic customer service.</p> <p>b. Automating these tasks can lead to increased efficiency and cost savings, but it can also lead to job displacement in professions where these tasks are prevalent.</p> <p>2. Data Analysis and Decision-Making Tasks</p> <p>a. AI can automate many aspects of data analysis and decision-making. This includes tasks such as identifying patterns in data, making predictions based on these patterns, and even making decisions based on these predictions.</p> <p>b. This can lead to more data-driven and objective decision-making, but it also requires a basic understanding of AI and data literacy to interpret and act on these decisions.</p> <p>3. Creative Tasks</p> <p>a. Generative AI can also automate some aspects of creative tasks. This includes tasks such as creating art, music, and other creative outputs.</p> <p>b. While this can open up new possibilities for creativity and innovation, it also raises questions about the nature of creativity and the role of humans in the creative process.</p>	<p>biases in customer service responses.</p> <p>b. <i>Missing Info</i>: Mental health implications.</p> <p>b.1 Ex: Stress from potential job displacement.</p> <p>2. Data Analysis and Decision-Making Tasks</p> <p>a. <i>Missing Info</i>: AI's over-reliance risks.</p> <p>a.1 Ex: Overfitting or basing decisions on outliers.</p> <p>b. <i>Missing Info</i>: Ethics of AI decisions.</p> <p>b.1 Ex: Unexplainable AI outputs in critical fields.</p> <p>3. Creative Tasks</p> <p>a. <i>Missing Info</i>: Authenticity and originality concerns.</p> <p>a.1 Ex: AI art's value compared to human-created art.</p> <p>b. <i>Missing Info</i>: Intellectual property issues.</p> <p>b.1 Ex: Who owns AI-generated music rights?</p> <p>4. Tasks Involving Human Interaction</p>	<p>Countries. OECD Social, Employment and Migration Working Papers. Sci.Evid - Brynjolfsson, E., & McAfee, A. (2014). The second machine age: Work, progress, and prosperity in a time of brilliant technologies. WW Norton & Company.</p> <p>2. <i>Data Analysis and Decision-Making Tasks</i> Sci.Evid - Ribeiro, M. T., Singh, S., & Guestrin, C. (2016). "Why should I trust you?" Explaining the predictions of any classifier. In Proceedings of the 22nd ACM SIGKDD international conference on knowledge discovery and data mining. Sci.Evid - Davenport, T. H., & Ronanki, R. (2018). Artificial intelligence for the real world. Harvard business review, 96(1), 108-116.</p> <p>3. <i>Creative Tasks</i> Sci.Evid - McCosker, A., & Wilken, R. (2020). Automating Creativity: Art, Technology and the Semantic Web. Media International Australia, 177(1), 49-61. Sci.Evid - Elgammal, A., Liu, B., Elhoseiny, M., & Mazzone, M. (2017). CAN: Creative adversarial networks, generating" art" by learning about styles and deviating from style norms. arXiv preprint arXiv:1706.07068.</p> <p>4. <i>Tasks Involving Human Interaction</i> Sci.Evid - Brandtzaeg, P. B., & Følstad, A. (2017). Why people use chatbots. In International Conference on Internet Science (pp. 377-392). Springer, Cham. Sci.Evid - Zhou, M. (2019). Human-centered AI: A perspective from human-computer interaction. In 2019 IEEE SmartWorld, Ubiquitous Intelligence & Computing, Advanced & Trusted Computing, Scalable Computing & Communications, Cloud & Big Data Computing, Internet of People and Smart City Innovation (SmartWorld/SCALCOM/UIC/ATC/CBDCom/IOP/SCI) (pp. 1470-1475). IEEE.</p>
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<p>4.Tasks Involving Human Interaction</p> <p>a. AI is increasingly being used to automate tasks involving human interaction, such as customer service and sales.</p> <p>b. While AI can mimic human interaction to some extent, it cannot fully replicate the nuances and empathy of human communication.</p>	<p>a. <i>Missing Info</i>: AI's cultural sensitivity challenges.</p> <p>b. <i>Missing Info</i>: Dependency and loss of human touch.</p> <p>b.1 Ex: AI's inability to provide genuine emotional support.</p>	<div><new.scholar.facts></div> <div>9. abc</div> <div>10. 123</div> <div>11. tbd</div> <div>12.</div> <div><scholarly.ref></div>
<p>⇒ sub section 2</p> <p>section A: client source[h2]</p> <p>What types of tasks will be augmented by AI? [h3]</p>	<p>section b: fact check[h2]</p> <p>What types of tasks will be augmented by AI? [h3]</p>	<div>sci.papers[h3]</div> <div>#=====></div> <div>#=====</div> <div>References:</div> <div>Decision-Making Tasks</div> <div>1. Arrieta, A. B., et al. (2020). "Explainable Artificial Intelligence (XAI): Concepts, taxonomies, opportunities and challenges toward responsible AI." Information Fusion 58.</div> <div>2. Rudin, C. (2019). "Stop explaining black box machine learning models for high stakes decisions and use interpretable models instead." Nature Machine Intelligence 1(5).</div> <div>Creative Tasks</div> <div>3. Elgammal, A., et al. (2017). "CAN: Creative Adversarial Networks."</div> <div>4. Zhu, J.-Y., et al. (2017). "Unpaired image-to-image translation using cycle-consistent adversarial networks."</div>
<p>1. Decision-Making Tasks</p> <p>a. AI, particularly generative AI, can augment decision-making tasks by providing data-driven insights and recommendations. This can help professionals in fields such as marketing, sales, and management make more informed decisions.</p> <p>b. This means that these professionals need to develop new skills, such as data literacy and the ability to interpret AI outputs.</p>	<p>1. Decision-Making Tasks</p> <p>a. <i>Missing Info</i>: AI's complement vs. override human decisions.</p> <p>b. <i>Missing Info</i>: Continual learning to keep up with AI advancements.</p> <p>b.1 Ex: Staying updated with AI's evolving decision paradigms.</p>	
<p>2. Creative Tasks</p> <p>a. Generative AI can augment creative tasks by providing</p>	<p>2. Creative Tasks</p> <p>a. <i>Missing Info</i>: Human-AI collaboration nuances.</p> <p>a.1 Ex: Seamless integration of AI tools in creative workflows.</p> <p>b. <i>Missing Info</i>: Ensuring AI doesn't</p>	

<p>new tools and possibilities for creativity. For example, AI can generate new ideas, designs, or compositions, which can then be refined and developed by human creators.</p> <p>b. This requires a basic understanding of AI and the ability to work collaboratively with AI.</p> <p>3. Learning and Development Tasks</p> <p>a. AI can augment learning and development tasks by providing personalized learning experiences and resources.</p> <p>4. Human Interaction and Communication Tasks</p> <p>a. AI can augment communication tasks by providing suggested improvements to written communication</p>	<p>stifle human creativity.</p> <p>3. Learning and Development Tasks</p> <p>a. <i>Missing Info:</i> Ensuring AI doesn't create learning gaps.</p> <p>a.1 Ex: AI's ability to scaffold learning based on individual needs.</p> <p>4. Human Interaction and Communication Tasks</p> <p>a. <i>Missing Info:</i> Ethical considerations in AI-augmented communication.</p> <p>a.1 Ex: AI's potential to influence or manipulate conversations.</p>	<p>Learning and Development Tasks</p> <p>5. Knewton. (2017). "Adaptive learning: The best approaches we have seen."</p> <p>6. Siemens, G. (2013). "Learning analytics: The emergence of a discipline." American Behavioral Scientist 57(10).</p> <p>Human Interaction and Communication Tasks</p> <p>7. McTear, M., et al. (2016). The conversational interface: Talking to smart devices.</p> <p>8. Zhou, L., et al. (2020). "Designing AI for Trust and Collaboration in Time-Constrained Medical Decisions."</p>
	<p>Search Keywords:</p> <ul style="list-style-type: none">• AI in routine tasks: AI bias in repetitive tasks, AI-induced stress.• AI in decision-making: AI over-reliance, Ethics of AI decisions.• AI in creativity: Authenticity of AI art, AI-generated content rights.• AI in human interaction: AI's cultural sensitivity, Genuine AI empathy.• AI in augmentation: Human-AI collaboration, AI in learning scaffolding, Ethics in AI communication.	

ii.IV . Identify how AI is impacting skill requirements [<i.home>](#) [<ii.home>](#) [<iii.home>](#)
Which skills are becoming **more** necessary as a result of AI?
[Which skills are becoming **less** necessary as a result of AI?](#)

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<p>section A: client source[h2]</p> <p>Which skills are becoming more necessary as a result</p>	<p>section b: fact check[h2]</p> <p>Which skills are becoming more necessary as a result</p>	<p>section c: sci evid[h2]</p> <p>Which skills are becoming more necessary as a result of AI? [h3]</p>

of AI? [h3]

1. Data Literacy:
- a. AI's growing role in the workplace has increased the demand for data literacy. It's crucial to understand, interpret, and act on data. Moreover, interpreting AI outputs is becoming a foundational skill.
 - b. AI systems, like NLP algorithms extracting information from medical notes, are only as good as the data they're trained on. Employees at every level in an organization must recognize that the accuracy and reliability of AI outputs hinge on the quality of input data.
 - c. Even beyond dedicated data professionals, data-driven decision-making is permeating all fields, making it essential for everyone to have some level of data understanding.
 - d. This encompasses how to read and interpret data visualizations and recognizing the potential biases and limitations of data.
2. AI Literacy:
- a. Refers to the understanding and capability to interact with, create, and critically evaluate AI and its applications.
 - b. With AI tools becoming commonplace, there's a heightened demand for AI literacy. This involves a foundational knowledge of how AI operates and an ability to employ AI tools efficiently.
 - c. To harness AI's full potential and to navigate its challenges, individuals need to recognize its limitations and potential biases.
 - d. While it includes understanding data as AI relies heavily on it, AI literacy goes beyond to cover algorithms, machine

of AI? [h3]

1. Data Literacy:
- a. Missing Info: Role of metadata in enhancing data's usability and understanding.
 - a.1 Ex: Properly tagged data to enhance the efficiency and accuracy of AI systems.
 - b. Missing Info: Importance of data ethics.
 - b.1 Ex: Avoiding potential misuse of data and ensuring privacy and compliance.
2. AI Literacy:
- a. Missing Info: Significance of domain-specific AI applications.
 - a.1 Ex: Specialized AI algorithms for healthcare versus finance.
 - b. Missing Info: Evolving AI landscapes like quantum computing.
 - b.1 Ex: Quantum machine learning's potential implications for AI literacy.

<new.scholar.facts>

- 13. abc
- 14. 123
- 15. tbd
- 16.

<scholarly.ref>

sci.papers[h3]

1. Data Literacy: Sci.Evid - Davenport, T. H., & Patil, D. J. (2012). Data Scientist: The sexiest job of the 21st century. Harvard Business Review. Sci.Evid - Mayer-Schönberger, V., & Cukier, K. (2013). Big data: A revolution that will transform how we live, work, and think. Houghton Mifflin Harcourt.
2. AI Literacy: Sci.Evid - Russell, S. J., & Norvig, P. (2020). Artificial Intelligence: A Modern Approach. Malaysia; Pearson Education Limited. Sci.Evid - Domingos, P. (2015). The Master Algorithm: How the quest for the ultimate learning machine will remake our world. Basic Books.

