

Designers can create designs that nudge users toward the most desirable option.

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Digital Nudging: Guiding Online User Choices through Interface Design

LIFE IS FULL of choices, often in digital environments. People interact with e-government applications; trade financial products online; buy products in Web shops; book hotel rooms on mobile booking apps; and make decisions based on content presented in organizational information systems. All such choices are influenced by

the choice environment, as reflected in this comment: “What is chosen often depends upon how the choice is presented.”¹⁶ Why? People have cognitive limitations, so their rationality is bounded,²⁷ and heuristics and biases drive their decision making.³⁴ Designers of choice environments, or “choice architects,”³² can thus use these heuristics and biases to manipulate the choice environment to subtly guide users’ behavior by gently “nudging” them toward certain choices.

These observations are more than theory. We are being nudged every

day of our lives. Supermarkets position items with the highest markups at eye level to nudge customers into making unplanned purchases. Likewise, supermarkets limit the number of units customers are allowed to buy,

» key insights

- **Heuristics and biases influence offline and online behavior.**
- **User-interface design influences choices, even unintentionally.**
- **Thorough design and testing can help achieve a designer’s intended behavioral effects.**

Figure 1. The decoy effect in reward-based crowdfunding; screenshot shows the decoy condition.

Book project

Project description

The children's book "Allen and the Wolf Pack" tells the story of Allen, a nine-year old boy who is trying to discover the reason for an unnaturally long winter. Although it is end of May, spring has not come, winter storms continue to bluster, and the winds carry the voices of howling wolves. When Allen goes to the forest to find the reason for this confused state, he gets lost, but he survives with the help of wolves that care for him until he is found. The book is full of adventure, mystery, excitement, and great characters.

Decision

Which of the following reward options do you prefer?

PAY \$ 10 - GET an eBook.

PAY \$ 20 - GET a hardback of the book.

PAY \$ 20 - GET an eBook and a hardback of the book.

Decoy
Option

thereby influencing their buying decisions; customers subconsciously anchor their decisions on the maximum number and adjust downward from there, resulting in purchases of greater quantities.³⁶ This effect has been demonstrated in the context of everyday items; for example, introducing a quantity limit of 12 cans of soup helped double the average quantity purchased from 3.3 to seven cans.³⁶ Nudges are not, however, used only by marketers trying to sell more products or services; for example, when asking people to consent to being an organ donor, simply changing defaults can influence people's choices. Setting the default to "dissent," whereby donors have to opt out, rather than "consent" whereby donors have to opt in, can nearly double the percentage of organ donors.¹⁵ These examples show that largely imperceptible nudges are effective in a variety of offline contexts.

As in offline environments, online environments offer no neutral way to present choices. Any user interface, from organizational website to mobile app, can thus be viewed as a digital choice environment.³⁷ Digital choice environments nudge people by deliberately presenting choices or organizing workflows, making digital nudging—"the use of user-interface design elements to guide people's behavior in digital choice environments"³⁷—a powerful tool in any choice architect's toolbox. Choosing the most effective nudge involves trade-offs, however, because predicting the consequences of implementing certain nudges is not always possible.

Existing guidelines for implementing nudges have been developed primarily for offline environments, and digital nudging has only recently begun to attract programmer interest; see, for example, Gregor and Lee-Archer¹⁰ and Weinmann et al.³⁷ In addition, guidelines that are effective offline may not always be directly transferred to a digital context; for example, online users are more willing to disclose information but are also more cautious about accepting default options.² To this end, this article shows how designers can consider the effects of nudges when designing digital choice environments.

Figure 2. The decoy effect in reward-based crowdfunding; adding a decoy option can make another option more attractive.

