Evidence-based design heuristics for idea generation



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How do product designers create multiple concepts to consider? To address this question, we combine evidence from four empirical studies of design process and outcomes, including award-winning products, multiple concepts for a project by an experienced industrial designer, and concept sets from 48 industrial and engineering designers for a single design problem. This compilation of over 3450 design process outcomes is analyzed to extract concept variations evident across design problems and solutions. The resulting set of patterns, in the form of 77 Design Heuristics, catalog how designers appear to introduce intentional variation into conceptual product designs. These heuristics provide 'cognitive shortcuts' that can help designers generate more, and more varied, candidate concepts to consider in the early phases of design.

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ow do designers successfully create novel product concepts? One suggested approach is to first generate a wide range of concepts to consider (Cross, 1994; Liu, Bligh, & Chakrabarti, 2003). This requires the ability to create a large number of concepts that differ from each other so that the set of concepts covers the space of possible designs (Gero, 1990; Goel & Pirolli, 1992; MacLean, Young, Bellotti, & Moran, 1991; Simon, 1981). Logically, the idea generation process benefits from considering as many different concepts as possible (Akin & Lin, 1995; Atman, Chimka, Bursic, & Nachtman, 1999; Brophy, 2001; Liu et al., 2003). However, generating a diverse set of concepts can be challenging because designers tend to fixate on specific design specifications, which leads them to generate more concepts with similar features (Purcell & Gero, 1996; Sio, Kotovsky, & Cagan, 2015). For example, Jansson and Smith (1991) observed designers replicating similar solutions to concepts provided as examples, and even including their flaws. Across studies, designers appear to consider only a small set of related concepts when generating ideas (Ball, Evans, & Dennis, 1994; Chrysikou & Weisberg, 2005; Dong & Sarkar, 2011; Linsey et al., 2010;

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Purcell & Gero, 1996; Sio et al., 2015; Smith, 1998; Viswanathan & Linsey, 2013; Youmans & Arciszewski, 2014).

A number of approaches for facilitating idea generation during the early phases of conceptual design have been proposed (c.f. Clapham, 1997; Shah, Hernandez, & Smith, 2002; Smith, 1998). One approach distills knowledge about specific designs into an intermediate-level knowledge base by constructing composites from multiple examples. In Alexander's *pattern language* (Alexander, Ishikawa, & Silverstein, 1977), and Krippendorf's *design discourses* (2005), patterns common in successful design solutions are identified at a component level, linking the designer to a broad range of helpful guidance from past solutions in a refined form (Alexander et al., 1977). This composite knowledge about design has been referred to as *heuristic* knowledge (Fu, Yang, & Wood, 2015). Heuristics are described as 'mental shortcuts' that capture cognitive strategies that may lead to solutions (though not necessarily the best one) (Nisbett & Ross, 1980), and are ubiquitous in human reasoning (Goldstein et al., 2001). Heuristics capture important features of problem situations and solutions that tend to reoccur in experiences (Clancey, 1985).

In software design, Riel (1996) has described the heuristic approach as 'specific experience-based guidelines' that help developers make good decisions. Lawson (1979) observed architectural students solving puzzles through 'trial and error' heuristic approaches. Lawson (1980) concludes, 'An examination of protocols obtained from such closely observed design sessions reveal that most designers adopt strategies which are heuristic in nature... Heuristic strategies do not so much rely upon theoretical first principles as on experience and rules of thumb' (p. 132). When generating new concepts, designers appear at times to offer intuitive responses derived from 'large pools of experience' (Cross, 2011, p. 10) to make a 'best guess' at a new design. Consider the example in Figure 1, a desk chair that reclines to allow the user to lie beneath (rather than in front of) a computer screen.

In comparing this novel design to prototypical chairs, it is evident that the designer changed the user's direction of access. By moving the access point from in front of the screen to below it, an innovative design results. Further, this strategy, 'change direction of access,' may be a useful heuristic to apply in generating designs for other products. For example, applying the 'change direction of access' heuristic to a trackball controller may suggest side rather than top access, and accommodate thumb control rather than palm movements (see Figure 2). Design heuristics like this one may help designers create more, and more diverse, concepts, thereby increasing the likelihood that an innovative concept will result. Understanding how cognitive processes can be stimulated to generate design ideas may lead to more effective methods and tools to support conceptual design (Jin & Benami, 2010).



Figure 1 A design released by Altwork (http://altwork.com) positions the user under the workstation

In this paper, we examine evidence for design heuristics in the creation of multiple design concepts. First, we summarize prior research where design heuristics were derived from evidence in the field of product design, including approaches based on analysis of existing products and patents (e.g., Altshuller, 1984; Skiles et al., 2006). Next, we compile results across four research studies to identify a distinct set of heuristics evident in a diverse sample of design solutions. These solutions include an analysis of award-winning products created by many different designers. Uniquely, the present analysis examines design concepts from a professional designer working on a single design problem. In addition, two think-aloud protocol studies of industrial and engineering designers working on a novel design problem are included. These samples add value because they include multiple concepts generated for the same design problem. By considering alternative concepts, it is possible to observe how heuristics are used in the idea generation process, and how they facilitate exploring the space of concepts for a design problem. Compiling patterns observed across varied products, design tasks, and design processes, we identify a new set of 77 design heuristics. Each heuristic is presented with a written description and an example of its application in an existing consumer product. Finally, we discuss issues of the granularity of heuristic descriptions, and the use of heuristics as a concept generation tool for product designers.

l Heuristics in product design

How can we identify possible heuristics used in product design? Heuristics are learned from experience within a domain, and tend to be implicit and difficult to verbalize (Nisbett & Ross, 1980). The use of heuristics without conscious access has been documented in studies of experts including firefighters (Klein, 1993), scientists (Baker & Dunbar, 2000) and designers (Yilmaz & Seifert, 2011). However, this tacit knowledge about how to create designs

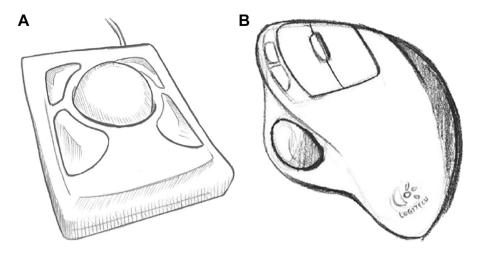


Figure 2 The original version of the Kensington Expert Mouse (www.kensington.com) used a center ball as a trackball, while the newer design by Logitech (www.logitech.com) positions the ball on the right side, under the thumb

may be observable by comparing designers' proposed solutions (Matthews, Wallace, & Blessing, 2000; Yilmaz, Seifert, Daly, & Gonzalez, 2016). Several existing heuristic approaches to idea generation have drawn conclusions based on empirical studies of product concepts (Perez, Linsey, Tsenn, & Glier, 2011) and design patents (Altshuller, 1984).

