

HUMAN-AI INTERACTION

Eunsuk Kang

Required reading:

Building Intelligent Systems by Geoff Hulten (2018), Chapter 8.

Guidelines for Human-AI Interaction. Saleema Amershi, et al., in CHI 2019.

Optional reading:

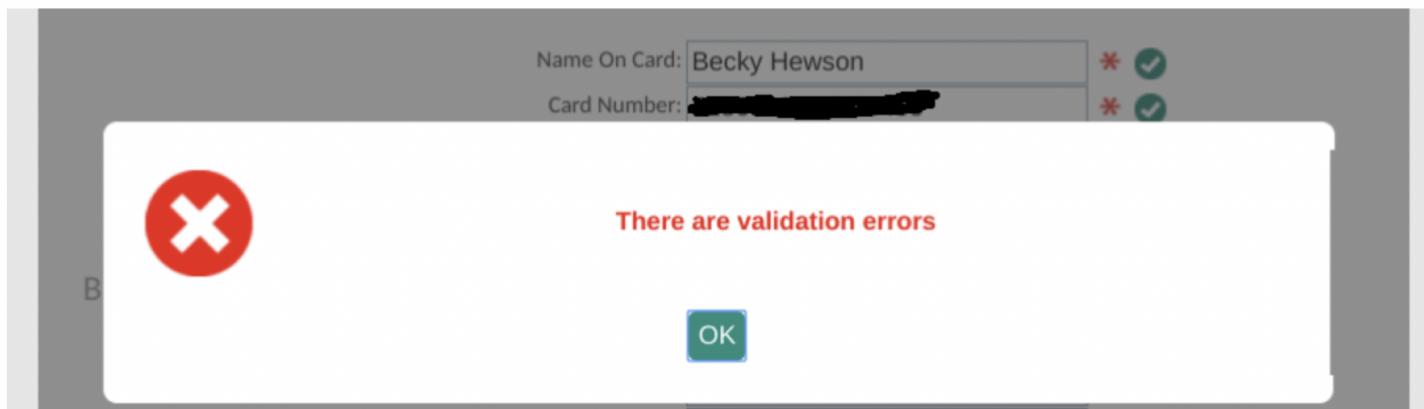
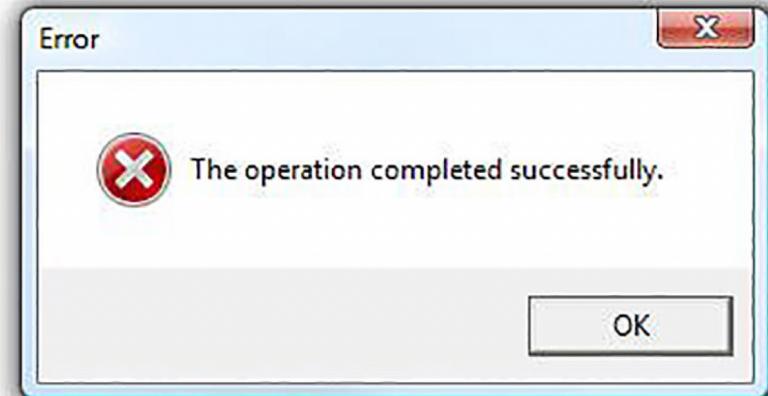
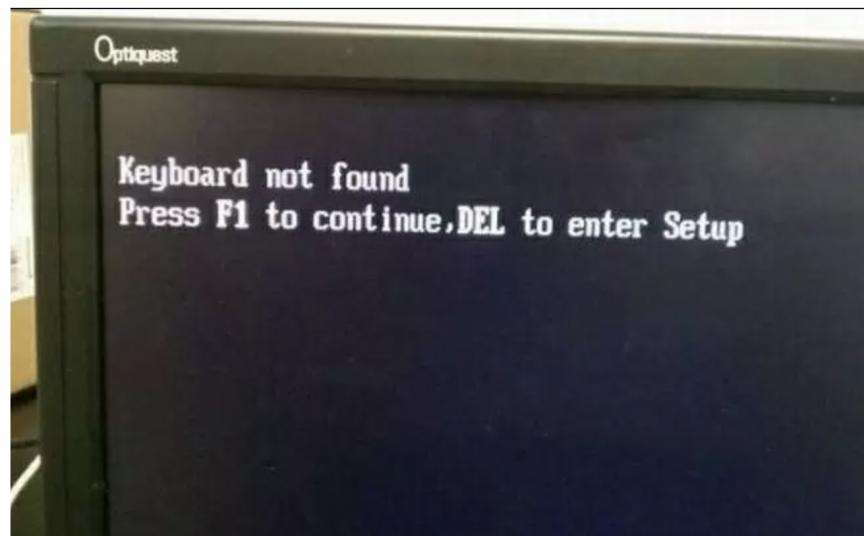
Will You Accept an Imperfect AI? Exploring Designs for Adjusting End-user Expectations of AI Systems. Kocielnik, et al., in CHI 2019

LEARNING GOALS

- Understand the risks of poor interaction design
- Understand the challenges behind designing human-AI interactions
- Understand the basic elements of user interaction design
- Consider design considerations for AI-based systems
 - Modes of interaction: Automate or augment?
 - Mental model: User understanding of what AI is doing
 - Dealing with errors: Guide user towards recovery & prevention
 - Feedback and control: Align user feedback with AI improvement