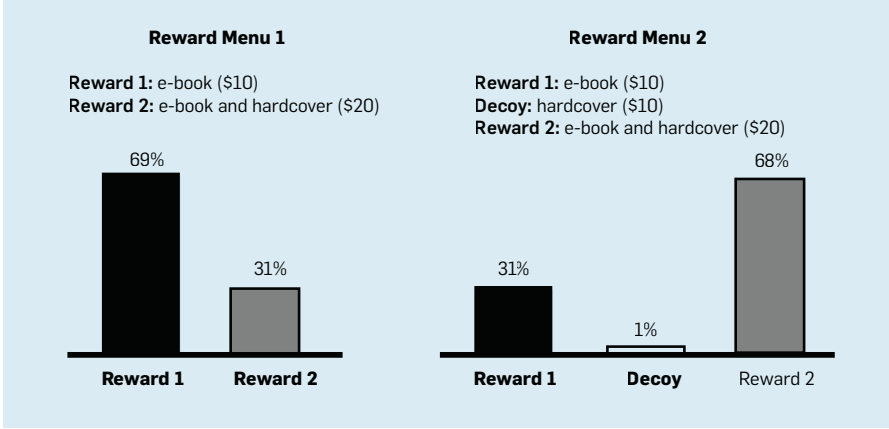


Figure 3. The scarcity effect in reward-based crowdfunding; limiting either reward changes pledging behavior of potential backers.

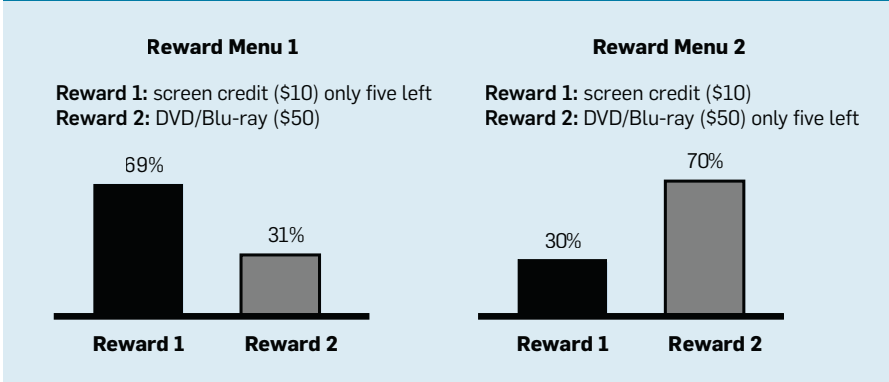
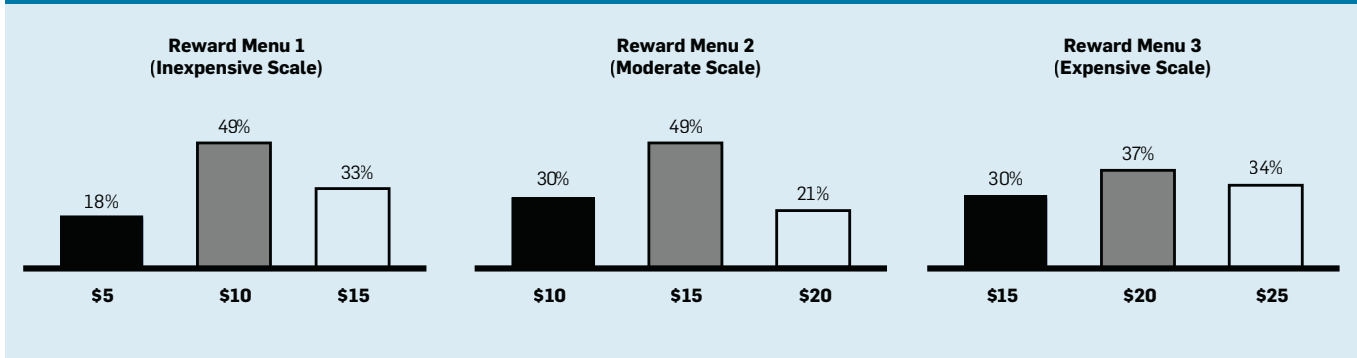


Figure 4. The middle-option bias in reward-based crowdfunding; even when the investment scale is increased, backers tended to select the middle option.