The theory of 'inventive problem solving' (known as TIPS or TRIZ) (Altshuller, 1984) involved identifying heuristics from successful patents in engineering. The TRIZ analysis focuses on identifying technical contradictions in mechanical engineering designs. For example, Ogot & Okudan (2007) describe a design tradeoff when 'increasing the stiffness of an airplane's wings to reduce vibration during flight (good) increases the weight of the plane (bad)' (p. 111). Altshuller (1984) analyzed thousands of engineering patents and abstracted forty principles, and noted that certain contradictions lend themselves to particular solutions. These were compiled into a contradiction matrix of system features (e.g., speed, weight, measurement accuracy) crossed with typical undesired results to index relevant design principles (Altshuller & Rodman, 1999; Altshuller, 1997, 2005; Orloff, 2003; Savransky, 2000). However, because TRIZ analysis requires the identification of technical tradeoffs first, it is most helpful for designs developed to the point of specific commitments to materials and mechanisms.

Learning to use the TRIZ system requires extensive training, effort and commitment (Ilevbare, Probert, & Phaal, 2013). The terminology and modeling methods are unique to TRIZ, and differ from those found in engineering design (Smith, 2003). However, in a classroom study with first-year engineering students, Ogot and Okudan (2007) trained teams of 4 students to use

TRIZ to generate concepts while other teams used traditional idea generation methods. They found that teams using the TRIZ method produced more unique solutions compared to other teams, along with more feasible concepts. This was replicated in another engineering classroom study where the TRIZ method was found to result in more novelty compared to sketch methods. In a third classroom study, engineering students using TRIZ improved the novelty and variety of concepts generated (Hernandez, Schmidt, & Okudan, 2013; Hernandez, Schmidt, Kremer, & Lin, 2014). Finally, an experimental study with graduate student and professional engineer teams found that TRIZ improved the novelty of solutions with only a ten minute training session (Chulvi, Gonzalez-Cruz, Mulet, & Aguilar-Zambrano, 2013).

Another approach to identifying design heuristics has examined existing products that 'transform,' or change into different configurations or states for use (Skiles et al., 2006). For example, a wooden chair may be designed to transform into a stepladder. Transformer products address each function set independently and at different times, while moving smoothly between states as needed (Weaver, Wood, Crawford, & Jensen, 2010). Based on analyses of 85 past patents, 40 analogies from nature, and 100 existing multistate products, three transformation design principles were extracted (expand/collapse, expose/cover, and fuse/divide) (Singh et al., 2007, 2009; Skiles et al., 2006; Weaver et al., 2008, 2010). A fourth principle, reorientation, was proposed in a later study (Haldaman & Parkinson, 2010). In addition, twenty subordinate 'facilitators' were extracted to support these principles. Example facilitators include using 'generic connections' to allow different modules to perform different functions; 'segmentation,' or dividing a single contiguous part into two or more parts; and 'fold,' or create relative motion between parts or surfaces by hinging, bending, or creasing. A study of engineering students found that encouraging the use of transformation principles and facilitators resulted in the generation of 25% more concepts (Weaver et al., 2009).

Several other studies have analyzed product designs to derive heuristics for idea generation. One study examined 197 award-winning innovative products, and organized the identified design features into categories (Saunders, Seepersad, & Hölttä-Otto, 2011). The thirteen 'innovation characteristics' identified in this analysis include 'additional function,' 'modified size,' 'expanded usage environment,' and 'user interactions.' Another study identified 'consumer variation' heuristics for designing for user differences (Cormier, Literman, & Lewis, 2011). Through an analysis of 31 product lines with 645 product models, 20 heuristics are identified and categorized into function, form, and information and control groups. Examples include, *Utilize (re)configurability when the product architecture is specific to handedness, Use system (re)configurability facilitated by modules when desired functionality is decoupled*, and *Utilize materials which have built-in flexibility for aesthetic modification*. Finally, a study of 46 bio-inspired products and systems resulted

in six 'scaling principles:' *change energy source, simplify system, change method, combine functions, directly transfer components*, and *change parameters* (Perez et al., 2011).

In these different approaches, various design heuristics were identified based on the design evidence considered. These approaches differ in the observed designs, with a focus on transforming (dual function) products in Weaver et al. (2010), award-winning innovative products in Saunders et al. (2011), consumer variation product lines in Cormier et al. (2011), and products at varied scales (in Perez et al., 2011). TRIZ (Altshuller, 2005) stands out for the large number of patents analyzed. However, in all of these approaches, only a final 'winning' concept is considered. The present study also includes a large sample of designs for award-winning consumer products. But uniquely, the present study adds samples of multiple candidate concepts generated by designers for a single design problem. The opportunity to observe the set of candidate concepts generated by a designer for a given problem provides a richer sample of variations among concepts than is captured by final product designs. Observations from a long-term design project by a very experienced designer added hundreds of concepts for a single design problem. The observation of idea generation sessions (rather than solely the 'winning,' final product) provides more evidence about how designers introduce variations in their concept sets through what Lawson (2012) calls 'knowing by doing.' By consolidating results across four empirical studies of concept generation, with varied contexts and more concepts sampled, we hoped to detect a broad array of design heuristics.

2 Method

For the present study, we compiled a larger database from four prior empirical studies (described in Table 1). The goal was to create a larger, rich dataset of design concepts from three different contexts, multiple design problems and multiple designers. The four studies included diverse datasets: (1) award-winning products from a wide range of consumer domains, (2) an expert industrial designer's sequential concept sketches from a two-year solo design project, and (3) a protocol study of engineering designers where student and practicing designers' think-aloud protocols were recorded as they worked on a novel product design task. A fourth study (4) replicated the think-aloud protocol study with industrial designers in order to compare concepts from the two design disciplines.

The process for extracting a design heuristic from award-winning product was as follows: For observed design concepts, major elements and key features of each concept were analyzed for functionality, form, and user-interaction features. A content analysis of the needs, design criteria, functions, and the design solution was performed for each concept. Then, potential heuristics were

Study	Research question Data collection		Source	
Study 1.	What are the strategies that	400 award-winning products from	Yilmaz, Seifert	
Product Analysis	successful designers use to create novel products?	a diverse range of design domains.	et al. (2016).	
Study 2.	How does an experienced	218 sequential concepts created by	Yilmaz and	
Case Study	designer add variation to concepts within a single long-term design problem?	an expert industrial designer over two years for a single design project (a universal access bath within an existing home).	Seifert (2011).	
Study 3.	How do different designers	Think-aloud protocols from 36	Daly, Yilmaz	
Protocol Analysis	create concepts within a single novel design task?	engineers at varying levels of expertise as they designed a novel product (a portable solar oven) in a 25-min session, with a total of 179 concepts	et al. (2012).	
~		generated.		
Study 4.	How does Design Heuristic	Think-aloud protocols from	Yilmaz, Daly	
Protocol Analysis	use differ among designers from different design disciplines?	12 industrial designers at varying levels of expertise working with the problem (in Study 3) for a total of 68 concepts generated.	et al. (2015).	