RISKS OF POOR INTERACTION DESIGN

POOR INTERACTION DESIGN CONFUSES USERS



POOR INTERACTION DESIGN ANNOYS USERS



POOR INTERACTION DESIGN CAUSES HARM

Alexa recorded a woman's private conversation and sent it to a random contact

Kyle Wiggers

@Kyle_L_Wiggers

May 24, 2018 7:38 AM

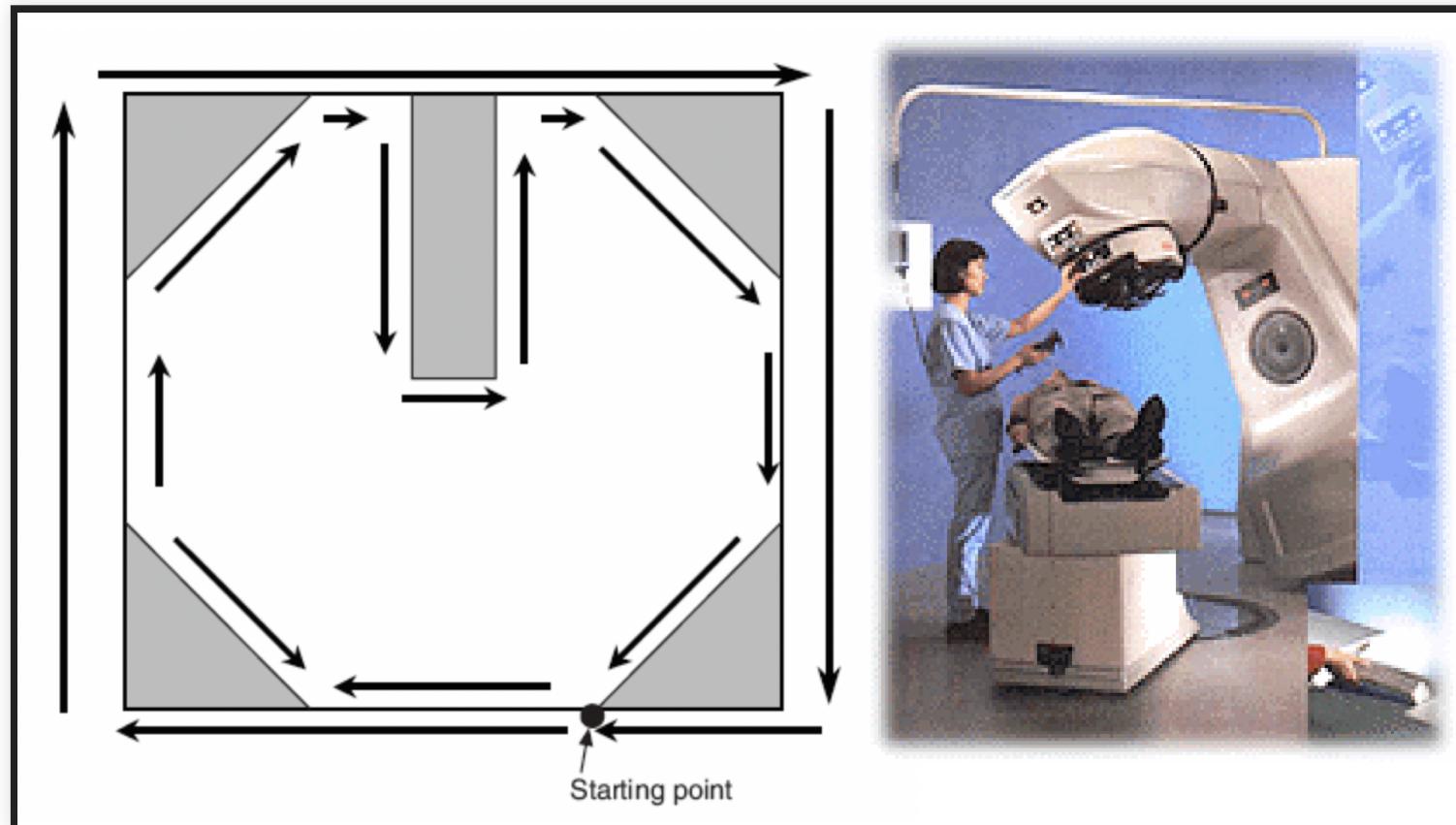
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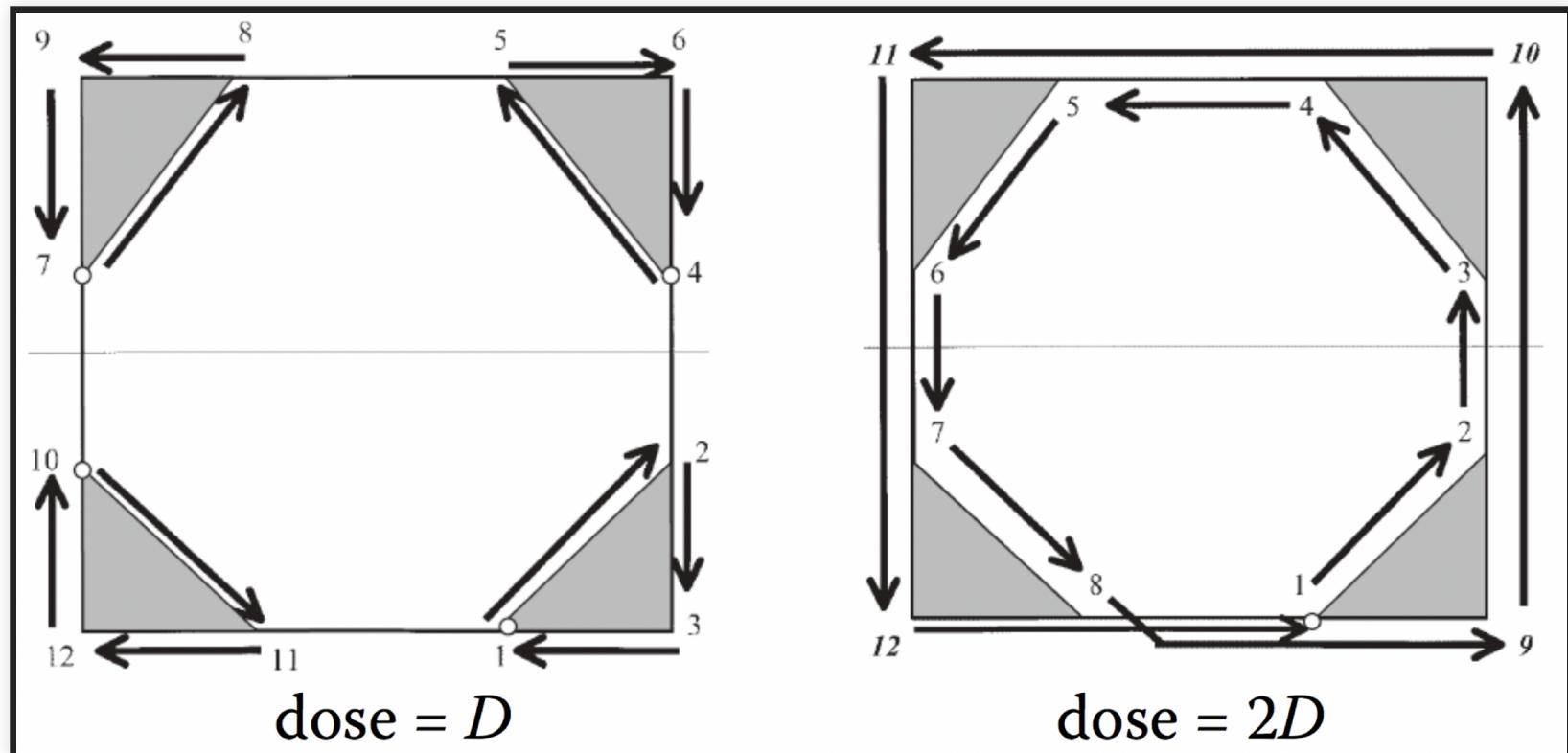


POOR INTERACTION DESIGN CAUSES HARM



- Radiation therapy system at Panama City public hospital (2001)
 - Therapist draws block shapes to determine treatment area
 - Software computes final radiation settings

POOR INTERACTION DESIGN CAUSES HARM



- Same shape drawn in different order, double the radiation dose
- 28 patients overdosed; 8 dead
 - Therapists charged with 2nd degree murder (but are they really to blame?)

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 - Cause physical (injuries, deaths) and societal harms (bias, misrepresentation)