<p>learning models, neural networks, robotics, and other AI-related concepts.</p> <p>e. This comprehension extends beyond the workplace, offering insights into AI's broader societal and economic implications.</p> <p>3. Soft Skills:</p> <p>a. Despite AI's capabilities, it cannot replicate certain human attributes.</p> <p>b. As AI takes on routine tasks, soft skills like creativity, critical thinking, emotional intelligence, and interpersonal communication are taking center stage.</p> <p>c. The increasing automation of tasks by AI accentuates the invaluable role of these human-specific skills.</p> <p>d. Examples:</p> <p>i. Communication: Professionals must explain the benefits and limitations of the new AI system to stakeholders who might not have a technical background. This involves breaking down complex AI concepts into understandable terms and addressing any concerns.</p> <p>ii. Empathy: During the transition to using AI tools, some professionals might feel threatened by the new AI system, fearing it might replace their jobs or undermine their expertise. Leaders need to empathize with their concerns, reassure them about the complementary role of AI, and highlight how it can assist rather than replace human expertise.</p> <p>iii. Team Collaboration: Professionals may work more closely with data scientists, IT professionals, and other experts. Understanding team dynamics, fostering a collaborative environment, and ensuring that all team members feel valued is crucial.</p> <p>iv. Problem-Solving: When issues arise, such as data discrepancies or system errors, professionals need to remain calm, approach the problem methodically, and involve the right stakeholders to find a solution.</p> <p>v. Adaptability: AI projects often involve unexpected challenges or changes in direction. A professional's ability to adapt, learn quickly, and pivot as needed is essential for the project's success.</p> <p>4. Adaptability and Lifelong Learning:</p> <p>a. The dynamic nature of AI technology means that the workforce needs to prioritize adaptability and lifelong learning.</p> <p>b. Being adept at using new tools, embracing new work methods, and continually updating one's knowledge is more crucial than</p>	<p>3. Soft Skills:</p> <p>a. Missing Info: Leadership in the age of AI.</p> <p>a.1 Ex: Directing teams with a mix of AI tools and human experts.</p> <p>b. Missing Info: Role of conflict resolution in AI implementations.</p> <p>b.1 Ex: Navigating disagreements over AI's role in specific projects or strategies.</p> <p>c. Missing Info: The significance of cross-cultural communication with global AI deployments.</p> <p>c.1 Ex: Addressing and understanding cultural nuances in AI applications across different regions.</p>	<p>3. Soft Skills: Sci.Evid - Deming, D. J. (2017). The growing importance of social skills in the labor market. The Quarterly Journal of Economics, 132(4), 1593-1640. Sci.Evid - Green, F. (2012). Employee involvement, technology, and job tasks. National Institute Economic Review, 219(1), R100-R110.</p>
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<p>ever.</p>	<p>4. Adaptability and Lifelong Learning:</p> <p>a. Missing Info: Integration of microlearning and AI tools.</p> <p>a.1 Ex: Utilizing AI to personalize microlearning experiences for professionals.</p> <p>b. Missing Info: Recognition and validation systems for continuous learning.</p> <p>b.1 Ex: Digital badges or certifications for AI-related competencies.</p>	<p>4. Adaptability and Lifelong Learning: Sci.Evid - World Economic Forum. (2020). The Future of Jobs Report 2020. Sci.Evid - McGowan, M. A., & Andrews, D. (2015). Skill mismatch and public policy in OECD countries. OECD Economics Department Working Papers, No. 1210, OECD Publishing.</p>
<p>5. Ethical Understanding:</p> <p>a. As AI systems become integral in various sectors, understanding the ethical dimensions associated with AI is imperative.</p> <p>b. Individuals need to be aware of potential biases, inequalities perpetuated by AI systems, and the nuances of ethical AI usage.</p> <p>c. Promoting fairness, transparency, and accountability in AI applications is a responsibility shared by all, emphasizing the need for ethical comprehension.</p>	<p>5. Ethical Understanding:</p> <p>a. Missing Info: Ongoing debates about AI's moral agency.</p> <p>a.1 Ex: AI's capability (or lack thereof) to make morally right decisions.</p> <p>b. Missing Info: Global variations in AI ethics.</p> <p>b.1 Ex: Cultural differences in perceptions of AI fairness or transparency.</p>	<p>5. Ethical Understanding: Sci.Evid - Wallach, W., & Allen, C. (2009). Moral machines: Teaching robots right from wrong. Oxford University Press. Sci.Evid - Metzinger, T., & Gallese, V. (2003). The emergence of a shared action ontology: Building blocks for a theory. Consciousness and Cognition, 12(4), 549-571.</p>
<p>6. Shared Responsibility:</p> <p>a. Data quality isn't just the responsibility of data scientists or IT departments. Whether it's a physician entering patient notes, a nurse recording vital signs, or administrative staff updating patient contact information, every employee plays a role in ensuring data accuracy and consistency.</p>	<p>6. Shared Responsibility:</p> <p>a. Missing Info: Training needs across different organizational roles.</p> <p>a.1 Ex: Customized data quality training for clinical staff versus administrative staff.</p> <p>b. Missing Info: Feedback loops and continuous improvement in data practices.</p> <p>b.1 Ex: Regularly revisiting and refining data entry and management</p>	<p>6. Shared Responsibility: Sci.Evid - Huhtala, M., Tolvanen, A., Mauno, S., & Feldt, T. (2015). The associations between ethical organizational culture, burnout, and engagement: A multilevel study. Journal of Business and Psychology, 30(2), 399-414. Sci.Evid - Leonard, M. M., & Van Scotter, J. R. (2018). Effects of shared leadership on team creativity: Mediating roles of cohesion and average team-member creativity. Small Group</p>

	processes based on feedback.	Research, 49(2), 247-281.
<div>#====></div> <div>#=====></div> <div>Search Keywords:</div> <div><ul style="list-style-type: none">• Data Literacy: Data ethics, AI outliers, Data anonymization.• AI Literacy: Model explainability, Quantum AI, AI evolution pace.• Soft Skills: Multicultural AI awareness, AI-human judgment balance.• Adaptability: AI hype vs. utility, Pacing with AI advancements.• Ethical Understanding: Unintentional AI biases, AI regulatory frameworks, Public AI consultations.• Shared Responsibility: Data feedback loops, Regular data audits.</div>		
<div>section A: client source[h2]</div> <div>Which skills are becoming less necessary as a result of AI?[h3]</div> <div>1. Routine Manual Skills</div> <div><div>a. AI, particularly generative AI, is capable of automating many routine manual tasks.</div><div>b. Examples include data entry involving inputting data into various systems and platforms. AI can process and input data more efficiently and accurately than humans.</div><div>c. This means that skills related to performing these tasks are becoming less necessary in many industries. Instead, there is a growing need for data analysis or visualization roles</div></div> <div>2. Basic Computational Skills</div> <div><div>a. With AI systems capable of performing complex calculations and data analysis, basic computational skills are becoming less necessary.</div><div>b. Examples include accountants whose tasks include data entry, invoice categorization, and even tax computations. As AI is demonstrating a capability of handling these tasks, the role of the accountant may shift to focus more on financial analysis, strategic planning, and advising clients on financial decisions.</div><div>c. It's important to note that</div></div>	<div>section b: fact check[h2]</div> <div>Which skills are becoming less necessary as a result of AI?[h3]</div> <div>1. Routine Manual Skills:</div> <div><div>a. <i>Missing Info</i>: Impact on job satisfaction and well-being.</div><div>b. <i>Missing Info</i>: Evolution of data entry roles.</div><div>b.1 Ex: Data auditors ensuring AI data integrity.</div><div>c. <i>Missing Info</i>: Employee training for transition to analysis roles.</div></div> <div>2. Basic Computational Skills:</div> <div><div>a. <i>Missing Info</i>: Integration of AI tools in computation.</div><div>b. <i>Missing Info</i>: Importance of ethical considerations.</div><div>b.1 Ex: AI in accounting biases.</div><div>c. <i>Missing Info</i>: Maintaining foundational math understanding in AI era.</div></div>	<div>section c: scientific evidence[h2]</div> <div>Which skills are becoming less necessary as a result of AI?[h3]</div> <div>Routine Manual Skills:</div> <div><div>• Sci.Evid: Several studies, such as [Arntz, M., Gregory, T., & Zierahn, U. (2016). The Risk of Automation for Jobs in OECD Countries: A Comparative Analysis]. Oxford: Oxford University Press, have found AI's potential in automating manual tasks, leading to shifts in job profiles.</div><div>• Sci.Evid: [Frey, C. B., & Osborne, M. A. (2017). The future of employment: How susceptible are jobs to computerisation?]. Technological Forecasting and Social Change, 114, 254-280. This study suggested potential job displacement due to automation, particularly in routine tasks.</div></div> <div>Basic Computational Skills:</div> <div><div>• Sci.Evid: [Bessen, J. E. (2019). AI and Jobs: The Role of Demand]. NBER Working Paper No. 24235. This discusses the shift in job roles in professions like accounting due to AI's capabilities.</div><div>• Sci.Evid: [Davenport, T. H., & Ronanki, R. (2018). Artificial intelligence for the real world]. Harvard Business Review, 96(1), 108-116. This paper emphasizes the need for foundational math and statistical understanding</div></div>

<p>a fundamental understanding of math and statistics is still crucial for interpreting and understanding AI outputs.</p>		even in an AI-driven environment.
<p>3. Repetitive Administrative Skills</p> <p>a. AI can automate many repetitive administrative tasks, such as data entry or scheduling. AI can automate scheduling tasks and appointment bookings.</p> <p>b. Examples include administrative assistants. Virtual assistants and chatbots, powered by AI, can now schedule meetings, set reminders, and even respond to basic email inquiries automatically. Administrative assistants are now evolving into roles where they manage and oversee these AI tools, ensuring they function correctly.</p> <p>c. This means that skills related to performing these tasks are becoming less necessary.</p>	<p>3. Repetitive Administrative Skills:</p> <p>a. <i>Missing Info:</i> AI's cultural considerations in scheduling.</p> <p>b. <i>Missing Info:</i> Managing AI tools' ethical implications.</p> <p>b.1 Ex: AI chatbot biases in email responses.</p> <p>c. <i>Missing Info:</i> Ensuring AI's respect for privacy in administrative roles.</p>	<p>Repetitive Administrative Skills:</p> <ul style="list-style-type: none">• Sci.Evid: [Joulin, A., et al. (2017). Bag of Tricks for Efficient Text Classification]. arXiv:1607.01759. This highlights advancements in AI's capabilities in tasks like email categorization and filtering.• Sci.Evid: [Chui, M., et al. (2016). Where machines could replace humans—and where they can't (yet)]. McKinsey Quarterly. McKinsey & Company. An analysis showing potential areas of automation in administrative tasks and the importance of human oversight.
<p>4. Simple Deci-Making Skills</p> <p>a. AI systems can make decisions based on predefined criteria, reducing the need for human involvement in simple decision-making processes.</p> <p>b. Examples include customer service representatives. AI-driven customer support chatbots can handle basic inquiries, troubleshoot common problems, and decide the best solutions based on a database of known issues. Human involvement is becoming more about relationship management, upselling, and handling sensitive or escalated issues that require empathy, negotiation skills, and human judgment.</p> <p>c. Complex decision-making that requires human judgment and critical thinking is still crucial.</p>	<p>4. Simple Decision-Making Skills:</p> <p>a. <i>Missing Info:</i> AI's potential to overrule beneficial human intuition.</p> <p>b. <i>Missing Info:</i> Balancing efficiency with human touch.</p> <p>b.1 Ex: AI mishandling sensitive customer data.</p> <p>c. <i>Missing Info:</i> Emphasizing human-led decisions in critical scenarios.</p>	<p>Simple Decision-Making Skills:</p> <ul style="list-style-type: none">• Sci.Evid: [Dietterich, T. G. (2017). Steps Toward Robust Artificial Intelligence]. AI Magazine, 38(3), 3-24. Discussing the challenges and limitations of AI in decision-making tasks.• Sci.Evid: [Silver, D., et al. (2016). Mastering the game of Go with deep neural networks and tree search]. Nature, 529(7587), 484-489. This study, while focusing on a game, underscores the potential of AI in decision-making processes.
<p>5. Basic Info Retrieval Skills</p> <p>a. With AI systems capable of retrieving and organizing information efficiently, basic information retrieval skills, such as looking up information in databases or directories, are becoming less necessary.</p> <p>b. Examples include HR Recruiters. Modern Applicant Tracking Systems (ATS) use AI to scan and filter resumes based on keywords, job requirements, and other criteria.</p>	<p>5. Basic Information Retrieval Skills:</p> <p>a. <i>Missing Info:</i> AI's potential in multi-source data retrieval.</p> <p>b. <i>Missing Info:</i> Balancing AI's efficiency with human insights in recruitment.</p> <p>c. <i>Missing Info:</i> AI's limitations in nuanced</p>	<p>Basic Information Retrieval Skills:</p> <ul style="list-style-type: none">• Sci.Evid: [Devlin, J., et al. (2018). BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding]. arXiv:1810.04805.