Table 1 Separate empirical studies of design concepts included in the cumulative database

hypothesized and design criteria for their application were identified. Other concepts in the dataset with the same design features were compared in order to explore commonalities in candidate heuristics. Finally, a heuristic would be defined at a level of generality that applied to multiple products, but was still specific to the observed design solution. For example, one heuristic was described as the 'hollowing out' of material, such as a brush handle with its mass reduced by using a hollow cylinder for a handle. This kept the heuristic's description as close as possible to the observed concepts; for example, different heuristics captured reducing material through flattening or folding. This extraction approach catalogs more specific innovations while ensuring the heuristics are general enough to fit several different observed concepts. Singh and colleagues (2009) describe a similar extraction method in their analysis of transforming products.

The product images in Figure 3 illustrate the process of extracting a heuristic from two of the 400 award-winning products included in the study. The first image shows a new product – a paint roller – where a commonly used mechanism in ballpoint pens (the ink storage and roller) is *applied in a new context* to solve the problem of delivering wall paint touchups. This heuristic also appears in the second image as a brush repurposed as a desk organizer design. The heuristics extracted identify independent components of the design, and are not exhaustive, such that other features of these designs might serve to identify other possible heuristics. In the first image, a second heuristic is also observable; namely, *Synthesize Functions*, where both paint storage and applicator are combined in the design. In this way, observed concepts sometimes provided evidence of multiple heuristics.

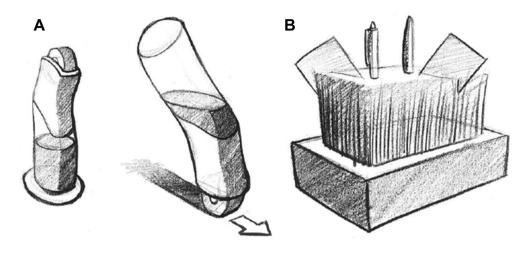


Figure 3 Example designs exhibiting the design heuristic, Apply existing mechanism in new way. On the left, the Rubbermaid Paint Buddy is a touch-up paint roller with onboard paint storage with a mechanism similar to ballpoint pens (http://www.idsa.org/awards/idea/computer-equip-ment/rubbermaid-paint-buddy). On the right, a desk organizer for pens and cards makes use of brush bristles to catch and hold these objects (http://ideasmodern.com/ideas/playful-pencil-organizer-pratonzolo/)

This extraction method for identifying design heuristics in existing products was applied to the design concepts in the remaining three studies (Daly, Christian, Yilmaz, Seifert, & Gonzalez, 2012; Daly, Yilmaz, Christian, Seifert, & Gonzalez, 2012; Yilmaz & Seifert, 2011; Yilmaz, Seifert et al., 2016). Study 2 provided 218 concepts created by a single, very experienced industrial designer over a two-year period (Yilmaz & Seifert, 2011). The design problem was to create a universal access bathroom to be installed in private homes. The designer worked on a large paper scroll to preserve his concepts as they were created. By examining sequential concepts, transitions between candidate concepts were evident. Across this set of designs, we observed that the same specific heuristics appeared repeatedly in this designer's work. For example, one heuristic addressed a change in how the functions of the product were controlled. In this example concept, the designer arranged components around the same central structure (a plumbing tube) (see Figure 4). This strategy was then observed in other designs, leading to a proposed heuristic, Align components around the center. This concept also suggests other heuristics, allowing the user to reorient the product according to their height, and *repeat* design elements.

The concepts collected from Studies 3 and 4 involved a 'think aloud' protocol (Dorst & Cross, 2001; Ericsson & Simon, 1993) of engineering and industrial designers' process while creating solutions for a novel product problem (the design of a solar oven for use in an outdoor setting). Forty-eight designers generated 247 different concepts for this single design problem. For example, one of the designers generated a concept for a portable backpack container that allowed cooking using sunlight (see Figure 5).

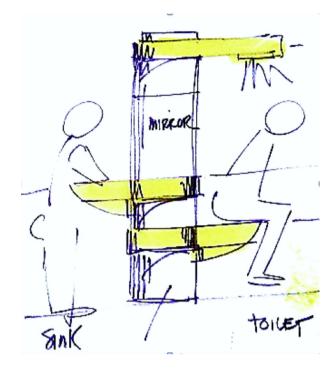


Figure 4 Example concept combining the heuristics Align components around center, Allow user to reorient, and Repeat. (Courtesy of Allen Samuels, Industrial Designer.)

> Next, three independent coders with advanced degrees (one with an M.F.A. in industrial design, one with a Ph.D. in engineering education, and one a senior student in mechanical engineering) worked as a team to examine each concept in the collected database. The coders considered each concept both individually and in its concept set sequence for evidence of heuristic use. The three coders worked collaboratively to refine heuristic definitions, and all decisions about identified heuristics were argued to consensus. Because the coders worked as a team during the extensive analysis, no measure of reliability was possible. The collaborative identification of heuristic use across these observed concepts occurred over a period of six weeks.

3 Results

The analysis of this combined sample of 3457 products and design concepts across four empirical studies resulted in the observation of 77 distinct design heuristics. Each of the identified heuristics was observed in at least four different concepts across the sample datasets. These heuristics addressed design goals such as adding functionality, using fewer resources, saving space, providing visual consistency, and forming new relationships among design elements. The 77 Design Heuristics are shown in Figure 6. This set of 77 Design Heuristics includes only those necessary to account for the data in these four studies. Each Design Heuristic is described, and illustrated with a commercial product where the heuristic is evident.