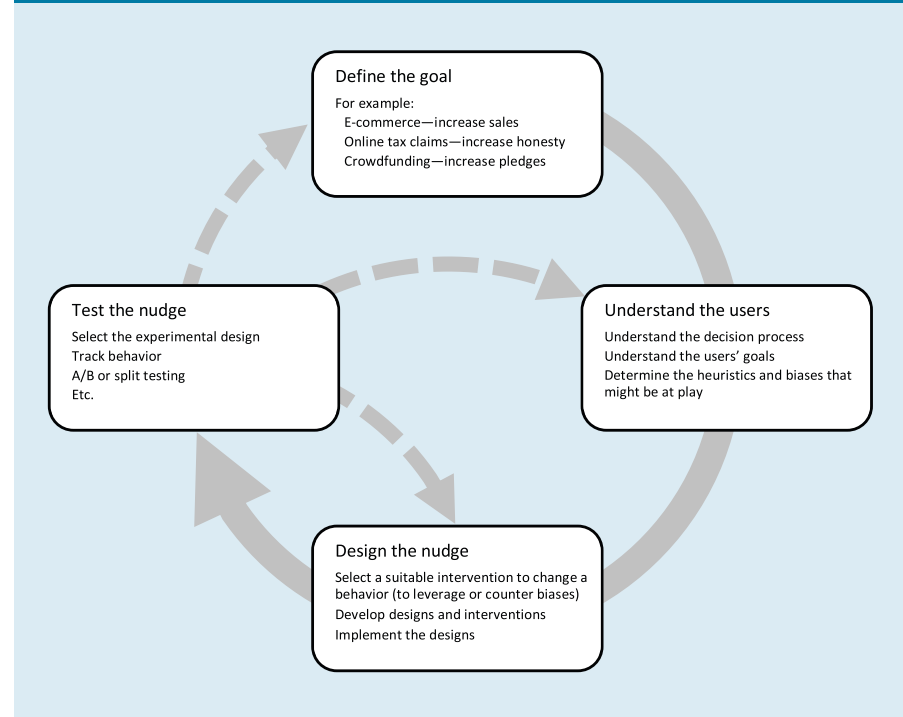
Guiding Choices

As in offline contexts, online decision making is almost always influenced by heuristics and biases; consequently, the concept of digital nudging applies not only to online consumers’ decision making but also to various other contexts, from e-health systems to social media apps to organizational information systems. Whereas such factors as presenting reviews or highlighting markdowns are well known for having a strong effect on user behavior in general, digital nudges influence decisions at the point and moment of decision making.^{a,22} In particular, digital nudging works by either modifying what is presented—the content of a choice^{6,35}—or how it is presented—the visualization of a choice—as in, say, changing the design of the user interface.¹⁶ For example, the mobile payment app Square presents a “tipping” option by default, so customers must select “no tipping” if they prefer not to give a tip; this modification is likely an attempt to nudge people into giving tips, motivating them to tip even where tipping is uncommon.³

To illustrate the effects of digital nudges, we briefly explore the results of a series of experiments in the context of reward-based crowdfunding.^{28,33,38} In reward-based crowdfunding, project creators collect small amounts of money from a large number of people, or “backers.” Backers pledge money for projects and receive non-financial

a Digital nudging, with its focus on the design of digital choice environments, can be viewed as a subset of persuasive computing/technology, which is generally defined as technology designed to change attitudes or behaviors and includes aspects of human-computer interaction beyond interface design.^{8,26}

Figure 5. Designing digital nudges follows a cycle; based on Datta and Mullainathan⁵ and Ly et al.¹⁹



rewards in return (such as an e-book).¹ To test how digital nudges influence backers’ pledges, researchers at the University of Liechtenstein modified the content and/or visualization of a choice environment to nudge backers toward a particular option through three particular heuristics and biases, known as the “decoy effect,”³³ “scarcity effect,”³⁸ and “middle-option bias.”²⁸

Decoy effect. The decoy effect increases an option’s attractiveness by presenting the option alongside an unattractive option no one would reasonably choose—the decoy.¹³ In a study conducted in the context of crowdfunding (N = 96), the researchers showed how decoys can nudge users to select certain rewards;³³ when backers were presented with a choice of receiving

an e-book in return for a \$10 pledge or both an e-book and a hardcover book for a \$20 pledge, most backers chose to pledge \$10. However, when a third option—a decoy nudge—was included that offered only the hardcover book in return for a \$20 pledge (see Figure 1), most backers chose to pledge \$20 to receive both the e-book and the hardcover book. Including the decoy option thus led many backers to move from the \$10 pledge to the \$20 pledge (see Figure 2).

Scarcity effect. People tend to perceive scarce items as more attractive or desirable.⁹ In the context of crowdfunding (N = 166), the researchers showed that limiting the availability of rewards—a “scarcity nudge”—can lead them to choose a particular re-

ward.³⁸ For a fictitious movie project, backers were offered a choice between two rewards: pledge \$10 to be listed in the screen credits or pledge \$50 to receive the movie on a DVD/Blu-ray disc (see Figure 3). When the availability of the low-price reward was limited, 69% of the backers chose that reward, as in Figure 3, left side, whereas when the availability of the high-price reward was limited, 70% chose that reward, as in Figure 3, right side. Merely presenting information about the limited availability of either reward, even the higher-price one, thus caused more backers to choose that reward.