USABILITY CONCEPTS

DIMENSIONS OF USABILITY

<https://www.nngroup.com/articles/usability-101-introduction-to-usability/>

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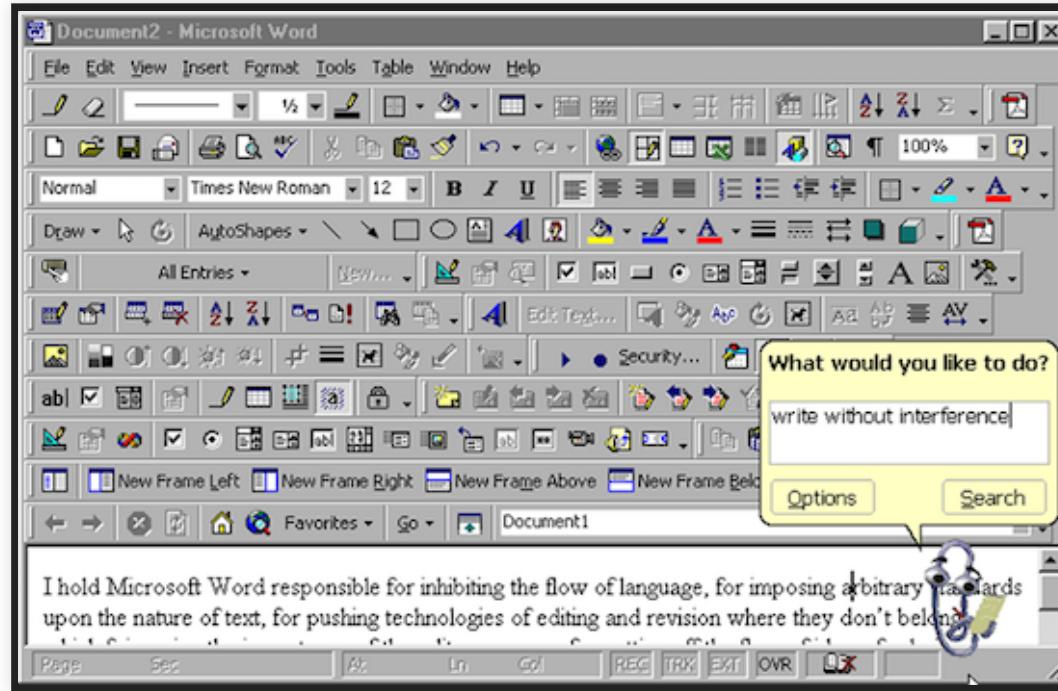
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- Satisfaction: How pleasant is it to use the design?

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INTERACTION COST



- Mental and physical effort needed to perform a desired task
 - Task memorization & recall, context switch, track system state
 - Reading, scrolling, clicking, typing, waiting for UI changes
- **Goal of usable design:** Minimize interaction cost while allowing users to perform their tasks

USABILITY & AI



- AI has potential to greatly reduce interaction costs
 - Automate tasks through personalization & predictions
- But also introduces new challenges
 - **Unpredictability:** AI makes mistakes, sometimes unexpectedly
 - **Opaqueness:** User has difficulty understanding how system works
 - **Evolution:** AI behavior changes over time, surprising users

DESIGN CONSIDERATIONS FOR AI

- **Modes of interaction:** Automate or augment?
- **Mental model:** User understanding of what AI is doing
- **Dealing with errors:** Guide user towards recovery & prevention
- **Feedback and control:** Align user feedback with AI improvement

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 - Organize: Display a set of items in an order
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- Augment when:
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 - Difficult to communicate user's need to AI
 - User enjoys performing the task (e.g., driving)

AUTOMATE OR AUGMENT? WHY?

The screenshot shows a Microsoft PowerPoint slide titled "Measuring Progress?" with slide number 47. The slide contains a bulleted list about software engineering progress. The "Design" tab is selected in the ribbon. A "Design Ideas" panel is open on the right, displaying two design variations of the slide. The top variation has a red gradient background and the bottom one has a dark blue/black background. Both variations include a quote from the slide content.

46

47 Measuring Progress?

48 • "I'm almost done with the app. The frontend is almost fully implemented. The backend is fully finished except for the one stupid bug that keeps crashing the server. I only need to find the one stupid bug, but that can probably be done in an afternoon. We should be ready to release next week."

49

50

51

52

53 47 15-313 Software Engineering ISR institute for SOFTWARE RESEARCH

54

55 Tap to add notes

56

Slide 47 of 74 Notes 29%

Design Ideas

Measuring Progress?

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Design transformations in PowerPoint

AUTOMATE OR AUGMENT? WHY?



Fall detection in a smartwatch

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- Cost: What is the effect of a wrong prediction?
 - If possible, provide a way to undo the action of AI

FACTORS TO CONSIDER

Slide design transformations:

The screenshot shows a Microsoft PowerPoint slide titled "Measuring Progress?". The slide contains a bulleted list:

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Below the slide, there is a "Notes" section with the text "Tap to add notes". On the right side of the slide, there is a "Design Ideas" panel. This panel displays two versions of the slide: a dark-themed version at the top and a light-themed version at the bottom. Both versions show the same content as the main slide. The "Design Ideas" panel also includes a preview of the slide with a timestamp "6" at the bottom right.

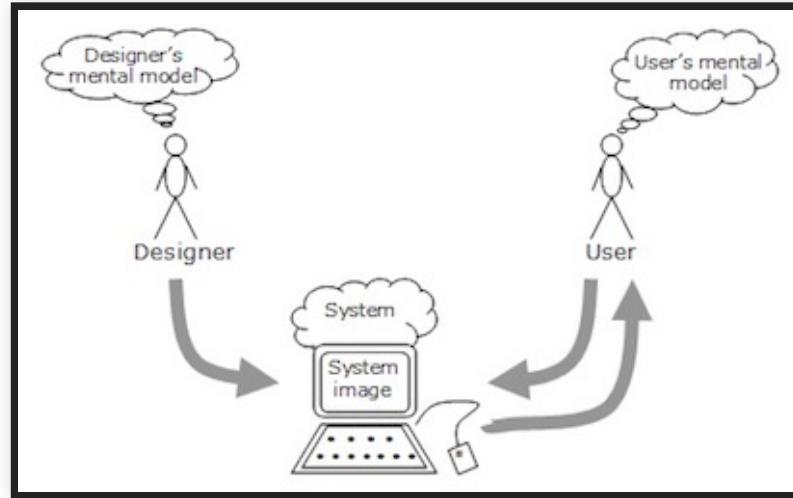
Fall detection:



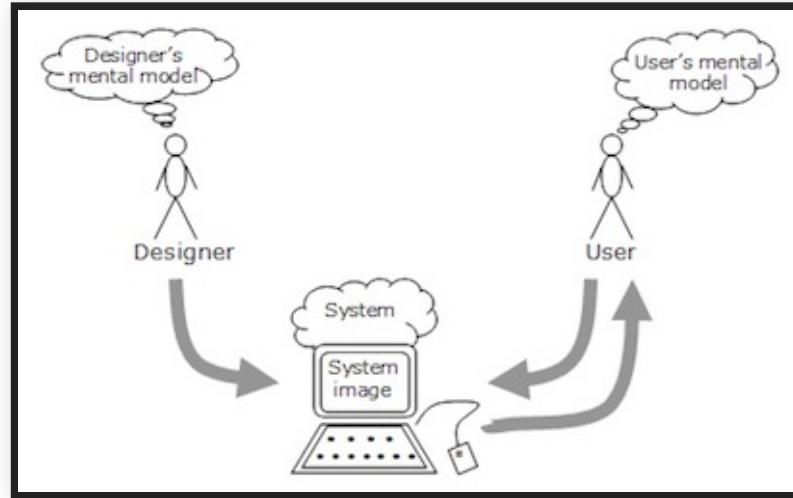
Q. Forcefulness, frequency, cost?