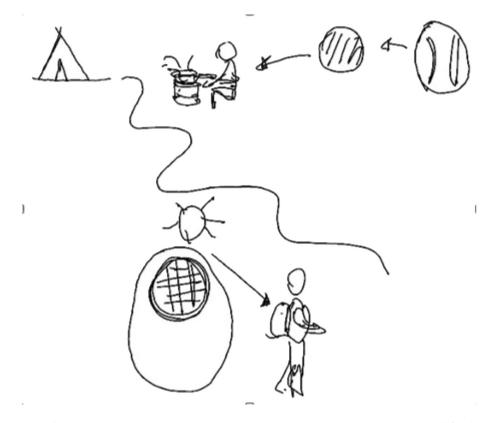


Figure 5 A concept for a solar oven generated by a designer using an Attach product to user heuristic, along with an Add functions heuristic. The industrial designer described a context in which the user was a hiker, and designed an integrated backpack with a heating element and pot attached to it. This would allow the user to warm food throughout the day while traveling

The observations supporting this set of 77 Design Heuristics (capitalized when referring to heuristics from this set) are shown in Table 2. An important feature of this compilation of heuristics across studies is that each heuristic was observed multiple times (at least four) in different products and product concepts, and all were observed in solutions from more than one designer. The sole exception is *expose interior*, which was observed only one concept (in Study 4) but included because it is well known (e.g., watches or clocks) and may facilitate the goal of considering a variety of candidate concepts.

Only seven heuristics were observed in just one of the four studies. The frequency of observation for each heuristic in the compiled dataset ranged from 4 to 274, indicating high variation in frequency of use. Only 12% of the observed instances of Design Heuristic use occurred in Study 1 (product analysis), but over half of the Design Heuristics (39) were observed in that particular study. Across the four studies (analyzed sequentially), the number of new heuristics identified decreased from 39 to 25 to 5 to 1. Even though the design problem and setting changed with each study, a great number of

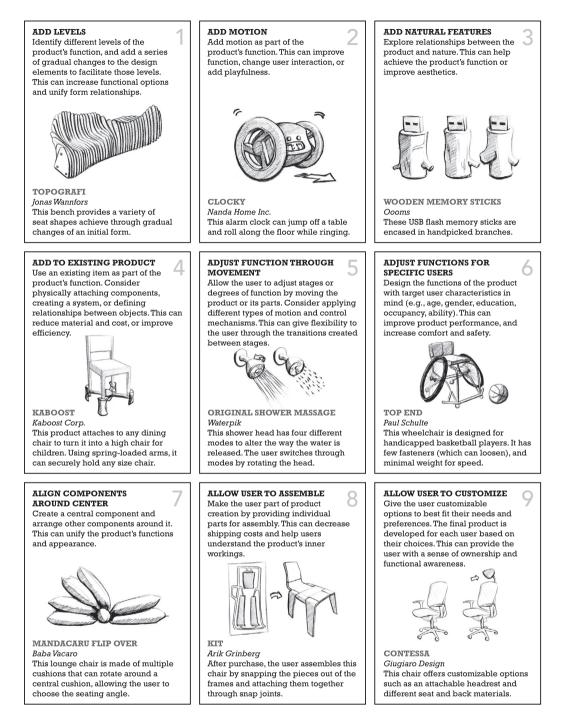
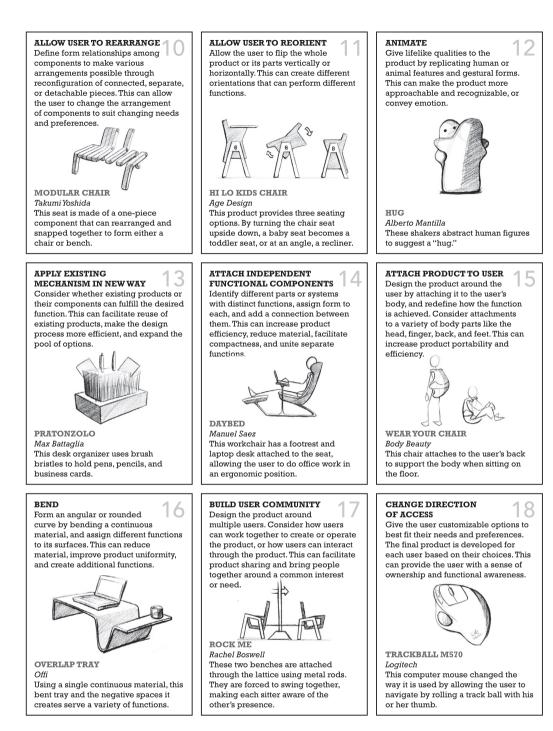
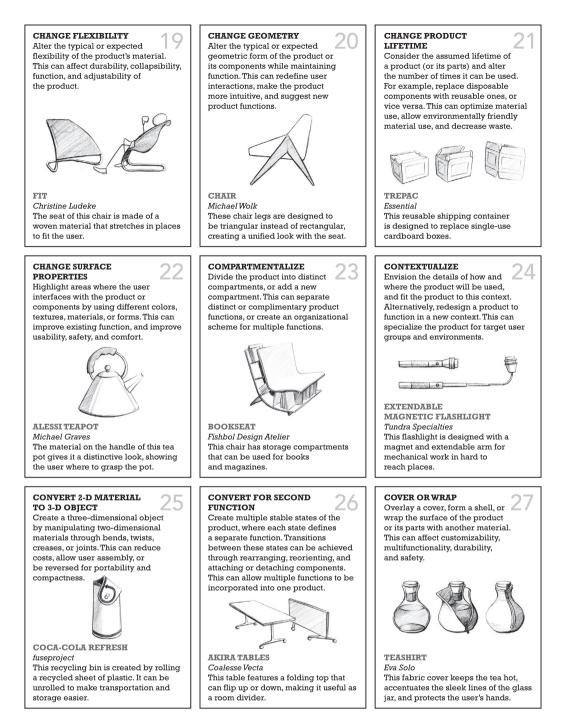
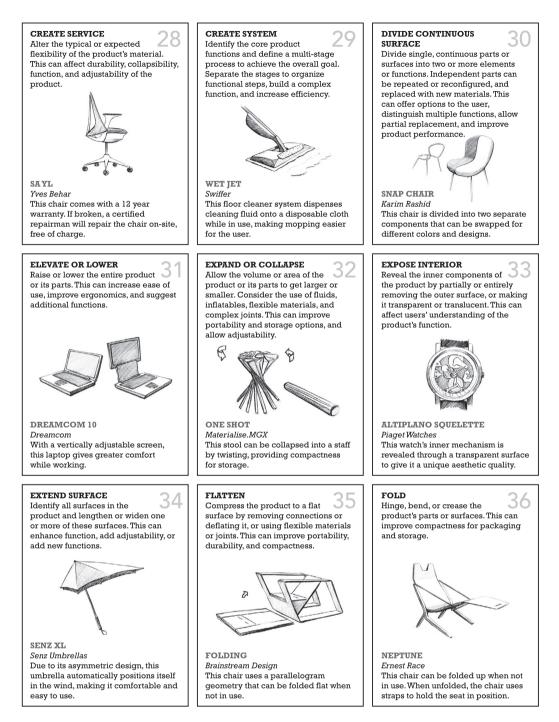


Figure 6 The 77 Design Heuristics identified across four studies of award-winning product designs, a solo professional design project, and protocol studies of engineers and industrial designers working on a novel problem. Each is illustrated with a description and an example consumer product where the Design Heuristic is evident. (Courtesy of Design Heuristics, Inc.)