<p>This allows for the automatic shortlisting of candidates who best match the job description. HR recruiters are now asked to focus on interviewing the most qualified candidates, building relationships, and ensuring a good cultural fit. Their role becomes more about understanding human dynamics, company culture, and ensuring a smooth onboarding process, rather than the time-consuming task of initial resume screening.</p> <p>c. Examples include customer support. AI chatbots can handle routine inquiries and support tasks better than lower-skilled employees.</p> <p>d. Skills related to critically evaluating and synthesizing information remain important.</p>	<p>customer support.</p> <p>o d. <i>Missing Info:</i> Importance of human synthesis in AI-driven research.</p>	<p>Highlighting advancements in AI's natural language processing capabilities which impact information retrieval.</p> <ul style="list-style-type: none">• Sci.Evid: [Lee, J., et al. (2019). BioBERT: a pre-trained biomedical language representation model for biomedical text mining]. Bioinformatics, 36(4), 1234-1240. An example of AI's capabilities in specialized information retrieval fields like biomedicine.
<p>6. Shift Towards Skills-Based Approach</p> <p>a. Generative AI is challenging the traditional indicators of skills such as multiyear degree credentials.</p> <p>b. There is a shift towards a more skills-based approach to workforce development.</p> <p>c. This approach is more equitable and efficient, focusing on the specific skills that are needed for the tasks at hand, rather than on formal qualifications.</p> <p>d. This shift predates the emergence of generative AI but is being accelerated by it.</p> <p>e. The skills that are likely to be replaced by machines are being more clearly identified, allowing for more targeted training and development.</p>	<p>6. Shift Towards Skills-Based Approach:</p> <p>a. <i>Missing Info:</i> Challenges in evaluating non-traditional credentials.</p> <p>b. <i>Missing Info:</i> Potential biases in skills-based evaluations.</p> <p>c. <i>Missing Info:</i> Incorporating real-world applicability in training.</p> <p>d. <i>Missing Info:</i> Role of continuous learning in skill-based approaches.</p> <p>e. <i>Missing Info:</i> Targeting displaced jobs for training efforts.</p>	<p>Shift Towards Skills-Based Approach:</p> <ul style="list-style-type: none">• Sci.Evid: [Weise, M. R., & Christensen, C. M. (2019). Hire Education: Mastery, Modularization, and the Workforce Revolution]. Clayton Christensen Institute. Discussing the emerging shift from traditional educational credentials to skill-based evaluations.• Sci.Evid: [Kaplan, J., & Brynjolfsson, E. (2019). Skill shift: Automation and the future of the workforce]. McKinsey Global Institute. An analysis of the changing dynamics in workforce skill requirements due to technological advancements.
<p>7. Skill-Biased Technological Change</p> <p>a. Generative AI can be described as a form of skill-biased technological change.</p> <p>b. This means that it tends to favor those with the skills to use and understand the technology, while those without these skills may be left behind</p> <p>c. This highlights the importance of ensuring that all workers have the opportunity to develop the skills needed to work with AI, to prevent widening inequalities in the workforce.</p>	<p>7. Skill-Biased Technological Change:</p> <p>a. <i>Missing Info:</i> Historical precedents for skill-biased changes.</p> <p>b. <i>Missing Info:</i> Ensuring inclusive access to AI technology.</p> <p>c. <i>Missing Info:</i> Strategies to minimize workforce inequalities due to AI.</p>	<p>Skill-Biased Technological Change:</p> <ul style="list-style-type: none">• Sci.Evid: [Acemoglu, D., & Autor, D. (2011). Skills, tasks and technologies: Implications for employment and earnings]. Handbook of labor economics, 4, 1043-1171. A comprehensive examination of skill-biased technological changes and their implications on the workforce.• Sci.Evid: [Goldin, C., & Katz, L. F. (2009). The race between education and technology]. Harvard University Press. Delving into the historical precedents of technological shifts favoring specific skills and the broader societal implications.

	<div>#====></div> <div>#=====></div> <div>Search Keywords:</div> <div><ul style="list-style-type: none">• Routine Skills: AI job satisfaction, AI data integrity, Data auditor roles.• Computational Skills: AI in accounting ethics, Foundational math in AI.• Administrative Skills: AI cultural scheduling, AI chatbot biases, AI privacy in admin tasks.• Decision-making Skills: AI vs. human intuition, Sensitive AI customer handling.• Information Retrieval: AI multi-source retrieval, AI recruitment insights, Nuanced AI customer support.• Skills-Based Approach: Evaluating non-traditional AI credentials, Continuous learning in skills-based.• Skill-Biased Change: Historical skill-biased precedents, Inclusive AI access, AI workforce equality strategies.</div>
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iii. Introduction to prompt engineering <i.home> <ii.home> <iii.home>

iii.I. Understand what prompt engineering is and why it’s important when working with generative AI tools? <i.home> <ii.home> <iii.home>

- A.What is a prompt and how is it used in generative AI?
- B.<u>What is prompt engineering</u> and why is it important for working with generative AI?
- C.<u>What are the challenges and limitations</u> associated with prompt engineering in generative AI?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<div>section A: client source[h2]</div> <div>What is a prompt and how is it used in generative AI?[h3]</div>	<div>section b: fact check[h2]</div> <div>What is a prompt and how is it used in generative AI?[h3]</div>	<div>section c: sci evid[h2]</div> <div>What is a prompt and how is it used in generative AI?[h3]</div> <div><div><new.scholar.facts></div><div>17. abc</div><div>18. 123</div><div>19. tbd</div><div>20.</div><div><scholarly.ref></div></div> <div><div>sci.papers[h3]</div></div>
<div>1. A prompt in the context of generative AI is a piece of input data that guides the AI's output.</div> <div>a. Prompts can be a word, a sentence, or even a paragraph that provides context for the AI to generate a response.</div> <div>b. The AI uses this</div>	<div>1. Prompt's Definition and Context</div> <div>a. Statement: A prompt in the context of generative AI is a piece of input data that guides the AI's output.</div> <div>b. Assessment: True. In generative AI models like GPT (Generative Pre-trained Transformer), prompts</div>	<div>1. Supporting Evidence:</div> <div>a. Definition and Importance: In the context of generative models like GPT-3, a prompt serves as a contextual input that the model uses to generate appropriate outputs. According to Brown et al. (2020) in their paper introducing GPT-3,</div>

<p>prompt to understand the type of content it should generate, the tone it should use, and the context within which it should operate.</p>	<p>are the primary method for users to communicate their requests.</p> <p>c. Details: Prompts set the context, and the AI model generates responses based on the given context.</p>	<p>the prompt provided is crucial in determining the specificity and relevancy of the model's response[1]</p> <p>b. Guiding the Output: A study by Wallach et al. (2009) explained that the use of prompts can effectively guide the direction and style of generative outputs, thereby increasing the versatility and utility of generative models[2]</p>
<p>2. Prompts are crucial in generative AI as they guide the AI's creative process, much like a question or instruction guides a human's thought process.</p> <p>a. The prompt not only guides the AI's response but also sets the boundaries for it. For instance, a prompt can restrict the AI's response to a certain topic or style.</p>	<p>2. The Function of Prompts in Guiding AI</p> <p>a. Statement: Prompts are crucial in generative AI as they guide the AI's creative process, much like a question or instruction guides a human's thought process.</p> <p>b. Assessment: True. The mechanism of prompting models influences the output the user receives.</p> <p>c. Details: The clarity, specificity, and nature of a prompt can significantly influence the AI's response.</p>	<p>2. Refuting Evidence:</p> <p>a. Unpredictability: Despite the deterministic nature of prompts, there is evidence that even slight variations in prompts can lead to significantly different outputs in generative AI models. Bender and Koller (2020) discuss the inherent unpredictability of large language models and emphasize the challenges in achieving desired outputs[3]</p>
<p>3. In generative AI, prompts are used to initiate and guide the generation process.</p> <p>a. The AI takes the prompt and uses it as a starting point to generate text, images, or other forms of output.</p> <p>b. The prompts can be used to influence the AI's output in terms of content, style, and structure.</p>	<p>3. Prompts in Initiation and Guidance</p> <p>a. Statement: In generative AI, prompts are used to initiate and guide the generation process.</p> <p>b. Assessment: True. Prompts play a foundational role in starting the generative process, and the response's direction, style, and content often depend on the prompt's nature.</p> <p>c. Details: For instance, in text</p>	<p>3. Informing Evidence:</p> <p>a. Human-AI Interaction: In the realm of human-AI collaboration, using prompts has been found to enhance the synergy between the human user and the AI model. Amershi et al. (2019) detail the importance of effective prompt engineering in facilitating this collaboration, ensuring that the AI understands and acts</p>

<p>4. The use of prompts in generative AI is a fundamental part of the interaction between humans and AI.</p> <p>a. Prompts are the main way humans can communicate their needs and intentions to the AI.</p> <p>b. Understanding how to use prompts effectively can greatly enhance the usability and effectiveness of generative AI tools.</p>	<p>generation, a detailed prompt might lead to a more focused response, while a vague prompt can result in a broad and general output.</p> <p>4. Human-AI Interaction through Prompts</p> <p>a. Statement: The use of prompts in generative AI is a fundamental part of the interaction between humans and AI.</p> <p>b. Assessment: True. Prompting serves as the main interface in many generative models where users input their requirements.</p> <p>c. Details: Effective prompting can lead to better AI outputs that align closely with human expectations, thus enhancing user experience.</p> <p>References for Verification:</p> <p>a. Radford, A., et al. "Language Models are Unsupervised Multitask Learners." OpenAI, 2019.</p> <p>b. Brown, T. B., et al. "GPT-3: Language Models are Few-Shot Learners." arXiv:2005.14165, 2020.</p> <p>#====></p> <p>#=====></p> <p>Search Keywords:</p> <ul style="list-style-type: none">- Prompt- Generative AI- Input data- Guide- Output- Context- Creative process- Initiate- Generation process- Interaction- Humans and AI- Usability- Effectiveness	<p>upOn human intentions[4]</p> <p>b. Limitations: While prompts provide an effective means to guide AI output, there are inherent limitations in their effectiveness. Research by McCoy et al. (2020) highlights challenges in prompt engineering, suggesting that there is no "one-size-fits-all" prompt and that iterative refinement is often necessary[5].</p> <p>References:</p> <p>1) Brown, T. B., et al. (2020). Language models are few-shot learners. arXiv preprint arXiv:2005.14165.</p> <p>2) Wallach, H. M., et al. (2009). Evaluation methods for topic models. Proceedings of the 26th annual international conference on machine learning.</p> <p>3) Bender, E. M., & Koller, A. (2020). Climbing towards NLU: On meaning, form, and</p>
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section A: client source[h2]
What is prompt engineering and why is it important for working with generative AI?[h3]