Middle-option bias. People presented with three or more options (ordered sequentially, as by price) tend to select the middle option.⁴ Testing the effect of the middle-option bias in the context of crowdfunding (N = 282), the researchers showed that backers can be nudged into choosing the reward presented in the middle.²⁸ They tested it by varying the pledges of the offered rewards by, in particular, shifting the scales such that Condition 1: \$5, \$10, \$15; Condition 2: \$10, \$15, \$20, and Condition 3:

\$15, \$20, \$25. The researchers told the participants that their pledge would be doubled as a reward if the project would be successful. However, irrespective of the scale, most backers tended to choose the middle option, and by shifting the scales, the researchers could nudge the participants toward selecting rewards associated with higher pledge amounts (see Figure 4).

These examples show that designers can create digital nudges on the basis of psychological principles of human decision making to influence people’s online behavior. Unintended effects may arise, however, if designers of digital choice environments are unaware of the principles. For example, in the context of crowdfunding, presenting decoys or limiting the availability of rewards without considering their effect can unintentionally lead backers to select lower-price rewards; that is, as virtually all user-interface design decisions influence user behavior,^{20,30} designers must understand the effects of their designs so they can choose whether to nudge users or reduce the effects of nudges.

Designing a Digital Nudge

While a number of researchers have suggested guidelines for selecting and implementing nudges in offline contexts,^{5,6,16,19,21,31} information systems present unique opportunities for harnessing the power of nudging. For example, Web technologies allow real-time tracking and analysis of user behavior, as well as personalization of the user interface, and both can help test and optimize the effectiveness of digital nudges; moreover, mobile apps can provide a wealth of information about the context (such as location and movement) in which a choice is made. Given these advantages, information systems allow rapid content modification and visualization to achieve the desired nudging effect.

Drawing on guidelines for implementing nudges in offline contexts, we now highlight how designers can create digital nudges by exploiting the inherent advantages of information systems. Just as developing an information system follows a cycle, as in, say, the systems development life cycle—planning, analysis, design, and implementation—so does designing choices to nudge users (see Figure 5)—define the goal, understand the users, design the nudge, and test the nudge. We discuss each step in turn, focusing on the decisions designers must make.

Step 1: Define the goal. Designers must first understand an organization’s overall goals and keep them in mind when designing particular choice situations. For instance, the goal of an e-commerce platform is to increase sales, the goal of a governmental taxing authority’s platform is to make filing taxes easier and encourage citizens to be honest, and the goal of project creators on crowdfunding platforms is to increase pledges and overall donation amounts. These goals determine how choices are to be designed, particularly the type of choice to be made. For example, subscribing to a newsletter is a binary choice—yes/no, agree/disagree—selecting between items is a discrete choice, and donating monetary amounts is a continuous choice, though it could also be presented as a discrete choice. The type of choice determines the nudge to be used (see the table here). The choice architect, however, must consider not only the goals but also the ethical im-

Applying the digital nudging design cycle (selected examples).

Step 1 Type of choice to be influenced	Step 2 Heuristic/Bias	Step 3 Example design elements and user-interface patterns and possible nudges and mechanisms
Binary (yes/no)	Status quo bias (defaults)	Radio buttons (with default choice)
Discrete choice (such as two products)	Status quo bias (defaults)	Use of defaults in Radio buttons Check boxes Dropdown menus
	Decoy effect	Presentation of decoy option(s) in Radio buttons Check boxes Dropdown menus
	Primacy and recency effect	Positioning of presentation of desired option(s) Earlier (primacy) Later (recency)
	Middle-option bias	Addition of higher- and lower-price alternatives around preferred option Ordering of alternatives Modification of the option scale
Continuous	Anchoring and adjustment	Variation of slider endpoints Use of default slider position Predefined values in text boxes for quantities
	Status quo bias (defaults)	Use of default slider position
Any type of choice	Norms	Display of popularity (social norms) Display of honesty codes (moral norms)
	Scarcity effect (loss aversion)	Use of default slider position

plications of deliberately nudging people into making particular choices, as nudging people toward decisions that are detrimental to them or their wellbeing is unethical and might thus backfire, leading to long-term negative effects for the organization providing the choice.³⁰ In short, overall organizational goals and ethical considerations drive the design of choice situations, a high-level step that influences all subsequent design decisions.