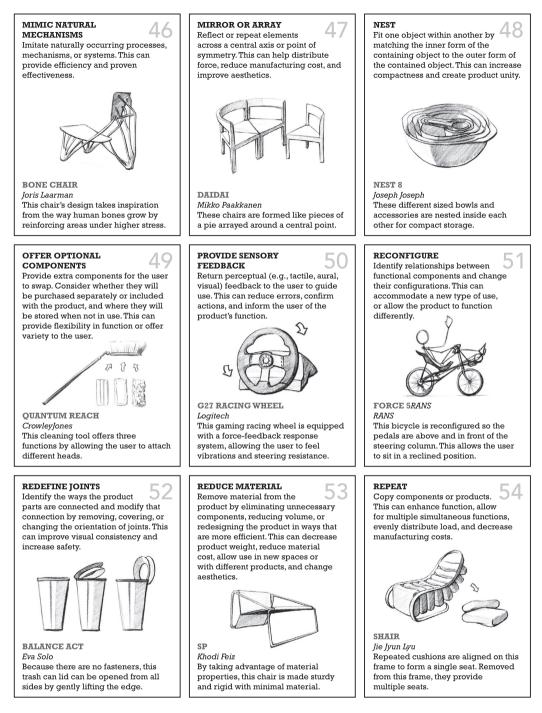
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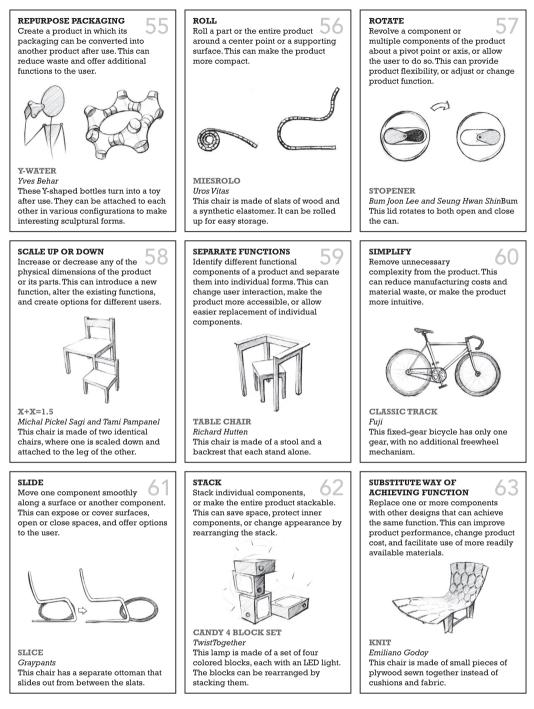


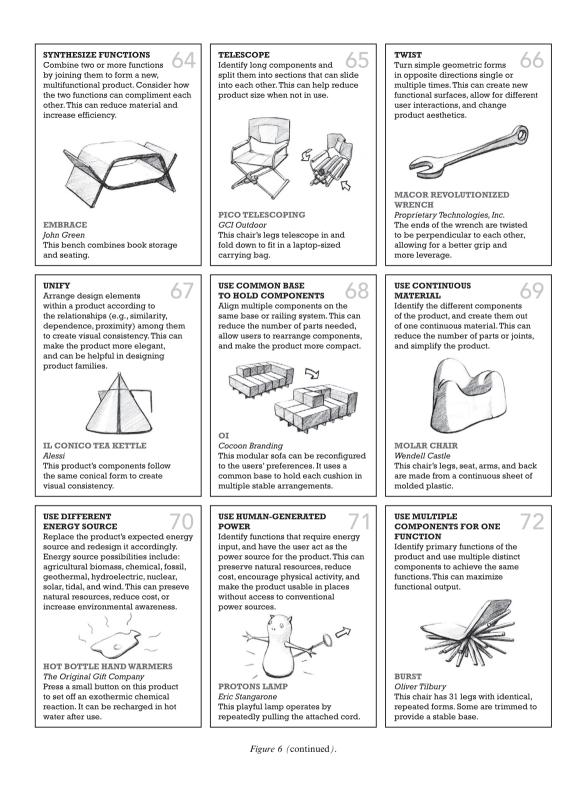
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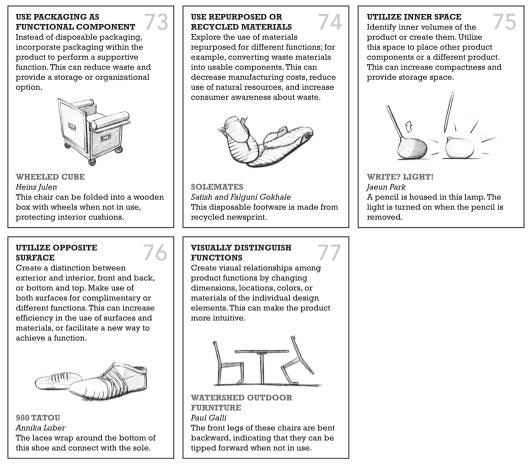


Figure 6 (continued).

previously identified heuristics were observed in each study. This suggests the identification of heuristics had reached a point of saturation across the entire set of concepts in this compiled dataset.

The data observed led to seventy heuristics across the four studies. Splitting seven observed heuristics into two separate heuristics subsequently created seven new heuristics. For example, *Replace materials with recycled ones* included both the use of recycled material and recyclable products. This heuristic was then redefined into two: *Use repurposed or recycled materials*, and *Make product recyclable*. The intent in adding these seven heuristics was to provide clarification of their meaning given that two subcategories appeared evident in the concepts reviewed (see Table 3).