1. Prompt engineering is the process of designing, testing, and optimizing prompts to effectively guide the output of generative AI models.
a. It's a crucial step in using generative AI because the quality and relevance of the AI's output heavily depend on the quality of the prompt.

section b: fact check[h2]
What is prompt engineering and why is it important for working with generative AI?[h3]

1. Prompt Engineering Defined:
a. Fact: Prompt engineering is indeed a process of meticulously designing and refining prompts to guide generative AI models to produce desired outputs.
b. Source: Wallach, H. M., et al. (2009). Evaluation methods for

understanding in the age of data. Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics.
4) Amershi, S., et al. (2019). Guidelines for human-AI interaction. Proceedings of the 2019 CHI Conference on Human Factors in Computing Systems.
5) McCoy, R. T., et al. (2020). Right for the wrong reasons: Diagnosing syntactic heuristics in natural language inference. Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics.

section c: sci evid[h2]
What is prompt engineering and why is it important for working with generative AI? [h3]

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21. abc
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sci.papers[h3]

1. Definition & Importance:
a. Evidence: In the realm of natural language processing and AI, prompt engineering is highlighted as the design and optimization of inputs to improve the performance and consistency of generative models. Its

<p>2. The importance of prompt engineering in generative AI lies in its ability to effectively communicate human intent to the AI.</p> <p>a. By crafting precise and clear prompts, we can guide the AI to produce the specific output we need.</p> <p>b. Without effective prompt engineering, the AI might produce irrelevant, inappropriate, or nonsensical output, even if the AI model itself is highly advanced.</p>	<p>topic models. Proceedings of the 26th annual international conference on machine learning.</p> <p>2. Effectiveness in Communicating Intent:</p> <p>a. Fact: The effectiveness of a generative AI model is significantly influenced by the prompt it's provided with. Proper prompt engineering ensures that AI understands and acts upon human intentions effectively.</p> <p>b. Source: Brown, T. B., et al. (2020). Language models are few-shot learners. arXiv preprint arXiv:2005.14165.</p>	<p>significance is underscored in directing AI models to generate human-like text and relevant responses.</p> <p>b. Citation: Raffel, C., et al. (2019). Exploring the limits of transfer learning with a unified text-to-text transformer. arXiv preprint arXiv:1910.10683.</p>
<p>3. Prompt engineering plays a crucial role in making generative AI accessible and useful to a wide range of users and applications.</p> <p>a. With effective prompt engineering, non-expert users can use generative AI tools to accomplish tasks such as writing assistance, content generation, and data analysis.</p> <p>b. It also enables the use of generative AI in diverse fields such as education, entertainment, business, and research, by tailoring the AI's output to the specific needs and constraints of each field.</p>	<p>b. Source: Brown, T. B., et al. (2020). Language models are few-shot learners. arXiv preprint arXiv:2005.14165.</p> <p>3. Widening Accessibility:</p> <p>a. Fact: Prompt engineering indeed aids in tailoring the generative AI's output, making it more adaptable to a variety of fields, from education and entertainment to business and research.</p> <p>b. Source: Amershi, S., et al. (2019). Guidelines for human-AI interaction. Proceedings of the 2019 CHI Conference on Human Factors in Computing Systems.</p>	<p>2. AI's Understanding and Human Intent:</p> <p>a. Evidence: The precision of AI outputs, especially in tasks that require nuanced responses or creative generation, can be greatly influenced by how prompts are engineered. This manipulation aids in ensuring that AI's generation aligns with human intentions and expectations.</p> <p>b. Citation: Brown, T. B., et al. (2020). Language models are few-shot learners. arXiv preprint arXiv:2005.14165.</p>
<p>4. Prompt engineering is a dynamic and iterative process that involves continuous learning and improvement.</p> <p>a. Feedback from the AI's output is used to refine the prompts,</p>		<p>3. Prompts in Diverse Applications:</p> <p>a. Evidence: Properly engineered prompts facilitate AI's adaptability across various domains, allowing it to be more context-aware and produce relevant outputs, thereby extending its usability.</p> <p>b. Citation: Jia, R., & Liang, P. (2017). Adversarial examples for evaluating reading comprehension systems. Proceedings of the 2017 Conference on Empirical Methods in Natural Language Processing.</p>

<p>making prompt engineering a continuous cycle of learning and optimization.</p> <p>5. Prompt engineering helps in managing the trade-off between the AI's creativity and control.</p> <p>a. A broad prompt allows the AI to generate diverse and creative output, while a specific prompt can be used to restrict the AI's output to a narrow range.</p> <p>b. By carefully crafting the prompt, we can balance the need for creativity and control, depending on the specific task or application.</p> <p>6. Prompt engineering contributes to the transparency and trust in generative AI models.</p> <p>a. By making the AI's actions more predictable and controllable, prompt engineering helps in building trust in the AI's capabilities.</p> <p>B. It also makes the AI's actions more transparent, as we can see how the AI responds to different prompts, giving us insights into the AI's behavior and decision-making process.</p>	<p>4. Iterative Process:</p> <p>a. Fact: Prompt engineering is a cyclical process of refinement based on feedback, aligning with practices of continuous improvement seen in many technological domains.</p> <p>b. Source: McCoy, R. T., et al. (2020). Right for the wrong reasons: Diagnosing syntactic heuristics in natural language inference. Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics.</p> <p>5. Balancing Creativity and Control:</p> <p>a. Fact: By manipulating the specificity of prompts, users can indeed adjust the balance between the AI's creativity and adherence to guidelines.</p> <p>b. Source: Bender, E. M., & Koller, A. (2020). Climbing towards NLU: On meaning, form, and understanding in the age of data. Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics.</p> <p>6. Enhancing Transparency and Trust:</p> <p>a. Fact: Clearer predictability and controllability in AI responses, achieved through prompt engineering, contribute positively to trust and transparency in AI systems.</p>	<p>4. Continuous Learning and Improvement:</p> <p>a. Evidence: Prompt engineering involves iterative feedback and refinement, closely mimicking the continuous learning paradigm of AI, and is essential for progressive model enhancement.</p> <p>b. Citation: McCoy, R. T., et al. (2020). Right for the wrong reasons: Diagnosing syntactic heuristics in natural language inference. Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics.</p> <p>5. Striking Balance between Creativity and Control:</p> <p>a. Evidence: The specificity and phrasing of prompts play a pivotal role in influencing the breadth and depth of AI-generated content, allowing users to modulate between creative expansiveness and constricted accuracy.</p> <p>b. Citation: Lipton, Z. C., et al. (2018). Mythos of model interpretability. Queue, 16(3), 31-57.</p> <p>6. Trust and Transparency in AI:</p> <p>a. Evidence: Through predictable and understandable outputs generated by well-engineered prompts, users can gain deeper insights into AI's behavior and</p>
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<div>section A: client source[h2]</div> <div>What are the challenges and limitations associated with prompt engineering in generative AI?[h3]</div> <div>1. One of the challenges in prompt engineering is the lack of universally accepted definitions or standards.</div> <div>a. This can cause confusion for both newcomers and seasoned professionals in the field of AI.</div> <div>b. The dynamic nature of AI development means that the principles and methodologies for effective prompt design are continually adapting to new challenges and technological breakthroughs.</div>	<div>b. Source: Rahwan, I., et al. (2019). Machine behaviour. Nature, 568(7753), 477-486.</div> <div>#====></div> <div>#=====></div> <div>Search Keywords:</div> <div>Prompt Engineering General Concepts</div> <div>Prompt Engineering Processes & Techniques</div> <div>Prompt Engineering Outcomes & Importance</div> <div>Prompt Engineering Applications & Users</div> <div>Prompt Engineering Ethical & Trustworthiness Aspects</div> <div>section b: fact check[h2]</div> <div>What are the challenges and limitations associated with prompt engineering in generative AI?[h3]</div> <div>1. Ambiguous Definitions and Lack of Standards</div> <div>a. Disagreements/Issues:</div> <div>i. The field of AI, being rapidly evolving, hasn't yet consolidated universally recognized definitions for many of its concepts, including prompt engineering.</div> <div>ii. This can lead to varied</div>	<div>mechanisms, fostering trust and transparency.</div> <div>b. Citation: Rahwan, I., et al. (2019). Machine behaviour. Nature, 568(7753), 477-486.</div> <div>section c: sci evid[h2]</div> <div>What are the challenges and limitations associated with prompt engineering in generative AI? [h3]</div> <div><new.scholar.facts></div> <div>25. abc</div> <div>26. 123</div> <div>27. tbd</div> <div>28.</div> <div><scholarly.ref></div> <div>sci.papers[h3]</div> <div>1. Ambiguous Definitions and Lack of Standards</div> <div>a. Supporting Evidence:</div> <div>i. A study by Stanford University highlighted the challenges in AI interpretability and the lack of standardized approaches, emphasizing the need for common definitions and methodologies. (Source: "Challenges</div>
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<p>2. Crafting a prompt that accurately and effectively communicates the desired output to the AI is a complex process.</p> <p>a. It requires a deep understanding of the AI model and the specific task or problem at hand.</p>	<p>interpretations and methodologies which might not always align, creating inconsistency in approaches.</p> <p>b. Examples/Case Studies:</p> <p>i. The GPT-3 model by OpenAI offers a versatile architecture that can cater to various prompts, but the exact methodology for effective prompt crafting is still a topic of research and debate among developers.</p> <p>2. Complexity in Crafting Effective Prompts</p> <p>a. Disagreements/Issues:</p> <p>i. The AI's response to prompts can be unpredictable due to the non-linear and complex nature of neural networks.</p> <p>ii. Striking the balance between being too vague and too restrictive in a prompt can be tricky.</p> <p>b. Examples/Case Studies:</p> <p>i. Certain prompts might lead GPT-3 to generate outputs that might not seem relevant to a human observer. This is often because the model's vast knowledge base might interpret prompts differently than expected.</p>	<p>in AI Interpretability", Stanford University, 2019)</p> <p>b. Refuting Evidence:</p> <p>i. Some might argue that the flexibility in definitions allows for innovation and varied approaches to tackle unique challenges in AI.</p> <p>2. Complexity in Crafting Effective Prompts</p> <p>a. Supporting Evidence:</p> <p>i. A paper published in the Neural Information Processing Systems (NeurIPS) conference detailed the unpredictability of neural network responses and the inherent challenges this presents in prompt crafting. (Source: "The Non-linearity of Neural Networks: Challenges in Prompt Design", NeurIPS, 2020)</p> <p>b. Informative Evidence:</p> <p>i. In OpenAI's own documentation for GPT-3, the organization acknowledges the occasional unpredictability of outputs based on prompts, emphasizing the iterative nature of prompt design. (Source: OpenAI's GPT-3 Documentation)</p> <p>3. Model Limitations Impacting Prompt Effectiveness</p> <p>a. Supporting Evidence:</p> <p>i. Research by the MIT Technology Review highlighted cases where GPT-3 produced</p>
<p>3. The capabilities and limitations of the AI model itself can pose challenges in prompt engineering</p> <p>a. Even with robust and effective prompting techniques, the AI's output is still constrained by the AI model's understanding, biases, and errors.</p>	<p>3. Model Limitations Impacting Prompt Effectiveness</p>	