Step 2: Understand the users. People's decision making is susceptible to heuristics and biases. Heuristics, commonly defined as "rules of thumb,"¹⁴ can facilitate human decision making by reducing the amount of information to be processed when addressing simple, recurrent problems. Conversely, heuristics can influence decisions negatively by introducing cognitive biases—systematic errors—when one faces complex judgments or decisions that should require more extensive deliberation.⁷ Researchers have studied a wide range of psychological effects that subconsciously influence people's behavior and decision making.^b In addition to the middle-option bias, decoy effect, and scarcity effect described earlier, common heuristics like the "anchoring-and-adjustment" heuristic, or people being influenced by an externally provided value, even if unrelated; the "availability" heuristic, or people being influenced by the vividness of events that are more easily remembered; and the "representativeness" heuristic, or people relying on stereotypes when encountering and assessing novel situations,³⁴ influence how alternatives are evaluated and what options are ultimately selected. Other heuristics and biases that can have a strong effect on choices include the "status quo bias," or people tending to favor the status quo so they are less inclined to change default options;¹⁸ the "primacy and recency effect," or people recalling options presented first or last more vividly, so those options have a stronger influence on choice;²⁴ and "appeals to

Questions Designers Need to Address

Define goals:

- ▶ What is the use scenario?
- ▶ What are the overall organizational goals?
- ▶ What specific goals are to be achieved in this situation?
- ▶ What are the ethical implications of nudging people into making a certain decision?

Understand the decision process:

- ▶ What are the users' goals?
- ▶ What are the users' decision-making processes?
- ▶ What heuristics might influence users' choices?

Design the nudge:

- ▶ What types of nudges could counter the influence of biases?
- ▶ What types of nudges could increase the influence of biases?
- ▶ What nudges could inadvertently influence users' choices?
- ▶ How can the design of the user interface be modified to include the preferred nudges?
- ▶ How can we analyze users' behavior to adapt the choice environment dynamically?

Test the nudge:

- ▶ How effective are the various nudges?
- ▶ Does the effectiveness differ across users?
- ▶ Do the nudges fit the context and the goals?
- ▶ Do we have a thorough understanding of the users' decision-making process?

norms," or people tending to be influenced by the behavior of others.²³ Understanding these heuristics and biases and the potential effects of digital nudges can thus help designers guide people's online choices and avoid the trap of inadvertently nudging them into decisions that might not align with the organization's overall goals.

Step 3: Design the nudge. Once the goals are defined (see Step 1: Define the goal) and the heuristics and biases are understood (see Step 2: Understand the users), the designer can select the appropriate nudging mechanism(s) to guide users' decisions in the designer's intended direction. Common nudging frameworks a designer could use to select appropriate nudges include the Behavior Change Technique Taxonomy,²¹ NUDGE,³¹ MINDSPACE,⁶ and Tools of a Choice Architecture.¹⁶ Selecting an appropriate nudge and how to implement it through available design elements, or user-interface patterns, is determined by both the type of choice to be made—binary, discrete, or continuous^c—and

the heuristics and biases at play; see the table for examples. For example, a commonly used nudge in binary choices is to preselect the desired option to exploit the status quo bias. When attempting to nudge people in discrete choices, choice architects can choose from a variety of nudges to nudge people toward a desired option. For example, in the context of crowdfunding, with the goal of increasing pledge amounts, choice architects could present the desired reward option as the default option; add (unattractive) choices as decoys; present the desired option first or last to leverage primacy and recency effects; or arrange the options so as to present the preferred reward as the middle option. When attempting to nudge people in continuous choices (such as when soliciting monetary donations), choice architects could pre-populate input fields (text boxes) with a particular value so as to exploit the "anchoring and adjustment" effect. Likewise, when using a slider to elicit numerical responses, the position of the slider and the slider endpoints serve as implicit anchors. Presenting others' choices next to rewards to leverage people's tendency to conform to norms or presenting limited availability of rewards to exploit the


b See Stanovich²⁰ for a taxonomy of rational thinking errors and biases; see also Wikipedia for an extensive list of cognitive biases that influence people's online and offline behavior (https://en.wikipedia.org/wiki/List_of_cognitive_biases).

c In most cases, the type of decision is an externality, and many decisions allow for only one type; for example, consenting to something (whether organ donation or signing up for a newsletter) would normally always be a binary choice—yes/no.


scarcity effect can be used to nudge people in binary, discrete, or continuous choices.