MENTAL MODEL

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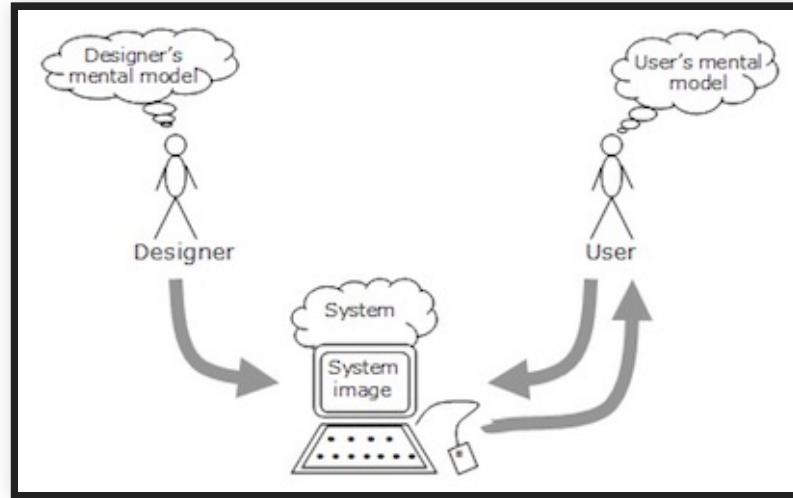


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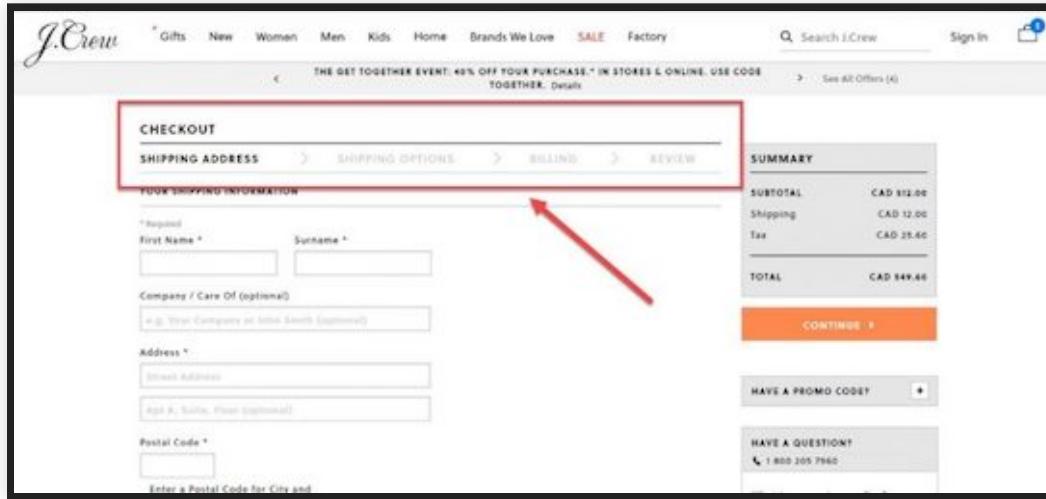
- What the user believes about the system
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- What the user believes about the system
 - "How does the system work? How does it respond to my actions?"
 - User plans actions and reacts to system based on this mental model
- Challenge: Aligning system with the user's mental model
 - Inherent mismatch between user's & designer's models
 - User's model may be preconceived based on prior experience
 - User's model and/or system evolves over time

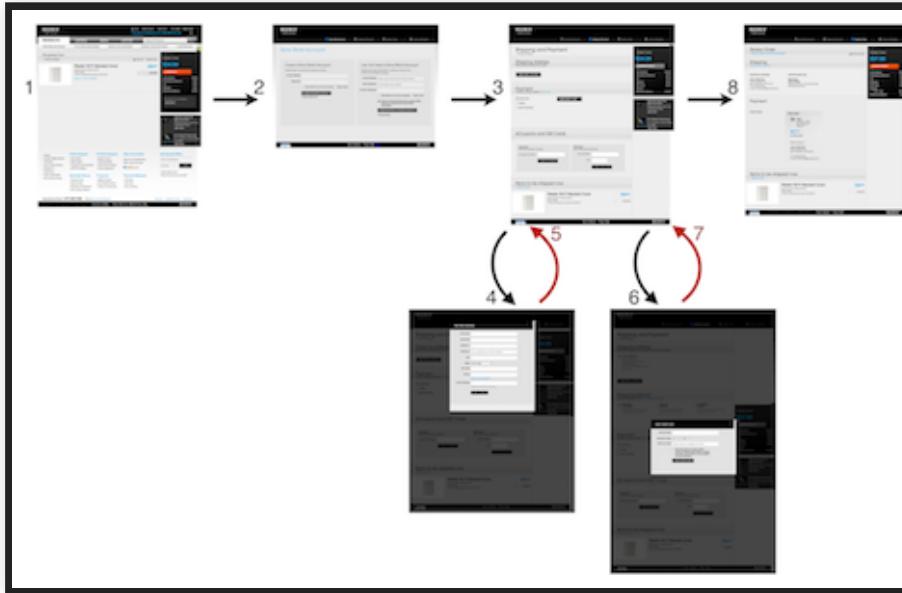
EXAMPLE: SHOPPING CART CHECKOUT



Mental model for shopping cart = A linear sequence of familiar steps

1. Browse for items
2. Add items to cart
3. Choose checkout
4. Enter shipping & billing data
5. Press submit
6. Get confirmation

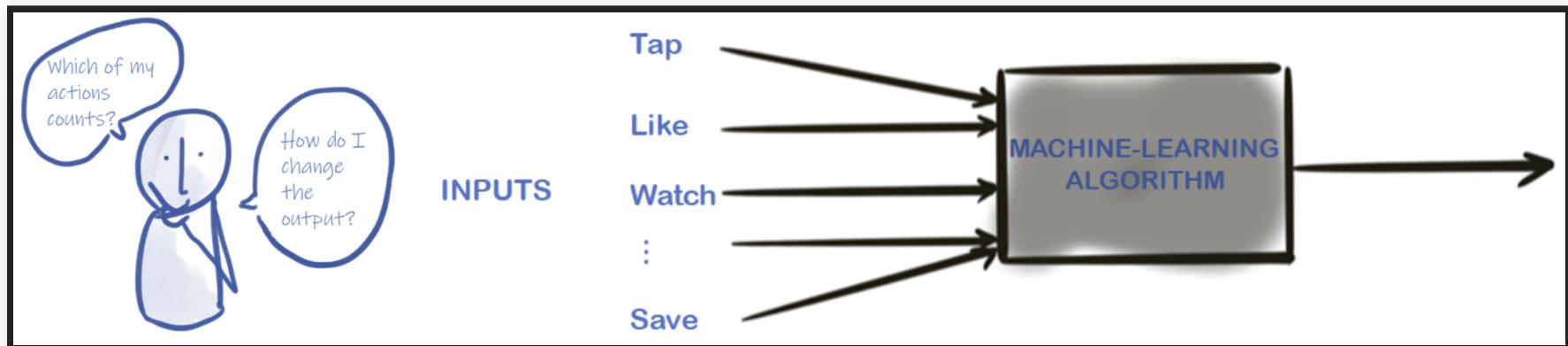
BREAKING MENTAL MODEL



- Anti-pattern: Interrupt linear flow & bring user back to a previous step
 - Create an account, open a new dialog to enter preferred address...
 - Breaks user's mental model => failure to convert into sales
- ~60% of customers abandon their shopping cart

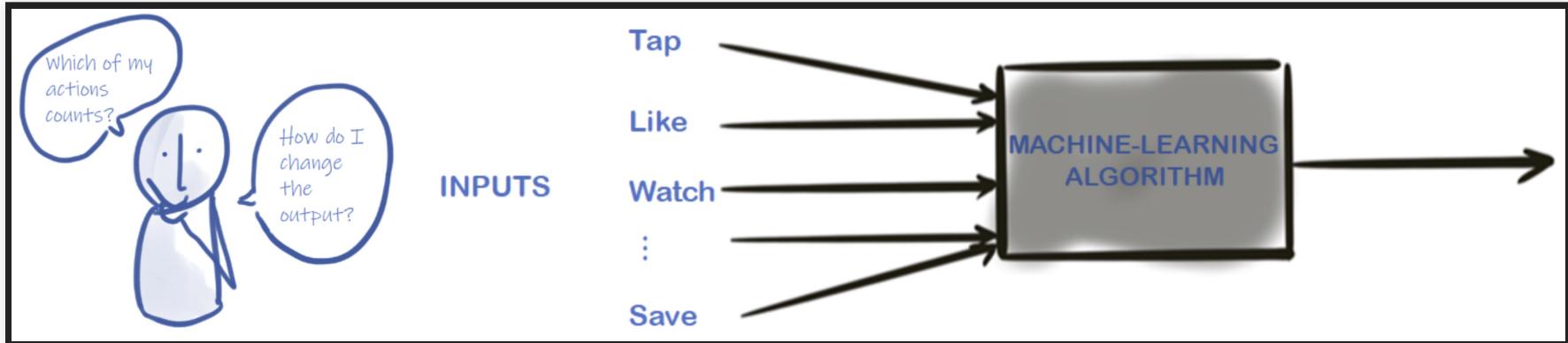
<https://baymard.com/blog/checkout-process-should-be-linear>