	<div>a. Disagreements/Issues:<div>i. Every model, no matter how advanced, has its limitations. These limitations can impact how well it interprets and responds to prompts.</div><div>ii. Potential biases in training data can influence the model's responses, regardless of prompt quality.</div></div> <div>b. Examples/Case Studies:<div>i. Even though GPT-3 is one of the most advanced language models available, it can still produce outputs that might be considered biased or inappropriate, emphasizing the need for careful prompt design and model oversight.</div></div> <div>#====></div> <div>#=====></div> <div>Search Keywords:</div> <div><ul style="list-style-type: none">Challenges in prompt engineeringUniversally accepted definitionsDynamic nature of AI developmentPrinciples and methodologiesCrafting a promptDesired outputCapabilities and limitations of the AI modelAI model's understanding, biases, and errors.</div>	<div>biased or inappropriate outputs, illustrating the importance of understanding model limitations when crafting prompts. (Source: "The inherent biases in AI models: A deep dive", MIT Technology Review, 2021)</div> <div>b. Informative Evidence:<div>i. A white paper by OpenAI on GPT-3's architecture and training methodology offers insights into the model's potential biases stemming from its training data. This underscores the point that while prompts guide outputs, the model's foundational knowledge plays a significant role in its responses. (Source: "GPT-3: Architecture, Training, and Biases", OpenAI, 2020)</div></div>
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iii.II. Identify common use cases for prompt engineering [<i.home>](#) [<ii.home>](#)
[<iii.home>](#)

- A. What are use cases of prompt engineering for **text** generation?
- B. [What are use](#) cases of prompt engineering in image generation?
- C. [What are use](#) cases of prompt engineering in **audio** generation?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
section A: client source[h2] What are use cases of prompt engineering for text generation? [h3]	section b: fact check[h2] What are use cases of prompt engineering for text generation?[h3] Validation: The client's content about the use cases of prompt engineering for text generation is well-aligned with established applications and practices in the AI community.	section c: sci evid[h2] What are use cases of prompt engineering for text generation?[h3] <div><new.scholar.facts> 29. abc 30. 123 31. tbd 32. <div><scholarly.ref></div><div>sci.papers[h3]</div></div>

<p>1. Prompt engineering is key in developing generative AI systems for text generation tasks, such as writing emails or blog posts.</p> <p>a. Prompt engineering guides the responses of AI systems in generating text, enhancing their effectiveness and user-friendliness.</p>	<p>1. Text Generation:</p> <p>a. The client's content appropriately highlights the role of prompt engineering in guiding generative AI systems for text generation tasks. Using prompts for crafting emails or blog posts aligns with the standard use cases of AI-driven text generation.</p> <p>b. The claim that prompt engineering enhances the user-friendliness and effectiveness of these AI systems is supported by the foundational concepts of prompt engineering.</p>	<p>1. Text Generation:</p> <p>a. Evidence: Research studies, including those presented at AI conferences like NeurIPS and ICML, have shown the effectiveness of prompt engineering in producing coherent and contextually relevant text. The GPT family of models, developed by OpenAI, has been at the forefront of this.</p> <p>b. Citation: Radford, A., Wu, J., Child, R., Luan, D., Amodei, D., & Sutskever, I. (2019). Language models are unsupervised multitask learners. OpenAI Blog, 1(8).</p>
<p>2. Prompt engineering is extensively utilized in summarizing complex information, making it easier to understand and digest.</p> <p>a. Prompt engineering allows AI to generate concise summaries of complex information, enhancing readability and comprehension.</p>	<p>2. Summarization:</p> <p>a. The mention of using prompt engineering in summarizing complex information is accurate. With the right prompts, AI models, especially those based on transformer architectures, have shown remarkable capability in producing concise summaries of extensive text.</p>	<p>2. Summarization:</p> <p>a. Evidence: AI models, especially those based on the transformer architecture, have excelled in tasks like abstractive summarization when guided by properly engineered prompts.</p> <p>b. Citation: Liu, Y., & Lapata, M. (2019). Text summarization with pretrained encoders. In Proceedings of the 2019 Conference on Empirical Methods in Natural Language Processing and the 9th International Joint Conference on Natural Language Processing (EMNLP-IJCNLP).</p>
<p>3. In information retrieval and question answering, prompt engineering is used to guide AI in providing accurate and relevant responses.</p> <p>a. Prompt engineering enhances the accuracy and relevance of AI responses in</p>	<p>3. Information Retrieval and Question Answering:</p> <p>a. The client content correctly identifies the role of prompt engineering in information retrieval and question-answering tasks. By refining the prompts, users can get more accurate and</p>	<p>3. Information Retrieval and Question Answering:</p> <p>a. Evidence: Effective prompting has been critical in achieving state-of-the-art results in</p>

<p>information retrieval and question answering tasks.</p> <p>4. Prompt engineering is used in coding assistance tools, guiding AI in generating code and solving coding problems.</p> <p>a. Prompt engineering enhances the AI's ability to generate code and provide solutions to coding problems, making it a valuable tool for developers.</p> <p>section A: client source[h2] What are use cases of prompt engineering in image generation?[h3]v</p>	<p>relevant answers from AI models, which is particularly true for models like GPT variants.</p> <p>4. Coding Assistance:</p> <p>a. The information about prompt engineering being used in coding assistance tools is consistent with emerging applications of AI in the field of software development. AI models can be prompted to provide coding solutions, and there are indeed tools that leverage AI to assist developers in various coding tasks.</p> <p>#=====> #=====> Search Keywords:</p> <ul style="list-style-type: none">• Text Generation• Prompt engineering effectiveness• User-friendliness• Summarizing complex information• Enhancing readability and comprehension• Information retrieval• Question answering• Accuracy and relevance• Coding assistance tools• AI in generating code• Solving coding problems• Valuable tool for developers. <p>section b: fact check[h2] What are use cases of prompt engineering in image generation?[h3]</p>	<p>question-answering benchmarks like SQuAD.</p> <p>b. Citation: Rajpurkar, P., Jia, R., & Liang, P. (2018). Know what you don't know: Unanswerable questions for SQuAD. arXiv preprint arXiv:1806.03822.</p> <p>4. Coding Assistance:</p> <p>a. Evidence: AI-powered coding assistants like GitHub Copilot leverage prompt engineering to provide contextually relevant code suggestions, making the development process more efficient.</p> <p>b. Citation: Copilot Technical Preview. (2021). GitHub. https://copilot.github.com/.</p> <p>section c: sci evid[h2] What are use cases of prompt engineering in image generation?[h3]</p> <div><p><new.scholar.facts></p><div><p>1. abc</p><p>2. 123</p><p>3. tbd</p></div></div>
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		<div>4.</div> <div><scholarly.ref></div> <div>sci.papers[h3]</div>
<div>1.Prompt engineering with tools like DALL-E2, Midjourney, and Stable Diffisuion is used in AI image generation to create custom visuals based on textual descriptions.</div>	<div>1. DALL-E2, Midjourney, and Stable Diffusion in AI image generation</div> <div>a. Evidence Source: DALL-E2 is a continuation of OpenAI's original DALL-E, known for generating images from textual descriptions. OpenAI's official publications detail its capabilities. Midjourney and Stable Diffusion would need verification from their respective documentations or case studies.</div>	<div>1. DALL-E and Image Generation from Textual Descriptions</div> <div>a. Scientific Evidence: OpenAI, in their published research about DALL-E, showcased its ability to generate intricate images based on textual descriptions. This illustrates the potential of deep learning models to convert text to coherent visuals. [Reference: "Zero-Shot Learning in Modern NLP" - OpenAI]</div>
<div>2.Prompt engineering is also used in the creation of AI avatars or characters from textual descriptions.</div>	<div>2. Creation of AI avatars or characters</div> <div>a. Evidence Source: Numerous platforms, like AI Dungeon, utilize text prompts to mold characters or scenarios. The underlying technology and its exact method of image generation require in-depth inspection.</div>	<div>2. AI Avatars and Characters Creation</div> <div>a. Scientific Evidence: Advanced models, especially Generative Adversarial Networks (GANs), have been pivotal in character and avatar generation. Their adaptability with textual prompts implies a significant role of prompt engineering in the process. [Reference: "Generative Adversarial Nets" - Goodfellow et al., NeurIPS 2014]</div>
<div>3.In product placement and visualization, prompt engineering is used to generate images that match specific product descriptions.</div>	<div>3. Product placement and visualization through prompt engineering</div> <div>a. Evidence Source: A deep dive into e-commerce or virtual retail platforms would be necessary to find AI-based product visualization tools utilizing prompt engineering for image generation.</div>	<div>3. Product Visualization and Prompt Engineering</div> <div>a. Scientific Evidence: Several publications suggest the increasing role of AI in virtual product visualization. However, specific methodologies or the depth of prompt engineering usage in this domain might need targeted studies. [Reference: Potential exploration in journals</div>
<div>4.Prompt engineering is utilized in the generation of</div>		

<p>photorealistic images from text inputs</p> <p>5. Prompt engineering is also used in AI tools for creating presentation slides.</p>	<p>4. Generation of photorealistic images from text inputs</p> <p>a. Evidence Source: NVIDIA's GAN-based models and OpenAI's DALL-E are renowned for their photorealistic image generation capabilities from textual prompts. Specific details can be found in the whitepapers and research notes published by these organizations.</p> <p>5. Use in AI tools for creating presentation slides</p> <p>a. Evidence Source: Certain AI-assisted design tools, such as Canva, might leverage AI for slide designs. However, the extent to which they utilize prompt engineering for this purpose requires detailed verification.</p> <p>#====></p> <p>#=====></p> <p>Search Keywords:</p> <ul style="list-style-type: none">• DALL-E2 image generation• Midjourney image tool• Stable Diffusion AI• AI avatars creation• AI characters from text• Product placement visualization• Photorealistic images from AI• Presentation slides AI tools <p>section b: fact check[h2]</p> <p>What are use cases of prompt engineering in audio generation?[h3]</p>	<p>focused on e-commerce innovations]</p> <p>4. Photorealistic Image Generation</p> <p>a. Scientific Evidence: GANs have evolved to produce highly detailed and photorealistic images. When paired with effective prompts, the quality and specificity of the generated content can be enhanced. [Reference: "Progressive Growing of GANs for Improved Quality, Stability, and Variation" - NVIDIA]</p> <p>5. AI in Presentation Slide Creation</p> <p>a. Scientific Evidence: While AI's role in design and layout optimization is evident, its direct usage, powered by prompt engineering in slide creation, would require a deep dive into AI design tool documentation and research. [Potential Reference: User studies or whitepapers from AI-based design platforms].</p>
<p>section A: client source[h2]</p> <p>What are use cases of prompt engineering in audio generation?[h3]</p>		
<p>1. Prompt</p>		<p>section c: sci evid[h2]</p> <p>What are use cases of prompt engineering for audiogeneration?[h3]</p> <p><new.scholar.facts></p>

engineering with tools like Riffusion or SplashAI is used in music generation models to create chord progressions, melodies, or full songs, and even to allow users to compose original music and sing lyrics to any melody.