As the same heuristic can be addressed through multiple nudges, in most situations, designers have a variety of “nudge implementations” at their disposal. Unlike in offline environments, implementing nudges in digital environments can be done at relatively low cost, as system designers can easily modify a system’s user interface (such as by setting defaults, displaying/hiding design elements, or providing information on others’ pledges). Likewise, digital environments enable dynamic adjustment of the options presented on the basis of certain attributes or characteristics of the individual user (such as when a crowdfunding platform presents particular rewards depending on the backers’ income, gender, or age). Notwithstanding the choice of nudges, designers should follow commonly accepted design guidelines for the respective platforms (such as Apple’s Human Interface Guidelines and Microsoft’s Universal Windows Platform design guidelines) to ensure consistency and usability.

Step 4: Test the nudge. Digital environments allow alternative designs to be generated easily, so their effects can be tested quickly, especially when designing websites. The effectiveness of digital nudges can be tested through online experiments (such as A/B testing and split testing). Testing is particularly important, as the effectiveness of a nudge is likely to depend on both the context and goal of the choice environment and the target audiences. For example, a digital nudge that works well in one context (such as a hotel-booking site like <https://www.booking.com>) may not work as well in a different context (such as a car-hailing service like <https://www.uber.com>); such differences may be due to different target users, the unique nature of the decision processes, or even different layouts or color schemes on the webpages; a hotel may use colors and shapes that evoke calmness and cleanliness, whereas a car-hailing service may use colors and shapes that evoke speed and efficiency. As choice



Big-data analytics can be used to analyze behavioral patterns observed in real time to infer users’ personalities, cognitive styles, or even emotional states.



architects have various nudge implementations at their disposal, thorough testing is thus imperative for finding the nudge that works best for a given context and users.

Especially in light of the increasing focus on integrating user-interface design and agile methodologies, using discount usability techniques (such as heuristic evaluation, as introduced by Nielsen²⁵) is often recommended to support rapid development cycles (see, for example, Jurca et al.¹⁷). Likewise, agile methodologies include the quick collection of feedback from real users. However, such feedback from conscious evaluations should be integrated with caution because the effects of nudges are based on subconscious influences on behavior, and experimental evaluations can provide more reliable results. If a particular nudge does not produce the desired effect, a first step for system designers is to evaluate the nudge implementation to determine whether the nudge is, say, too obvious or not obvious enough (see Step 3: Design the nudge). In some instances, though, reexamining the heuristics or biases that influence the decision-making process (see Step 2: Understand the users) or even returning to Step 1: Define the goal and redefining the goals may be necessary (see the sidebar, “Questions Designers Need to Address”).

Conclusion


Understanding digital nudges is important for the overall field of computing because user-interface designers create most of today’s choice environments. With increasing numbers of people making choices through digital devices, user-interface designers become choice architects who knowingly or unknowingly influence people’s decisions. However, user-interface design often focuses primarily on usability and aesthetics, neglecting the potential behavioral effects of alternative designs. Extending the body of knowledge of the computing profession through insights into digital nudging will help choice architects leverage the effects of digital nudges to support organizational goals. Choice architects can use the digital nudging

design cycle we have described here to deliberately develop such choice environments.

One final note of caution is that the design of nudges should not follow a “one-size-fits-all” approach, as their effectiveness often depends on a decision maker’s personal characteristics.¹⁶ In digital environments, characteristics of users and their environment can be inferred from a large amount of data, allowing nudges to be tailored. System designers might design the choice environment to be adaptive on the basis of, say, users’ past decisions or demographic characteristics. Likewise, big-data analytics can be used to analyze behavioral patterns observed in real time to infer users’ personalities, cognitive styles, or even emotional states.¹² For example, Bayesian updating can be used to infer cognitive styles from readily available clickstream data and automatically match customers’ cognitive styles to the characteristics of the website (such as through “morphing”¹¹). Designers of digital choice environments can attempt to “morph” digital nudges on the basis of not only the organizational goals but also users’ personal characteristics.

Any designer of a digital choice environment must be aware of its effects on users’ choices. In particular, when developing a choice environment, designers should carefully define the goals, understand the users, design the nudges, and test those nudges. Following the digital-nudging design cycle we have laid out here can help choice architects achieve their organizational goals by understanding both the users and the potential nudging effects so intended effects can be maximized and/or unintended effects minimized.

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