MENTAL MODEL FOR AI-BASED SYSTEMS



- User: "What is AI doing, and how do I use it?"
 - Typically less transparent than traditional applications
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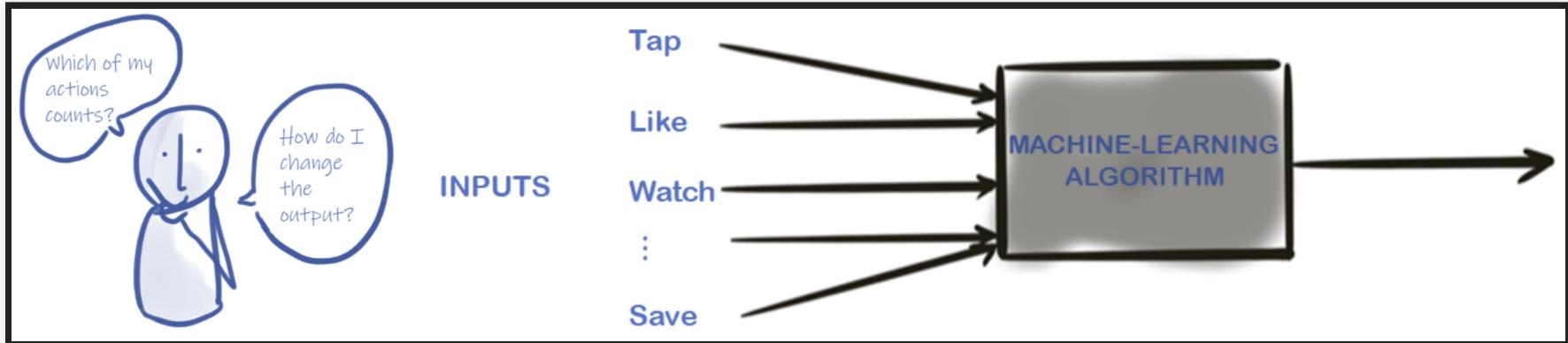
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- Lack of control over output: Why am I being given these recommendations? Why is the output displayed in this order?
- Lack of trust over output: How do I know the output is correct?

MENTAL MODEL FOR VOICE ASSISTANTS?



Q. Can you describe what it does? What it can't do?

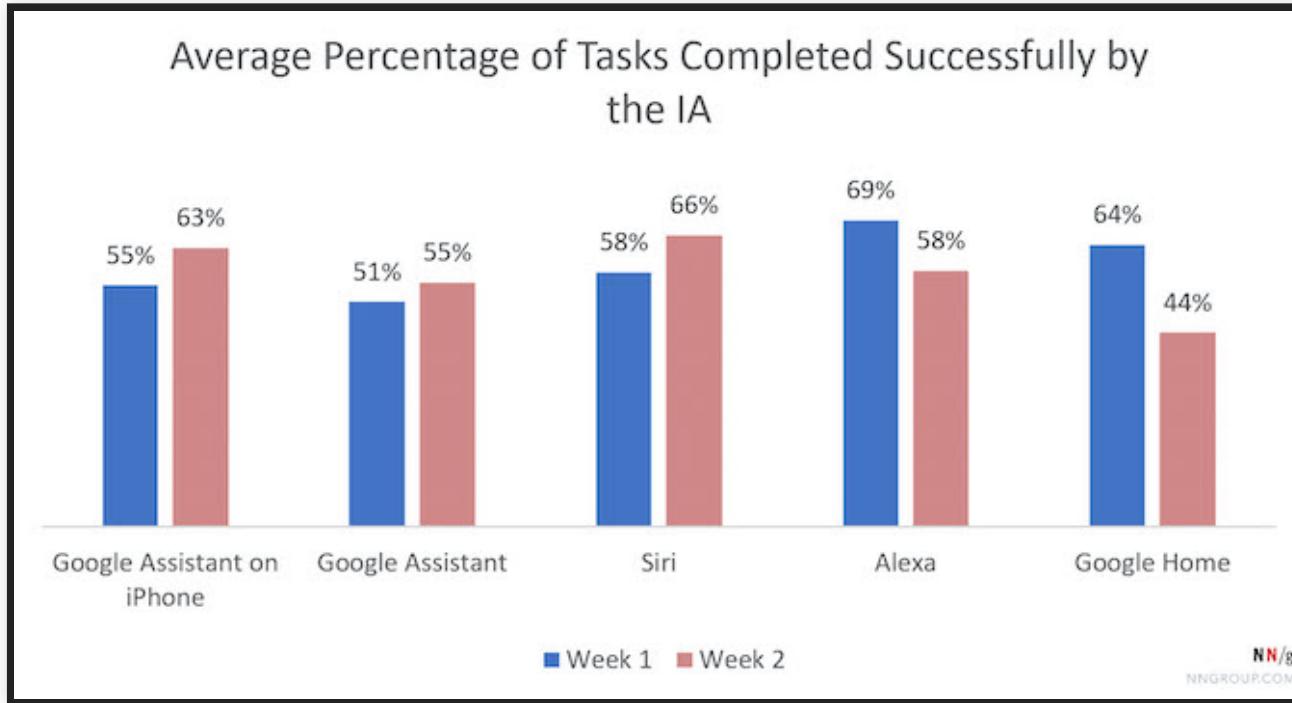
MENTAL MODEL FOR VOICE ASSISTANTS?



- Unclear, inconsistent mental model
 - An interface for other services?
 - "Handy helper"?
 - Knowledge repository? Fact-finding tool?

<https://www.nngroup.com/articles/mental-model-ai-assistants/>

MISALIGNMENT IN VOICE ASSISTANTS



- AI often fails to meet user expectations
 - (1) User doesn't know how to get AI to do X
 - (2) User says X, but AI can't do X well
- Users settle on simple tasks over time; small but limited improvements

MISALIGNMENT IN MENTAL MODELS

“So, this week, I realized that I don't use my IA nearly as much as I thought I did. I do use it often. However it's very much normally the same like five things over and over again.”