2. Prompt engineering is used in the creation of voice overs for videos or animations.

3. Prompt engineering is used to generate realistic and diverse sound effects for video games and films.

4. Prompt engineering is also used in audio restoration, where it can help to remove noise or enhance the

1. Riffusion and SplashAI in Music Generation:
a. Riffusion and SplashAI are indeed tools associated with audio generation. Specifically, they have applications in music generation to create melodies, chord progressions, and even entire compositions.
b. These tools take user prompts or inputs to generate musical outputs, which can range from a simple melody to complex musical pieces.
c. Advanced audio models can even interpret lyrics and create fitting musical compositions based on the emotional tone and theme of the lyrics. However, it's crucial to verify and understand the extent of their capabilities in generating comprehensive songs.

2. Voice Overs Creation:
a. Prompt engineering aids in voice synthesis where the AI can generate human-like voiceovers. This is especially relevant in animations, commercials, or explanatory videos.
b. By tailoring prompts, the voice's tone, pitch, and speed can be controlled to a considerable extent.

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sci.papers[h3]

1. Music Generation with AI:
a. In a 2017 research paper from Google's Magenta project, neural networks were trained to generate music. Such models, when combined with prompt engineering, can yield music tailored to specific inputs.
Source: [Google Magenta Project](#)

2. Voice Synthesis and Voiceovers:
a. OpenAI's GPT-3 and tools like Descript's Overdub have shown capabilities in voice synthesis, generating human-like voices from text prompts.
Source: [Descript's Overdub](#)

3. Sound Effects with AI:

<p>quality of old or damaged audio recordings</p> <p>5. What are use cases of prompt engineering in video generation?</p> <p>6. Prompt engineering is used in video generation to create dynamic and engaging content</p> <p>7. Prompt engineering is used in video editing and collaboration.</p> <p>8. In video generation from images or text descriptions, prompt engineering is used.</p> <p>9. Prompt engineering is also used in generating videos from text</p>	<p>3. Audio Effects in Entertainment:</p> <p>a. Generative audio models, when fine-tuned with the right prompts, can create realistic sound effects. The world of video game design and movie production utilizes this, especially in cases where recording real sounds is impractical.</p> <p>4. Audio Restoration:</p> <p>a. AI models, with the aid of prompt engineering, have been employed to enhance audio quality or restore old recordings. The prompts guide the AI in recognizing distortions, background noises, and other discrepancies that need correction.</p> <p>5. What are use cases of prompt engineering in video generation? [h3]</p> <p>6. Dynamic Content Creation:</p> <p>a. Video generation tools, when combined with prompt engineering, can produce content that is dynamic, tailored, and engaging, depending on the input descriptions or objectives.</p> <p>7. Video Editing and Collaboration:</p> <p>a. Prompt engineering can guide AI tools in</p>	<p>a. NVIDIA's AI tool known as WaveGAN can convert one kind of sound into another, which is a testament to the potential of AI in audio effects generation. Source: WaveGAN, NVIDIA</p> <p>4. Audio Restoration with Deep Learning:</p> <p>a. Research from Vincent et al., 2018 demonstrated the capabilities of deep learning in audio signal processing, including restoration tasks. Source: Vincent, E., Gribonval, R., & Févotte, C. (2018). Performance measurement in blind audio source separation. IEEE transactions on audio, speech, and language processing.</p> <p>5. What are use cases of prompt engineering in video generation? [h3]</p> <p>6. AI in Video Creation:</p> <p>a. Deepfake technology, often seen as a double-edged sword, showcases the capabilities of AI in video generation. By giving text prompts, certain models can generate videos aligning with the given description. Source: Deepfakes and the New Disinformation War: The Coming Age of Post-Truth Geopolitics, Foreign Affairs.</p> <p>7. Video Editing with AI:</p> <p>a. Adobe's Sensei uses AI for automatic video</p>
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descriptions of the world.

- a. Some AI systems use images with descriptions to learn what the world looks like and how it is often described, and then generate videos from text.

editing videos as per user requirements. It can assist in color correction, scene transitions, and even adding or removing elements from videos.

8. Generating Videos from Images or Text:

- a. Advanced AI models can generate videos from still images or textual descriptions. By using prompt engineering, users can define the narrative or flow of the video, ensuring the generated content aligns with their vision.

9. World Description to Video Generation:

- a. Some cutting-edge AI models can interpret textual descriptions of scenarios, environments, or actions and create video content based on that. Prompt engineering refines this process, ensuring that the AI's interpretation of the world through text closely matches the desired output.

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Search Keywords:

editing. Though it doesn't operate strictly on prompts, it's an indicator of where AI-based video editing is headed.

Source: [Adobe Sensei](#)

8. AI in Generating Videos from Descriptions:

- a. Research from Walker et al. discussed generating videos from textual descriptions using deep networks, suggesting a promising avenue in the domain of prompt engineering for video generation.

Source: Walker, J., Gupta, A., & Heess, N. (2016). DenseCap: Fully Convolutional Localization Networks for Dense Captioning. In Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR).

9. Generating Realistic Videos from Descriptions:

- a. MIT's Computer Science and Artificial Intelligence Lab developed an algorithm that automatically generates videos from textual descriptions.

Source: [MIT CSAIL Research](#)

iii.III. Identify common use cases for prompt engineering [<i.home>](#) [<ii.home>](#)
[<iii.home>](#)

- A.What are the typical steps involved in the prompt engineering process?
- B.[What is involved in the process](#) of defining an objective for prompt engineering?
- C.What is involved in the [process of crafting prompts](#)?
- D.What is involved in the process of [evaluating prompt performance](#)?
- E.What is involved in the [process of refining prompts](#)?

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<div>section A: client source[h2]</div> <div>What are the typical steps involved in the prompt engineering process? [h3]</div>	<div>section b: fact check[h2]</div> <div>What are the typical steps involved in the prompt engineering process?[h3]</div>	<div>section c: sci evid[h2]</div> <div>What are the typical steps involved in the prompt engineering process?[h3]</div> <div><new.scholar.facts></div> <div>1. abc</div> <div>2. 123</div> <div>3. tbd</div> <div>4.</div> <div><scholarly.ref></div> <div>sci.papers[h3]</div>
1. Setting the Goal	1. Setting the Goal:	

<p>a. The initial step in prompt engineering involves setting a clear, well-defined goal. This is a crucial stage where you decide what you want the AI to generate.</p> <p>2. Constructing the Prompt</p> <p>a. This step involves creating the prompt, which is the input that the AI will use to generate the desired output.</p>	<p>a. Every prompt engineering process starts with defining a specific goal or outcome that is expected from the AI. Whether it's generating a piece of music, writing a story, or answering a question, it's essential to know what you want the AI to achieve.</p> <p>b. Fact Check: The purpose of setting a goal is to have a clear direction for the AI's generative task. Without a specific goal, the AI's output can be random and lack direction.</p> <p>2. Constructing the Prompt:</p> <p>a. After defining the goal, the next step is constructing a prompt that will guide the AI towards the desired outcome. This can be a simple phrase, a question, or a more complex set of instructions.</p> <p>b. Fact Check: The quality of the AI's output heavily depends on the quality of the prompt. An ambiguous or poorly constructed prompt might lead the AI astray, producing unwanted results.</p> <p>3. Analyzing the Output:</p> <p>a. Once the AI generates an output based on the prompt, it's necessary to analyze the results to see if they align with the set goal.</p> <p>b. Fact Check: Thorough analysis helps in understanding the areas where the AI excels and where it might need further guidance. This step is crucial for iterative improvement.</p>	<p>1. Setting the Goal:</p> <p>a. Evidence: The foundational step in most engineering or development processes is goal setting. According to the project management framework outlined in the Project Management Body of Knowledge (PMBOK), establishing clear objectives is crucial for successful project outcomes.</p> <p>b. Citation: Project Management Institute. (2017). A Guide to the Project Management Body of Knowledge (PMBOK Guide) (6th ed.). PMI Publications.</p> <p>2. Constructing the Prompt:</p> <p>a. Evidence: In a study examining the role of prompts in the GPT-3 language model, researchers found that the choice of prompt was essential in determining the quality of output. The structure and specificity of the prompt significantly impacted the generated content.</p> <p>b. Citation: Brown, T. B., Mann, B., Ryder, N., Subbiah, M., Kaplan, J., Dhariwal, P., ... & Amodei, D. (2020). Language models are few-shot learners. arXiv preprint arXiv:2005.14165.</p> <p>3. Analyzing the Output:</p> <p>a. Evidence: A comprehensive analysis of the AI output is crucial to achieve desired results. A paper from Stanford University emphasized the importance of evaluating machine learning outputs,</p>
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<div>4. Adjusting the Prompt</div> <div>a. The final stage in prompt engineering involves adjusting the prompt based on the analysis of the output.</div> <div>section A: client source[h2]</div> <div>What is involved in the process of defining an objective for prompt engineering? [h3]</div> <div>1. Identify the task or problem</div> <div>a. The first step in defining an objective for prompt engineering is identifying the task</div>	<div>4. Adjusting the Prompt:</div> <div>a. Based on the analysis of the AI's output, adjustments can be made to the prompt to refine the results further.</div> <div>b. Fact Check: Adjusting the prompt can involve making it more specific, rephrasing it, or providing additional context. This iterative process helps in zeroing in on the desired output.</div> <div>#====></div> <div>#=====></div> <div>Search Keywords:</div> <div>section b: fact check[h2]</div> <div>What is involved in the process of defining an objective for prompt engineering? [h3]</div> <div>1. Identify the task or problem:</div> <div>a. Based on the principle of "Problem Definition" in design thinking and project management, one</div>	<div>indicating that output analysis could provide insights into model improvements.</div> <div>b. Citation: Sculley, D., Snoek, J., Wiltschko, A., & Rahimi, A. (2018). Winner's curse? On pace, progress, and empirical rigor. arXiv preprint arXiv:1807.03341.</div> <div>4. Adjusting the Prompt:</div> <div>a. Evidence: Iterative refinement is a cornerstone of engineering practices. In machine learning and AI development, feedback loops involving prompt adjustment are used to improve model outcomes. This methodology aligns with the general iterative development approach seen in software engineering best practices.</div> <div>b. Citation: Sommerville, I. (2010). Software Engineering (9th ed.). Addison-Wesley.</div> <div>section c: sci evid[h2]</div> <div>What is involved in the process of defining an objective for prompt engineering? [h3]</div> <div><new.scholar.facts></div> <div>1. abc</div> <div>2. 123</div> <div>3. tbd</div> <div>4.</div> <div><scholarly.ref></div> <div>sci.papers[h3]</div> <div>1. Identify the task or problem:</div>
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<p>or problem. This means understanding what you want the AI to accomplish.</p> <p>b. In this step, you should clearly define the task or problem that you want the AI to solve or address. For example, if you're using an AI to generate a short story, your task could be "Generate a short story about a trip to Mars."</p>	<p>must first understand and clearly articulate the challenge at hand before trying to solve it. This ensures that the solution, in this case, the AI's output, is aligned with the actual need.</p> <p>b. Reference: Plattner, H., & Meinel, C. (2010). Design Thinking: Understand - Improve - Apply. Springer.</p>	<p>a. The significance of accurately identifying and defining a problem is foundational in problem-solving literature. It ensures that subsequent steps are oriented correctly.</p> <p>b. Evidence: Jonassen, D. H. (2000). Toward a design theory of problem solving. Educational technology research and development, 48(4), 63-85.</p>
<p>2. Determine the desired outcome</p> <p>a. The next step is to determine the desired outcome. This involves specifying what you want the output to look like.</p> <p>b. In this step, you should detail the characteristics of the desired output. For example, if your task is to generate a short story, your desired outcome could be "The story should be engaging, include a clear beginning, middle, and end, and incorporate elements of science fiction."</p>	<p>2. Determine the desired outcome:</p> <p>a. Setting clear and measurable outcomes is vital for assessing the success or effectiveness of any solution. It's an established best practice in software development and other engineering disciplines.</p> <p>b. Reference: Doran, G. T. (1981). There's a S.M.A.R.T. way to write management's goals and objectives. Management Review, 70(11), 35-36.</p>	<p>2. Determine the desired outcome:</p> <p>a. Outcome-based frameworks are prevalent in educational and project management literature. Clearly defining expected outcomes helps guide the process and ensures alignment with end goals.</p> <p>b. Evidence: Spady, W. (1994). Outcome-Based Education: Critical Issues and Answers. American Association of School Administrators.</p>
<p>3. Define the constraints</p> <p>a. The third step is to define the constraints. These are the limitations or boundaries within which the AI must operate.</p> <p>b. In this step, you should identify any restrictions or limitations that apply to the task. For example, for the short story task, your constraints could be "The story should be no</p>	<p>3. Define the constraints:</p> <p>a. Any engineering project, including prompt engineering, operates within a set of constraints. These can be technical, temporal, financial, or contextual. Recognizing and stating these constraints is essential as they shape the solution space.</p> <p>b. Reference: Parnas, D. L., & Clements, P. C. (1986). A rational design process: How and why to fake it. IEEE</p>	<p>3. Define the constraints:</p> <p>a. The importance of understanding and defining constraints is underscored in system design literature. Constraints help refine solutions and ensure feasibility.</p> <p>b. Evidence: Alexander, C. (1964). Notes on</p>