Across the four studies, the majority (51%) of the design heuristic observations occurred in Study 2. This study analyzed designs from a single industrial

	Design heuristic	Study 1 Product analysis	Study 2 Longterm project	Study 3 Engineer protocols	Study 4 Ind. Design protocols	Total
1	Add levels	0	3	0	6	9
2	Add motion	4	0	4	0	8
3	Add natural features – split from 46					
4	Add to existing product	12	49	32	19	112
5	Adjust function through movement	17	76	35	12	140
6	Adjust functions for specific users	23	50	1	1	75
7	Align components around center	5	22	0	0	27
8	Allow user to assemble	4	0	0	0	4
9	Allow user to customize $-$ split from 6					
10	Allow user to rearrange – split from 51					
11	Allow user to reorient	5	0	0	0	5
12	Animate	16	0	0	0	16
13	Apply mechanism in new way	21	64	14	8	107
14	Attach independent functional components	0	145	95	34	274
15	Attach product to user	6	0	2	1	9
16	Bend	0	16	4	4	24
17	Build user community	4	0	1	1	6
18	Change direction of access	13	211	5	0	229
19	Change flexibility	8	12	17	10	47
20	Change geometry	0	12	25	0	37
21	Change product lifetime	8	4	0	2	14
22	Change surface properties	0	8	6	6	20
23	Compartmentalize	0	12	7	3	22
24	Contextualize	14	135	0	0	149
25	Convert 2-D material to 3-D object	9	8	4	1	22
26	Convert for second function	0	8	8	3	19
27	Cover or wrap	4	18	100	36	158
28 29	Create service – <i>split from 29</i>	6	0	14	4	24
29 30	Create system Divide continuous surface	6 0	31	32		24 74
30 31	Elevate or lower	0	31	52 66	11 27	124
32	Expand or collapse	11	49	10	4	74
32	Expose interior	0	49	10	4	1
34	Expose interior Extend surface	0	28	0 7	5	40
35	Flatten	0	28	4	3	40 10
36	Fold	0	25	48	23	96
37	Hollow out	0	0	4	3	7
38	Impose hierarchy on functions	11	0	3	8	22
39	Incorporate environment	0	8	6	4	18
40	Incorporate user input	Ő	Ő	5	2	7
41	Layer – <i>split from 48</i>	Ŭ	0	c	-	,
42	Make components attach/detachable	11	111	21	3	146
43	Make multifunctional	0	54	15	23	92
44	Make product recyclable – split from 74					
45	Merge surfaces	0	56	0	0	56
46	Mimic natural mechanisms	14	0	1	0	15
47	Mirror or array	0	7	7	7	21
48	Nest	13	32	11	6	62
49	Offer optional components	7	25	11	2	45
50	Provide sensory feedback	7	18	11	1	37
51	Reconfigure	0	28	10	2	40

Table 2 Observations of heuristics observed across Studies 1–4, presented in alphabetical order. Seven heuristics originated from subdividing other observed heuristics

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	Design heuristic	Study 1 Product analysis	Study 2 Longterm project	Study 3 Engineer protocols	Study 4 Ind. Design protocols	Total
52	Redefine joints	24	16	0	0	40
53	Reduce material	16	9	2	0	27
54	Repeat	14	64	69	23	170
55	Repurpose packaging	6	0	0	0	6
56	Roll	0	1	6	1	8
57	Rotate	0	26	5	2	33
58	Scale up or down	0	21	16	2	39
59	Separate functions – <i>split from 77</i>					
60	Simplify	22	37	0	0	59
61	Slide	0	14	7	1	22
62	Stack	0	2	26	9	37
63	Substitute way of achieving function	0	10	28	1	39
64	Synthesize functions	13	6	4	5	28
65	Telescope	0	0	4	0	4
66	Twist	4	0	0	0	4
67	Unify	7	31	4	3	45
68	Use common base for components	0	73	1	0	74
69	Use continuous material	8	22	0	0	30
70	Use different energy source	0	0	3	1	4
71	Use human-generated power	13	0	0	0	13
72	Use multiple components in one function	0	0	27	1	28
73	Use packaging as functional component	5	0	1	0	6
74	Use repurposed or recycled materials	14	5	12	3	34
75	Utilize inner space	7	31	14	12	64
76	Utilize opposite surface	8	0	15	10	33
77	Visually distinguish functions	0	22	34	10	56
	Total heuristic instances observed	414	1749	924	370	3457
	Percentage	12%	51%	27%	11%	
	Number of new heuristics identified	39	25	5	1	
Nun	ber of existing heuristics observed	_	34	50	49	70

designer working on a long-term project. Though fewer concepts (218) were included in this study compared to the other studies, the concepts from this setting were rich in heuristic observations, with many concepts including multiple heuristics (an average of 8 heuristics per concept in Study 2, compared to

Table 3 Seven new Design Heuristics originating from subdividing seven observed heuristics

Initial heuristics coded	Revised heuristic	New heuristic added
Implement characteristics from nature within the product	Mimic natural mechanisms	Add natural features
Include user in the assembly or the customization of the product	Allow user to assemble	Allow user to customize
Flip the direction of orientation	Reconfigure	Allow user to rearrange
Create systems for returning to manufacturer after life cycle ends	Create system	Create service
Add gradations or transitions to use	Add levels	Layer
Replace materials with recycled ones	Use repurposed or recycled materials	Make product recyclable
Visually separate primary functions from secondary functions	Visually distinguish functions	Separate functions

1.5 heuristics per product in Study 1). While the product analysis uncovered 39 different heuristics, this case study of a single designer showed evidence of 57 different heuristics. This designer also used a subset of heuristics more frequently. For example, *Change direction of access* was used 211 times in these concepts, perhaps reflecting the challenge of designing universal access functions within a home bathroom. Other heuristics frequently observed in this study were *Attach independent functional components, Make components attachable/detachable*, and *Contextualize* (envision how and where the product will be used). This suggests the designer and the problem may play a role in determining which heuristics are frequently employed during idea generation.

4 Discussion

Across four empirical studies, 77 Design Heuristics were identified. These heuristics were observed in multiple concepts and studies, and across designers and design settings. These results show that examining designers' concept sets during idea generation provides a rich source of information about how they introduce variation into concepts for a given problem. In comparison, analyses of existing or award winning products (Cormier et al., 2011; Haldaman & Parkinson, 2010: Perez & Linsev, 2011: Saunders et al., 2011: Singh et al., 2009; Skiles et al., 2006; Weaver et al., 2010; Yilmaz, Seifert et al., 2016) and patents (Altshuller, 2005) provide a single design concept for each design problem as observations. These observations may limit the opportunity to observe how designers create a concept set containing multiple, varied concepts to consider. In the combined studies presented here, the methodology added the collection of observations *during* the idea generation process. Observing the generation of multiple candidate concepts appears to give rise to heuristic patterns not evident when examining only final designs. Through systematic observation of multiple concepts created by many designers in varied design problems, we can attain a deeper understanding of the role of design heuristics in idea generation.