- User settles on a suboptimal mental model & fails to benefit from the full capabilities of AI

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- Improve/adjust the user's mental model
 - Set the user's expectations through onboarding
 - Increase transparency and explain decisions made by AI
 - Allow user to adjust system behavior to match their expectations

ONBOARDING: SET USER EXPECTATIONS

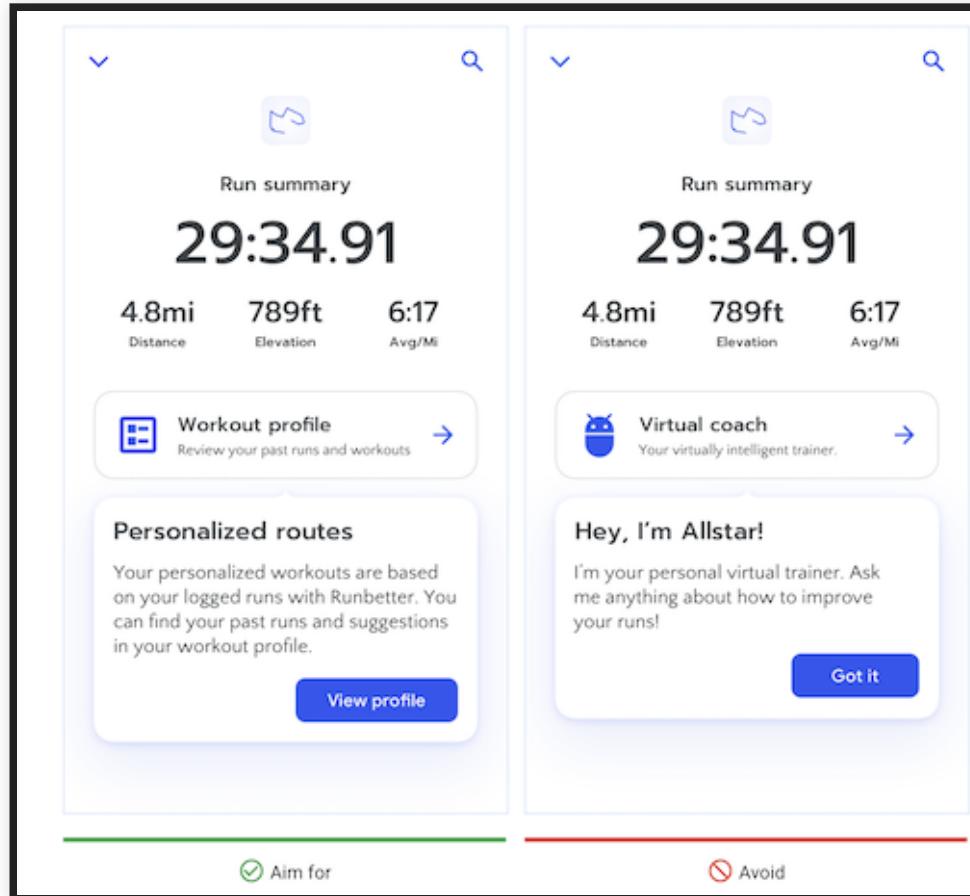
The image displays two screenshots of a writing application's user interface, illustrating how to set user expectations through onboarding.

Screenshot 1: This screenshot shows a document titled "Demo document". A blue callout box on the left says: "This is presumably what a document you've loaded into the app would look like. It seems like a really nice and simple interface!". A blue callout box on the right says: "Ooh! A pulsing hotspot! appears on the first highlighted grammatical error! Let me click...". The top navigation bar includes icons for back, forward, and file operations, along with status indicators (26 ALERTS, 17 PREMIUM) and a "CORRECT WITH ASSISTANT" button.

Screenshot 2: This screenshot shows the same "Demo document" with a different grammatical error underlined: "eight year old". A tooltip has appeared over the underlined text, titled "In-line corrections", with the sub-instruction: "Hover your mouse over underlined words to fix issues with one click.". A blue callout box on the right says: "Clicking the hotspot, I get a little tooltip explaining how this all works!". Another blue callout box below it says: "Ok, what's next? Let me close...". The top navigation bar is identical to the first screenshot.

- Provide examples of how it works

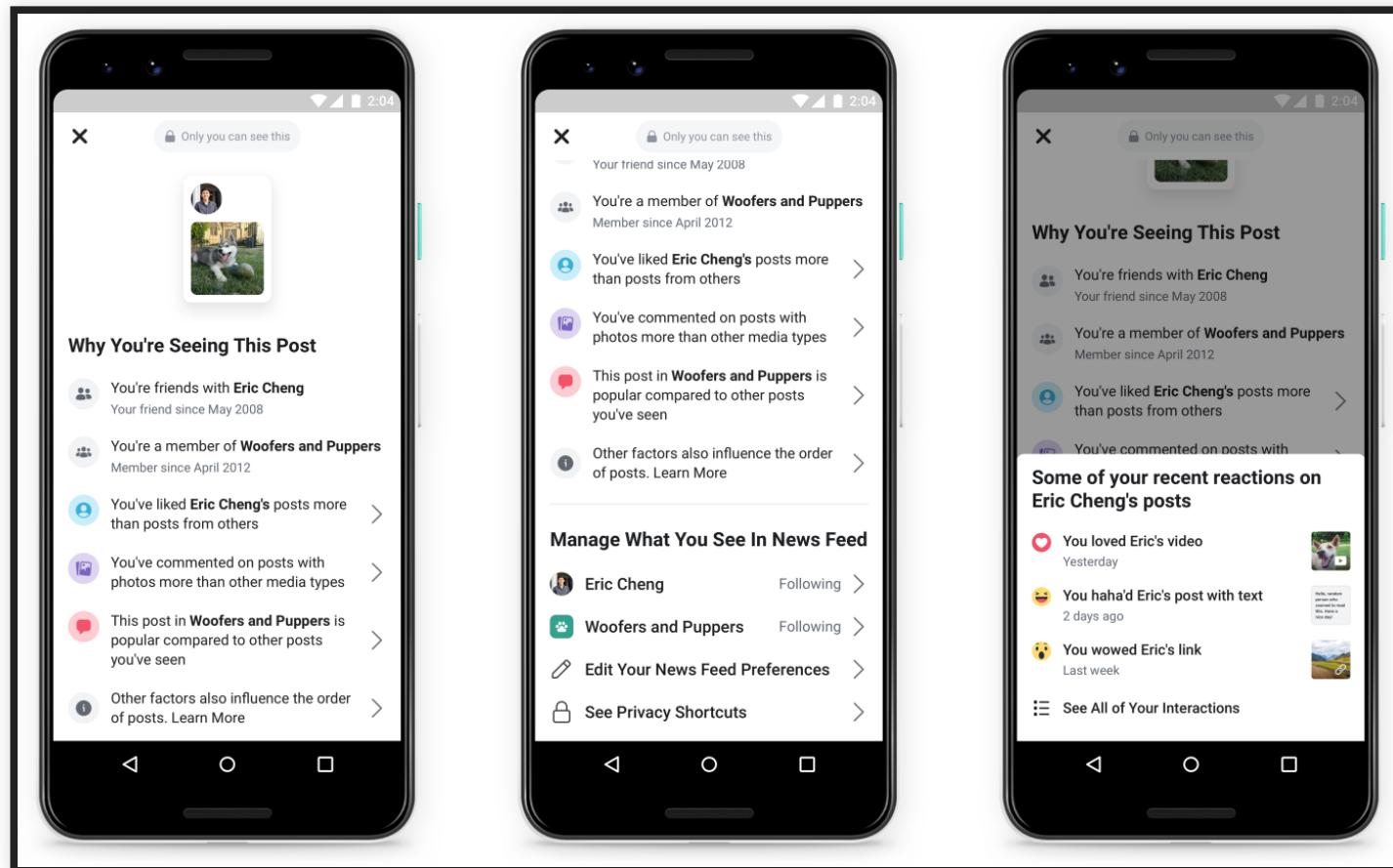
ONBOARDING: SET USER EXPECTATIONS



- Be clear about what system can(not) do

<https://pair.withgoogle.com/chapter/mental-models/>

TRANSPARENCY: EXPLAIN HOW DECISIONS ARE MADE



- Explain how the user's actions influence output

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- User errors: Mistakes made by users (e.g., click on a wrong button)
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 - Error taxonomies, human performance modeling, task analysis, ergonomic analysis, etc.,
- System errors: Failure to provide an outcome expected by the user
 - Our focus in this class

EXAMPLE: SCHEDULING ASSISTANT

The screenshot illustrates a scheduling assistant interface. On the left, labeled 'A', is an 'Inbox' section containing several email messages:

- John Bass: Saturday December 29, We will plan on Meeting at Al's Formal Wear at 1P...
- Kate Bush: keeping the lights on, John- ,Was lovely meeting you this weekend. ,Sorry...
- Daphne Co: Dinner, Hi Eric, ,Would you and Shanna like to meet us for...
- Sally Beck: Per Your Request, Sally, ,Please find attached the file that we discuss...