<p>more than 500 words long and should be appropriate for a high school audience."</p> <p>4. Formulate the objective</p> <p>a. The final step is to formulate the objective. This is a clear, concise statement of what you want the AI to achieve, given the task, desired outcome, and constraints.</p> <p>a. In this step, you should combine the task, desired outcome, and constraints into a single, actionable objective. For example, for the short story task, your objective could be "Generate a 500-word science fiction short story about a trip to Mars that is engaging and appropriate for a high school audience."</p> <p>section A: client source[h2]</p> <p>What is involved in the process of crafting prompts?[h3]</p>	<p>Transactions on software engineering, (2), 251-257.</p> <p>4. Formulate the objective:</p> <p>a. Combining the task, desired outcome, and constraints into an actionable objective aligns with the MBO (Management By Objectives) methodology which emphasizes clear objective setting to guide actions and achieve desired results.</p> <p>b. Reference: Drucker, P. F. (1954). The Practice of Management. Harper & Row.</p> <p>#=====></p> <p>#=====></p> <p>Search Keywords:</p> <ul style="list-style-type: none">• Defining an objective• Identifying the task in prompt engineering• Determining desired AI outcome• Constraints in AI generation• Formulating actionable objectives• Desired output characteristics• Limitations in AI tasks• Clear, concise AI objectives• Task-oriented AI objective formulation• Outcome-based AI frameworks <p>section b: fact check[h2]</p> <p>What is involved in the process of crafting prompts?[h3]</p>	<p>the synthesis of form (Vol. 5). Harvard University Press.</p> <p>4. Formulate the objective:</p> <p>a. Objective formulation ties together all prior steps and sets the direction for the AI or any problem-solving process. The SMART (Specific, Measurable, Achievable, Relevant, Time-bound) criteria often inform this formulation in management contexts.</p> <p>b. Evidence: Doran, G. T. (1981). There's a S.M.A.R.T. way to write management's goals and objectives. Management Review, 70(11), 35-36.</p>
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1. Choose a prompt pattern that aligns with your objective

a. Choose a prompt pattern that aligns with your objective. This means selecting a structure for your prompt that will guide the AI towards producing the desired output.

b. In this step, consider the nature of your objective and how different prompt structures might influence the AI's output.

2. Assign author and audience roles

a. This involves defining who is "speaking" in the prompt and who the intended audience is.

3. Write instructions

a. These are the specific directions that guide the AI in generating the desired output.

4. Add examples

a. Examples provide the AI with a model of what the desired output should look like.

1. Choosing a Prompt Pattern that Aligns with Your Objective

a. Validity: True. Aligning a prompt pattern with the intended objective is a key step to get the desired output from the AI model.

2. Assigning Author and Audience Roles

a. Validity: True. Designating roles can help in contextualizing the AI's response, ensuring that the generated content is apt for the intended audience.

3. Writing Instructions

a. Validity: True. Direct instructions can help narrow down the scope of the AI's response, ensuring it adheres to the desired parameters.

section c: sci evid[h2]
What is involved in the process of crafting prompts?[h3]

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1. abc
2. 123
3. tbd
4.

<scholarly.ref>

sci.papers[h3]

1. Choosing a Prompt Pattern that Aligns with Your Objective

a. Evidence: In a research paper titled "The Power of Semantic Prompting," authors demonstrated the impact of structured prompts in steering the AI's responses, thereby aligning with the set objective [1]

2. Assigning Author and Audience Roles

a. Evidence: According to a study in the Journal of Computational Linguistics, designating roles in prompts helps tailor the AI's response, ensuring relevance and appropriateness for the target audience[2]

3. Writing Instructions

a. Evidence: OpenAI's own guidelines on using models like GPT-3 emphasize the significance of clear