Of course, not all designers intentionally create a large set of candidate concepts for a given design problem. With expertise, and perhaps experience regarding when specific heuristics may prove useful, a more directed process may occur, where a designer can focus more quickly on promising concepts (Cross, 2016). Certainly, there is ample evidence that designers often consider only a small set of related concepts when generating ideas (Ball et al., 1994; Chrysikou & Weisberg, 2005; Dong & Sarkar, 2011; Linsey et al., 2010; Purcell & Gero, 1996; Sio et al., 2015; Smith, 1995; Viswanathan & Linsey, 2013; Youmans & Arciszewski, 2014). This small set of concepts in idea generation may also occur when designers fixate on specific design features (Jansson & Smith, 1991; Purcell & Gero, 1996; Sio et al., 2015). Logically, the idea generation process benefits from considering as many different concepts as possible (Akin & Lin, 1995; Atman et al., 1999; Brophy, 2001; Liu

et al., 2003) in order to cover the space of possible designs (Gero, 1990; Goel & Pirolli, 1992; MacLean et al., 1991; Simon, 1981). To do so, the evidence from the combined studies here suggests the use of design heuristics.

One open issue regarding design heuristic use is how to decide which heuristic to apply in any given design context. The data from existing design solutions collected in these studies suggests the heuristics are readily applicable across design problems. Other approaches, such as Design to Connect (Bleuze, Cioccib, Detandb, & De Baetsc, 2014), have tested whether organized cues for heuristic use are helpful. Their study found that including a set of 'design drivers' (e.g., usability, aesthetics, economy) did not improve performance of designers; instead, the student designers in their studies preferred an unstructured use of their connection guidelines. In studies with Design Heuristics, providing a subset of heuristics to designers to be selected at random has produced improved design outcomes (Daly, Christian et al., 2012; Daly, Yilmaz et al., 2012). In the open-ended idea generation process, less determinate methods like Design Heuristics may be preferable for creating alternative design concepts in the early phases of conceptual design.

Another question is whether the set of 77 Design Heuristics represent a definitive description, or whether more such heuristics may be uncovered in future research. In the present study, we analyzed concepts from 400 consumer products, 218 designs by a professional industrial designer, and 247 concepts from 48 different designers. This represents a large sample of design solutions across many different types of products and designers. Across these studies, the identification of new heuristics slowed, so that it appeared the readily evident heuristics had been uncovered, with only one new heuristic observed in the last study. However, further research on identifying new heuristics may identify new heuristics when different design problems are included, or when different designers' work is sampled. Because heuristics are based upon experiences, new design goals and contexts may give rise to innovation in heuristics as the field of product design (and designers' experiences) changes dynamically over time. In addition, the organization of these 77 Design Heuristics may be refined under further research (Design Heuristics, 2012). Finally, the empirical data described here was specific to the domain of product design. Future research should examine other domains, such as service design, software programs, and chemical engineering, to determine how heuristics may differ by domain.

What is the 'right' level of heuristic definition? Is it best to have few heuristics that capture more abstract similarities across designs, such as only three principles (*expand/collapse, expose/cover*, and *fuse/divide*) identified in transforming products (Singh et al., 2009)? Having a few, more general heuristics makes learning and remembering them easier, but requires more effort in deciding how to apply them within a new design problem. Alternatively, having more heuristics and conditions on their application, such as the 40 TRIZ principles

and contradiction matrix (Altshuller, 1997, 2005), may be easier to apply to specific problems. However, a system with more heuristics may be harder to learn and remember, and likely requires more training (Ilevbare et al., 2013).

Goel and Bhatta (2004) describe this issue of 'granularity' (Fu et al., 2015) as the problem of specifying generic relations (independent of any specific design situation) among abstract design elements. The specificity of an identified heuristic can be characterized at varied levels, from 'very general' (abstracted away from observed examples) to 'very specific' (closely tied to the observed example). At the extreme, a complete example, as in case-based design (Kolodner, 1993, 1997) and analogical approaches (Ball, Ormerod, & Morley, 2004; Bonnardel, 2000; Casakin, 2004; Christensen & Schunn, 2007; Helms et al., 2009; Linsey, 2007; Linsey et al., 2012; Perkins, 1997; Qian & Gero, 1996; Visser, 1996) provides specific information about implementation. However, application to new design problems requires the abstraction of heuristics with each use, costly in cognitive effort. Case approaches also raise the problem of access, or finding relevant analogies given the present design problem. This suggests a trade-off between heuristic specificity (that aids application) and generality (that increases relevance) that has consequences for the access and ease of heuristic application (Gray et al., 2016).

In the extraction of Design Heuristics, we propose a criterion of efficacy for heuristics: The success of heuristic definitions can be assessed based on their effectiveness in helping other designers create novel designs through their application during idea generation. Further research would then determine whether a candidate set of design heuristics captures design variations at a level useful in concept generation. The 77 Design Heuristics presented here offer an intermediate level of description that facilitates implementing the heuristic in a new problem context. The needed information about how to create a new concept is readily available within the heuristic. Yet, many decisions must still be made about how to apply the heuristic in a given problem. This includes the possibility of reapplying the same heuristic to the same problem again to create a different concept, as observed in Yilmaz and Seifert (2011). The challenges of organizing many heuristics during idea generation can be managed through an external representation of each heuristic and random selection among heuristics; then, if more concepts are desired, more heuristics can be considered. It is possible that further research might identify cues that indicate when specific heuristics are most relevant for application in a problem. Whether it is better to have 10 principles, or 77, or 1000, depends on what designers find helpful to their idea generation process.

In future research, it is important to compare the 77 Design Heuristics to other proposed methods of idea generation in order to assess its efficacy. Increasingly, studies are showing the advantages of specific idea generation methods, and suggesting which methods are more effective in given design circumstances

(Hernandez et al., 2013; Jensen, 2012; Jensen, Weaver, Wood, Linsey, & Wood, 2009; Ogot & Okudan, 2007; White, Wood, & Jensen, 2012). Empirical studies can identify which approaches work well for specific types of design problems, design domains, and types of designers. In addition, it is important to establish the value of generating multiple candidate concepts for later selection and implementation. The present findings provide evidence for a new tool to aid designers in the process of idea generation. In the past, the use of heuristics in idea generation likely depended solely upon the generalizations each designer was able to build from their own design experiences. The use of a shared, external tool like the 77 Design Heuristics may facilitate the creation of innovative concepts by even novice designers in the early stages of conceptual design.

5 Conclusion

Design heuristics offer a conceptual bridge between more general design theories and individual design precedents often provided to learners. The empirical observations presented here combine data from four studies of many designers working on a wide variety of products and problems in order to identify common patterns evident in their designs. The resulting identification of 77 Design Heuristics provides a collection of strategies grounded in observed use in concepts, and demonstrated across design problems, multiple concepts, and designers. This empirical approach to defining heuristic strategies is unique among the approaches in the field because it includes protocols from designers where more than one concept is sampled. By examining the candidate designs generated in addition to complete designs in the form of products and patents, rich information about how designers successfully create alternative concepts becomes evident. The results provide a collection of Design Heuristics suitable for use as a tool to explore possible alternative concepts. Design Heuristics may enhance the idea generation process by providing multiple strategies to consider, increasing the likelihood of innovative solutions.