On the right, labeled 'B', is a detailed view of an email from John Bass:

Saturday December 29
JB
John Bass
Wed Aug 22 2018 10:57 AM
Eric Ramiro

The message content is: "We will plan on Meeting at Al's Formal Wear at 1PM on that Saturday. I will see you all then." Jason has responded: "We think we've found an event".
Details:
Date: Sat Aug 25 2018
Time: 01:00 pm

Buttons at the bottom: Create Appointment, Cancel, Edit details.

On the far right, labeled 'C', is a large red circle highlighting the 'Create Appointment' button. At the top right of the main window, labeled 'D', is a blue 'Reply' button with a reply arrow icon.

- Analyze e-mail content for possible meeting scheduling
- Suggest creating a new meeting based on inferred information

Will You Accept an Imperfect AI? Exploring Designs for Adjusting End-user Expectations of AI Systems. Kocielnik, et al. (CHI 2019)

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- Provide meaningful error messages to user
 - Provide an explanation for the error
 - Suggest actions to fix the error (e.g., "Edit details" option)

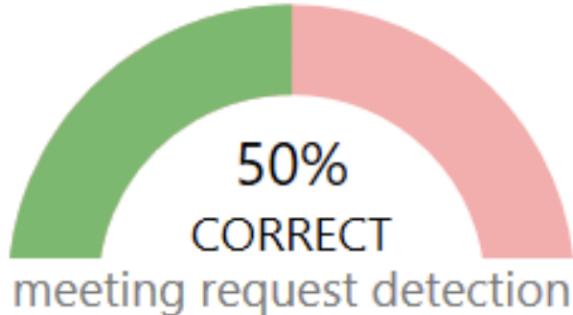
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 - Suggest actions to fix the error (e.g., "Edit details" option)
- Give user controls to recover from and mitigate the effect of an error
 - e.g., delete or modify incorrect meeting schedule

SETTING USER EXPECTATIONS FOR ML ERRORS



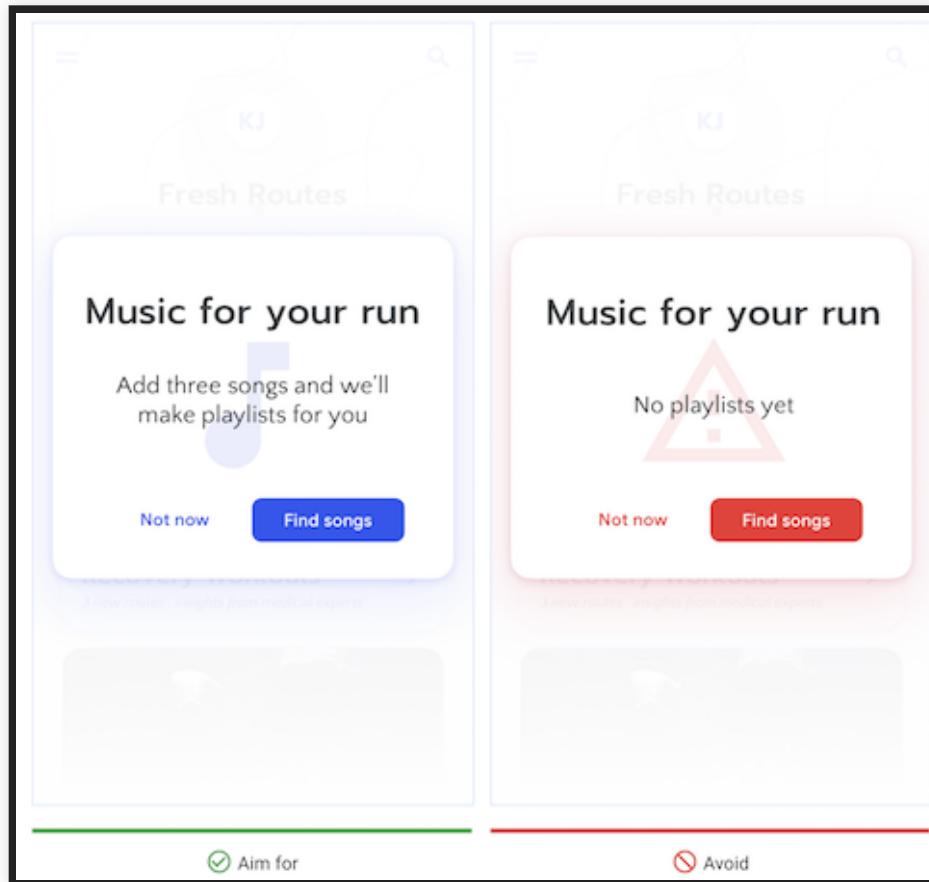
The Scheduling Assistant can correctly detect meeting requests about 50% of the time.



- Be upfront about how well the system performs (e.g., model accuracy)
- Temper the user's expectations and avoid surprises

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ERROR MESSAGES: SUGGEST USER ACTIONS



- Tell the user what the AI needs in order to behave as intended
- Guide the user towards ways to recover from/prevent further errors

<https://pair.withgoogle.com/chapter/errors-failing/>

ERRORS IN VOICE ASSISTANTS

CHANGE TO DIFFERENT EXAMPLE



“...sometimes it says it does – like the reminders and the sending messages. It says it will do it. But then at the end we found that it didn’t really send the message.”

- Q. How do we detect an error?
- Q. How can we notify/guide the user when an error occurs?