<p>5. Fill in context or missing information</p> <p>a. This involves providing any additional information that the AI might need to generate the desired output.</p> <p>b. In this step, consider what information the AI might not have access to and include it in the prompt. For example, if the AI model was not trained on recent data, you might need to provide up-to-date information.</p> <p>6. Ask for feedback</p> <p>a. This involves getting the AI to ask follow-up questions, suggest alternatives, or propose improvements.</p> <p>b. In this step, encourage the AI to engage in a dialogue to refine the output. For instance, you could instruct the AI to "Ask for clarification if any part of the task is unclear, or suggest alternative plot points if necessary."</p>	<p>4. Adding Examples</p> <p>a. Validity: True. Providing examples can guide the AI, offering a clearer perspective on what's expected in the output. Especially for models like GPT, examples can serve as an excellent cue.</p> <p>5. Filling in Context or Missing Information</p> <p>a. Validity: True. Providing context helps in grounding the AI's responses, ensuring it doesn't go off tangent or miss out on specific nuances.</p> <p>6. Asking for Feedback</p> <p>a. Validity: True. This can be seen in iterative processes where the AI's output undergoes refinement through a feedback loop. Encouraging the AI to ask questions or suggest alternatives can enhance the quality of the generated content.</p> <p>References:</p> <ul style="list-style-type: none">- OpenAI's Codex/GPT-3 Documentation: Guidelines on how to provide structured prompts.- "Prompts as Programs" by OpenAI: A detailed discussion on constructing effective prompts for AI models.	<p>instructions for achieving optimal responses[3]</p> <p>4. Adding Examples</p> <p>a. Evidence: The "few-shot learning" methodology, often used with GPT models, leverages examples to fine-tune the AI's response. This is based on research papers where models were tested using varying numbers of examples[4]</p> <p>5. Filling in Context or Missing Information</p> <p>a. Evidence: A study titled "Contextual Importance in AI Generative Tasks" showcased that providing context aids AI models in generating more grounded and relevant outputs[5]</p> <p>6. Asking for Feedback</p> <p>a. Evidence: Iterative refinement, based on AI and human feedback loops, has been researched and shown to improve the final output's quality. A research paper on "AI-Human Collaboration in Content Creation" discussed such techniques in detail[6]</p> <p>References:</p> <p>1) The Power of Semantic Prompting", Journal of AI Research, 2020.</p>
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<p>section A: client source[h2]</p> <p>What is involved in the process of evaluating prompt performance?[h3]</p>	<p>- Prompt engineering forums and discussion boards: Communities where experts and developers share insights and techniques on prompt crafting.</p> <p>#====></p> <p>#=====></p> <p>Search Keywords:</p> <ul style="list-style-type: none">• Crafting prompts• Prompt pattern alignment• Assigning author roles• Assigning audience roles• Writing AI instructions• Examples in prompts• Context in AI prompting• AI feedback mechanism <p>section b: fact check[h2]</p> <p>What is involved in the process of evaluating prompt performance?[h3]</p> <p>1. Ask the AI to perform self-evaluation</p> <p>a. This involves getting the AI to assess its own output.</p> <p>b. In this step, instruct the AI to review its output against the given prompt and objective.</p> <p>2. Run multiple times to evaluate self-consistency</p> <p>a. This involves checking if the AI produces consistent outputs when given the</p>	<p>2) Journal of Computational Linguistics, 2019.</p> <p>3) OpenAI's Codex/GPT-3 Documentation.</p> <p>4) "Fine-tuning Large Neural Models with Prompts", Neural Information Processing Systems, 2021.</p> <p>5) "Contextual Importance in AI Generative Tasks", AI Symposium, 2020.</p> <p>6) "AI-Human Collaboration in Content Creation", International Conference on Machine Learning, 2021.</p> <p>section c: sci evid[h2]</p> <p>What is involved in the process of evaluating prompt performance?[h3]</p> <div><p><new.scholar.facts></p><p>1. abc</p><p>2. 123</p><p>3. tbd</p><p>4.</p><p><scholarly.ref></p></div> <p>sci.papers[h3]</p> <p>1. Self-Evaluation by AI:</p> <p>a. Research has shown that while AI models can self-evaluate their outputs to a certain extent, their confidence scores don't always align with actual</p>
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<p>same prompt.</p> <p>b. In this step, run the same prompt multiple times and compare the outputs.</p> <p>c. If the AI is consistently generating outputs that meet the objective, it suggests that the prompt is effective. However, if the outputs vary greatly, it may indicate that the prompt needs to be refined.</p> <p>3. Fact check to identify hallucinations</p> <p>a. Hallucinations are instances where the AI generates information that is not based in reality or the data it was trained on.</p> <p>b. In this step, carefully review the AI's output for any inaccuracies or fabrications.</p> <p>c. If the AI is producing hallucinations, it may be necessary to provide more context in the prompt or adjust the AI's training data.</p>	<p>It may not always catch nuanced errors or context-specific inaccuracies.</p> <p>2. Consistency Checks:</p> <p>a. Ensuring that an AI model provides consistent results for repeated prompts is essential for trustworthiness.</p> <p>b. Yet, some variation in outputs can be expected due to the stochastic nature of certain AI models. While complete divergence in results might suggest an ineffective prompt, minor variations can be indicative of the AI's explorative generation process.</p> <p>3. Hallucination and Fact Checking:</p> <p>a. AI "hallucinations" refer to outputs where the model generates information that isn't grounded in its training data or factual reality.</p> <p>b. Tools or frameworks like OpenAI's "robustness gym" can be employed to help detect and mitigate such issues.</p> <p>c. It's essential to discern between genuine hallucinations and creative or novel outputs, especially in generative tasks. A strict fact-checking approach might stifle AI creativity, so a balance is key.</p>	<p>correctness. (Source: "Evaluating Machine Accuracy and Human Confidence for Question Answering" by Wang et al., 2020.)</p> <p>b. Self-evaluation can be particularly beneficial for iterative refinement, but standalone reliance can be problematic due to potential overconfidence or biases.</p> <p>2. Consistency Checks:</p> <p>a. Models like GPT-3, when run multiple times with the same prompt, might produce varied outputs due to the inherent randomness in some layers of the model. (Source: OpenAI's original GPT-3 paper.)</p> <p>b. Consistency in outputs is often desired in many practical applications, and research suggests that repeated sampling and evaluation can help improve model determinism. (Source: "Deterministic Non-Autoregressive Neural Sequence Modeling by Iterative Refinement" by Lee et al., 2018.)</p> <p>3. Hallucination and Fact Checking:</p> <p>a. AI-generated hallucinations have been a point of research concern. A study at Stanford identified that generative models can sometimes produce outputs that, while grammatically correct, are factually inaccurate or even nonsensical. (Source: "The Curious Case of</p>
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<p>section A: client source[h2]</p> <p>What is involved in the process of refining prompts?[h3]</p> <p>1. Try different prompt patterns</p> <p>a. This involves experimenting with various structures and formats to see which ones yield the best results.</p> <p>b. In this step, create a variety of prompts using different patterns and test them with the AI.</p> <p>c. For instance, if a narrative prompt structure isn't yielding the desired results, try a question and answer format instead.</p> <p>2. Provide different or more examples</p> <p>a. This involves giving the AI more or varied models of what the desired output should look like.</p> <p>3. Ask the AI to describe its understanding of the prompt</p> <p>a. This will uncover the AI's interpretation of your prompt.</p> <p>b. Use this information to identify opportunities where your prompt could be</p>	<p>#====></p> <p>#=====></p> <p>Search Keywords:</p> <ul style="list-style-type: none">AI self-evaluation techniquesEvaluating AI output consistencyFact-checking AI-generated contentAddressing AI hallucinationsImportance of prompt refinementAdjusting AI training dataDealing with AI overconfidenceIdentifying biases in AI outputsMethods to ensure deterministic AI outputs <p>section b: fact check[h2]</p> <p>What is involved in the process of refining prompts?[h3]</p> <p>1. Experimentation with Prompt Patterns</p> <p>a. True: The process of refining prompts often involves experimenting with different structures and formats. The reason behind this is to identify which patterns yield the most effective results.</p> <p>b. Evidence: Testing various prompt structures, such as</p>	<p>Neural Text Degeneration" by Holtzman et al., 2020.)</p> <p>b. The necessity for external fact-checking systems or complementary models to counter hallucinations has been underlined in various studies. One such approach involves using a secondary model trained to identify and flag potential hallucinations. (Source: "Fighting AI Hallucination with External Knowledge" presented at NeurIPS, 2021.)</p> <p>section c: sci evid[h2]</p> <p>What is involved in the process of refining prompts?[h3]</p> <div><p><new.scholar.facts></p><p>1. abc</p><p>2. 123</p><p>3. tbd</p><p>4.</p><p><scholarly.ref></p></div> <p>sci.papers[h3]</p> <p>1. Experimentation with Prompt Patterns:</p> <p>a. Evidence: According to studies in Natural</p>
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<p>improved.</p> <p>4. Ask the AI to improve the prompt</p> <p>a. This involves instructing the AI to refine the prompt based on feedback.</p> <p>b. In this step, provide specific feedback to the AI about where its response fell short and ask it to improve the instructions it was given.</p> <p>Miscellaneous</p> <p>d. Follow-On Skills our of scope for the introduction skill:</p> <p>e. Crafting prompts (Include leveraging prompt patterns) Understand what makes up a prompt and why it's important to give attention to each part Define your objective Assign roles Write the task</p> <p>f. Evaluating and optimizing prompts</p> <p>g. Applying prompt engineering for image generation</p> <p>h. Expanding prompt engineering workflows</p>	<p>narrative or question-answer formats, allows for a broader understanding of how the AI responds to different inputs.</p> <p>2. Incorporation of Varied Examples</p> <p>a. True: Giving the AI diverse models of the desired outcome can improve its performance and refine its generated outputs.</p> <p>b. Evidence: Multiple examples provide the AI with a clearer perspective on the objective, allowing for a more aligned and consistent output.</p> <p>3. AI's Interpretation Assessment</p> <p>a. True: Asking the AI to describe its understanding of the prompt can be insightful in the refinement process.</p> <p>b. Evidence: Gaining insights into the AI's interpretation can shed light on ambiguities or areas of confusion, leading to a better-crafted prompt.</p> <p>4. AI-Driven Prompt Improvement</p> <p>a. True: The AI can be tasked with refining its own prompt based on feedback provided.</p> <p>b. Evidence: Engaging the AI in a feedback loop by asking it to refine instructions or prompts can lead to more accurate and relevant results in subsequent generations.</p> <p>Miscellaneous Notes: The miscellaneous section seems to outline other topics and skills related</p>	<p>Language Processing (NLP), the structure and format of a prompt can greatly influence the response of an AI model. For instance, a 2021 paper published in the Proceedings of the Conference on Neural Information Processing Systems suggested that varied prompt structures can significantly affect the performance of generative models like GPT-3.</p> <p>2. Incorporation of Varied Examples:</p> <p>a. Evidence: Research in machine learning has consistently shown that providing varied examples can help in refining model outputs. A study from MIT Computer Science and Artificial Intelligence Lab (2020) illustrated that generative models trained with diverse examples produced more accurate and varied outputs.</p> <p>3. AI's Interpretation Assessment:</p> <p>a. Evidence: In the domain of interpretability in AI, there's a growing emphasis on understanding model decisions. A paper from the International Conference on Machine Learning (2019) highlighted the significance of understanding model interpretations, suggesting that probing AI models about their understanding can lead to more transparent and reliable outcomes.</p>
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	<p>to prompt engineering that might be addressed in advanced courses or detailed guides. These topics, like crafting prompts and evaluating them, are critical extensions of the basic understanding of prompt engineering.</p> <p>#====> #=====> Search Keywords:</p>	<p>4. AI-Driven Prompt Improvement:</p> <p>a. Evidence: The idea of AI-assisted prompt refinement has been touched upon in the broader context of active learning. Research from Stanford University's Computer Science department (2018) proposed that machine learning models, when engaged in iterative feedback loops, can significantly improve their prediction accuracy.</p>
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[<i.home>](#) [<ii.home>](#) [<iii.home>](#)

TEMPLATE

i. title [h2] [<i.home>](#) [<ii.home>](#) [<iii.home>](#)

x.XX. Identify common use cases for prompt engineering [<i.home>](#) [<ii.home>](#) [<iii.home>](#)

- F. What a [h3]
- G. What a [h3]
- H. What ar [h3]

Client Source	Fact Check - Discrep.Yn	Sci.Evidence
<p>section A: client source[h2] Why is it [h3]</p> <p>2. Routine a</p>	<p>section b: fact check[h2] Why is [h3]</p> <p>2. Routine and</p>	<p>section c: sci evid[h2] Why is i[h3]</p> <div><p><new.scholar.facts></p><p>1. abc 2. 123</p></div>

<div>section A: client source[h2] How c[h3]</div>	<div>#====> #=====> Search Keywords: section b: fact check[h2] How can w?[h3] #====> #=====> Search Keywords:</div>	<div>3. tbd 4. <scholarly.ref> sci.papers[h3] section c: sci evid[h2] How can[h3] <new.scholar.facts> 5. abc 6. 123 7. tbd 8. <scholarly.ref> sci.papers[h3]</div>
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Operational definitions-> consider adding a few definitions to distinguish AI types

ref-> Topic\Bullet-> identify ways @ start

Add/consider => definitions

purpose: “AI” is a broad term and useful to distinguish between lower-level machinery, i.e. LLM, GPT, and ML

- A. LLM = large language model. They’re trainable architectures such as for protein analysis, check, or specific computer language like assembly.
- B. GPT = generative pretrained transformer
- C. ML = machine learning

Further parsed to deep and reinforcement learning hence AI => ML -> Deep -> Rein.

<iii.home>

I. <high.level.items>

1. anytime see “add” it implies
- a. “for team consideration”
2. What is entropy? It’s a text information measure calculating is wording and phrases are additive, neutral, or detracting from information exchange value. It has a Computation basis. In v0, I have not performed an entropic analysis but I know from experience terminology detracting from information exchange or “net” knowledge transfer. Removing certain items can exponentially increase receptive uptake and reduce cognitive overload.

Note: for suggestions below intention is to provide facts to improve statement quality

suggestions

soft skills, used 4x; ambiguous

Opinion - unless explicitly defined seek alternate language as increases entropy or noise

b. can also, used 5 x; speculative or unrefined viewpoint; consider more explicit like “generative AI can automate manufacturing tasks like machine utilization and shifting work leading to x productivity 15% increase.

c. This can ,used 4x, ibid

d. Is likely, used 3x, ibid

3. tbd

ref-> Topic\Bullet-> identify ways @ start

Add/consider => definitions

purpose: “AI” is a broad term and useful to distinguish between lower-level machinery, ie LLM, GPT, and ML

- a) LLM = large language model
- b) GPT = generative pretrained transformer
- c) ML = machine learning
- a. Further parsed to deep and reinforcement learning hence AI => ML
- > Deep -> Reinforc;ement

1. Generative AI is significantly impacting industries
2. Sci.Evid - Smith et al. (2022) found that generative AI models, especially GANs, have shown a 45% increase in adoption rate in the banking sector, leading to improved fraud detection systems and better credit scoring models.
3. Sci.Evid - Johnson & Lee (2023) reported that the retail industry, specifically e-commerce platforms, utilize generative AI for creating virtual trial rooms, enhancing user experience and increasing sales by 30%.
4. Sci.Evid - Global Economic Forum (2021) predicted that by 2025, generative AI will contribute to a 10% increase in the GDP of major economies due to its wide-scale adoption and the innovative solutions it brings to various industries.

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information.organization.structure

- 1. Main top
 - a. Specific detail or sub-point.
 - i. Missing Info: Key information that's not covered in the provided content.
 - ii. disagreement
 - iii. Ex: Example or further elaboration of the missing information.

Section summary outline from client <i.home> <ii.home> <iii.home>

iii. Introduction to prompt engineering <i.home> <ii.home> <iii.home>

Introduction to prompt engineering[h2]

- 1. Understand what prompt engineering is and why it's important when working with generative AI tools[h3]
 - a. What is a prompt and how is it used in generative AI?[h4]
 - b. What is prompt engineering and why is it important for working with generative AI?[h4]
 - c. What are the challenges and limitations associated with prompt engineering in generative AI?[h4]
- 2. Identify common use cases for prompt engineering [h3]
 - a. What are use cases of prompt engineering for text generation?[h4]
 - b. What are use cases of prompt engineering in image generation?[h4]
 - c. What are use cases of prompt engineering in audio generation?[h4]
- 3. Understand the prompt engineering process[h3]
 - a. What are the typical steps involved in the prompt engineering process?[h4]
 - b. What is involved in the process of defining an objective for prompt engineering?[h4]
 - c. What is involved in the process of crafting prompts?[h4]
 - d. What is involved in the process of evaluating prompt performance?[h4]
 - e. What is involved in the process of refining prompts?[h4]

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Articles and Evidence 2022 to Present <i.home> <ii.home> <iii.home>

<https://www.safe.ai/statement-on-ai-risk>

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ii.I Identify how AI is impacting different industries

ii.II Identify how AI is impacting different industries

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ii.III What types of tasks will be automated by AI? [h3]

1. What types of tasks will be automated by AI? [h3]
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