<https://www.nngroup.com/articles/mental-model-ai-assistants/>

FEEDBACK AND CONTROL

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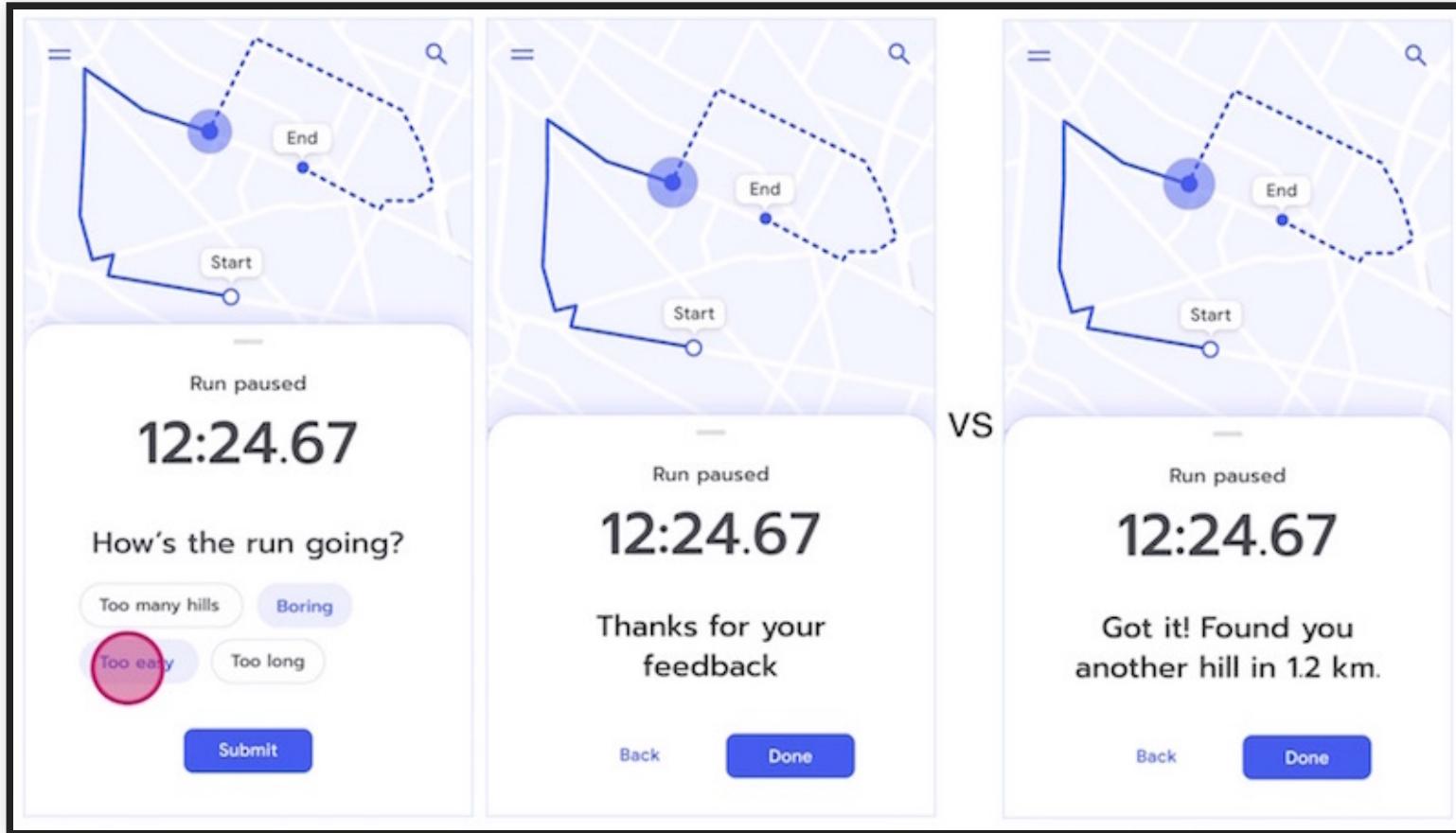
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 - Align feedback with improving interactions (and AI)
 - Acknowledge user feedback & respond immediately
- In addition to feedback, provide a way for user to adjust AI behavior

RESPONDING TO FEEDBACK



- When possible, respond to feedback with an adjustment to AI behavior

<https://pair.withgoogle.com/chapter/feedback-controls/>

GIVING USER CONTROL



- Provide a mechanism for user to adjust system behavior

GIVING USER CONTROL OVER ML BEHAVIOR



Adjust how aggressive you would want the Scheduling Assistant to be in detecting meetings in your emails:



Fewer detections
some requests
might be missed



More detections
more non-requests
might be suggested



- Provide a mechanism for the user to control the types of ML errors
- Scheduling assistant: Adjust thresholds to achieve trade-offs between precision vs recall

USER FEEDBACK IN VOICE ASSISTANTS

"All of the things that even Siri herself said she could do — for example 'I can send money via Venmo, just try and say this.' I tried and it didn't work, and maybe there are settings that I need to fix. But when those types of things happened, there was no button that said 'Hey, in order to make this work in the future, click this and we'll take you to the permissions or whatever'"

- Q. How do we collect user feedback? Implicit? Explicit?
- Q. What kind of control do we provide to the user?

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GUIDELINES FOR HUMAN-AI INTERACTIONS

Guidelines for Human-AI Interaction

INITIALLY	1  INITIALLY Make clear what the system can do. <small>Help the user understand what the AI system is capable of doing.</small>	2  INITIALLY Make clear how well the system can do what it can do. <small>Help the user understand how often the AI system may make mistakes.</small>					
DURING INTERACTION	3  DURING INTERACTION Time services based on context. <small>Time when to act or interrupt based on the user's current task and environment.</small>	4  DURING INTERACTION Show contextually relevant information. <small>Display information relevant to the user's current task and environment.</small>	5  DURING INTERACTION Match relevant social norms. <small>Ensure the experience is delivered in a way that users would expect, given their social and cultural contexts.</small>	6  DURING INTERACTION Mitigate social biases. <small>Ensure the AI system's language and behaviors do not reinforce undesirable and unfair stereotypes and biases.</small>			
WHEN WRONG	7  WHEN WRONG Support efficient invocation. <small>Make it easy to invoke or request the AI system's services when needed.</small>	8  WHEN WRONG Support efficient dismissal. <small>Make it easy to dismiss or ignore undesired system services.</small>	9  WHEN WRONG Support efficient correction. <small>Make it easy to edit, refine, or recover when the AI system is wrong.</small>	10  WHEN WRONG Scope services when in doubt. <small>Engage in disengagement or gracefully degrade the AI system's services when uncertain about a user's goals.</small>	11  WHEN WRONG Make clear why the system did what it did. <small>Enable the user to access an explanation of why the AI system behaved as it did.</small>		
OVER TIME	12  OVER TIME Remember recent interactions. <small>Maintain short-term memory and allow the user to make efficient references to that memory.</small>	13  OVER TIME Learn from user behavior. <small>Personalize the user's experience by learning from their actions over time.</small>	14  OVER TIME Update and adapt cautiously. <small>Limit sensitive changes when updating and adapting the AI system's behaviors.</small>	15  OVER TIME Encourage granular feedback. <small>Enable the user to provide feedback indicating their preferences during regular interaction with the AI system.</small>	16  OVER TIME Convey the consequences of user actions. <small>Immediately update, or convey how user actions will affect future behaviors of the AI system.</small>	17  OVER TIME Provide global controls. <small>Allow the user to globally customize what the AI system monitors and how it behaves.</small>	18  OVER TIME Notify users about changes. <small>Inform the user when the AI system adds or updates its capabilities.</small>

The Guidelines for Human-AI Interaction will help you create AI systems and features that are human-centered. We hope you use them throughout your design process – as you evaluate existing ideas, brainstorm new ones, and collaborate with the multiple perspectives involved in creating AI.

These guidelines synthesize more than 20 years of thinking and research in human-AI interaction. Learn more: <https://aka.ms/aiguidelines>.



HUMAN-AI INTERACTIONS

Human-AI interactions must be considered throughout the entire ML lifecycle!

- Requirements & design
 - Understand user needs & their mental models
 - Explicitly design system to match the mental model
- During interaction
 - Consider factors for interaction (automate vs augment, forcefulness, frequency)
- When errors occur
 - Provide an explanation & actionable information
 - Provide ways for user to adjust AI behavior
- Maintenance and evolution
 - Collect user feedback and improve model
 - Adjust system design to reduce mental model mismatch

SUMMARY

- Goal of usable design: Minimize interaction cost
 - Automation does not necessarily imply reduced cost!
- Interaction design considerations for AI
 - Modes of interaction: Automate or augment?
 - Mental model: User understanding of what AI is doing
 - Dealing with errors: Guide user towards recovery & prevention
 - Feedback and control: Align user feedback with AI improvement