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References

- Akin, O., & Lin, C. (1995). Design protocol data and novel design decisions. Design Studies, 16, 211–236.
- Alexander, C., Ishikawa, S. J. A., & Silverstein, M. J. A. (1977). A Pattern Language: Towns, Buildings, Construction. New York, NY: Oxford University Press.
- Altshuller, G. (1984). *Creativity as an Exact Science*. New York, NY: Gordon and Breach.
- Altshuller, G. (1997). 40 Principles: TRIZ Keys to Technical Innovation. Worcester, MA: Technical Innovation Center, Inc.

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- Altshuller, G. (2005). 40 Principles: TRIZ Keys to Technical Innovation. extended edition, Worchester, MA: Technical Innovation Center, Inc.
- Altshuller, G., & Rodman, S. (1999). The Innovation Algorithm: TRIZ, Systematic Innovation and Technical Creativity. Worchester, MA: Technical Innovation Center.
- Atman, C. J., Chimka, J. R., Bursic, K. M., & Nachtman, H. L. (1999). A comparison of freshman and senior engineering design process. *Design Studies*, 20(2), 131–152.
- Baker, L. M., & Dunbar, K. (2000). Experimental design heuristics for scientific discovery: the use of "baseline" and "known standard" controls. *International Journal of Human-computer Studies*, 53(3), 335–349.
- Ball, L. J., Evans, J., & Dennis, I. (1994). Cognitive processes in engineering design: a longitudinal study. *Ergonomics*, 37(11), 1753–1786.
- Ball, L. J., Ormerod, T. C., & Morley, N. J. (2004). Spontaneous analogising in engineering design: a comparative analysis of experts and novices. *Design Studies*, 25(5), 495–508.
- Bleuze, T., Cioccib, M.-C., Detandb, J., & De Baetsc, P. (2014). Engineering meets creativity: a study on a creative tool to design new connections. *International Journal of Design Creativity and Innovation*, 2(4), 203–223.
- Bonnardel, N. (2000). Towards understanding and supporting creativity in design: analogies in a constrained cognitive environment. *Knowledge-based Systems*, 13, 505–513.
- Brophy, D. R. (2001). Comparing the attributes, activities, and performance of divergent, convergent, and combination thinkers. *Creativity Research Journal*, 13(3&4), 439–455.
- Casakin, H. (2004). Visual analogy as a cognitive strategy in the design process: expert versus novice performance. *Journal of Design Research*, 4(2).
- Christensen, B. T., & Schunn, C. D. (2007). The relationship of analogical distance to analogical function and preinventive structure: the case of engineering design. *Memory & Cognition*, 35(1), 29–38.
- Chrysikou, E. G., & Weisberg, R. W. (2005). Following the wrong footsteps: fixation effects of pictorial examples in a design problem-solving task. *Journal of Experimental Psychology: Learning, Memory & Cognition, 31*(5), 1134–1148.
- Chulvi, V., Gonzalez-Cruz, M. C., Mulet, E., & Aguilar-Zambrano, J. (2013). Influence of the type of idea-generation method on the creativity of solutions. *Research in Engineering Design*, 24(1), 33–41.
- Clancey, W. J. (1985). Heuristic classification. Artificial Intelligence, 27(3), 289-350.
- Clapham, M. M. (1997). Ideational skills training: a key element in creativity training programs. *Creativity Research Journal*, 10(1), 33-44.
- Cormier, P., Literman, B., & Lewis, K. (2011, August 28–31). Empirically Derived Heuristics to Assist Designers with Satisfying Consumer Variation in Product Design. Paper presented at the ASME International Design Engineering Technical Conferences and Computers and Information in Engineering Conference, Washington, DC.
- Cross, N. (1994). Engineering Design Methods: Strategies for Product Design. Chichester, UK: John Wiley & Sons.
- Cross, N. (2011). Design Thinking. London: Bloomsbury.
- Cross, N. (2016). Personal communication.
- Daly, S. R., Christian, J. L., Yilmaz, S., Seifert, C. M., & Gonzalez, R. (2012). Assessing design heuristics for idea generation in an introductory engineering course. *International Journal of Engineering Education*, 28(2), 1–11.

- Daly, S. R., Yilmaz, S., Christian, J. L., Seifert, C. M., & Gonzalez, R. (2012). Design heuristics in engineering concept generation. *Journal of Engineering Education*, 101(4), 601–629.
- Design Heuristics. (2012). Retrieved from www.designheuristics.com.
- Dong, A., & Sarkar, S. (2011). Unfixing design fixation: from cause to computer simulation. Journal of Creative Behavior, 45(2), 147–159.
- Dorst, K., & Cross, N. (2001). Creativity in the design process: co-evolution of problem-solution. *Design Studies*, 22(5), 425-437.
- Ericsson, K. A., & Simon, H. A. (1993). Protocol Analysis: Verbal Reports as Data. Cambridge, MA: The MIT Press.
- Fu, K. K., Yang, M. C., & Wood, K. L. (2015). Design Principles: The Foundation of Design. Paper presented at the ASME International Design Engineering Technical Conferences, Boston, MA.
- Gero, J. S. (1990). Design prototypes: a knowledge representation schema for design. *AI Magazine*, 11(3), 26-36.
- Goel, A. K., & Bhatta, S. R. (2004). Use of design patterns in analogy-based design. *Advanced Engineering Informatics*, 18(2), 85–94.
- Goel, V., & Pirolli, P. (1992). The structure of design problem spaces. Cognitive Science, 16(3), 395–429.
- Goldstein, D. G., Gigerenzer, G., Hogarth, R. M., Kacelnik, A., Kareev, Y., Klein, G., et al. (2001). Why and how do simple heuristics work? In G. Gigerenzer, & R. Selten (Eds.), *Bounded Rationality: The Adaptive Toolbox* (pp. 173–190) Cambridge: MIT Press.
- Gray, C. M., Seifert, C. M., Yilmaz, S., Daly, S. R., & Gonzalez, R. (2016). What is the content of "Design Thinking"? Design Heuristics as conceptual repertoire. *International Journal of Engineering Education*, 32(3B), 1349–1355.
- Haldaman, J., & Parkinson, M. B. (2010). Reconfigurable Products and their Means of Reconfiguration. Paper presented at the ASME INternational Design Engineering Technical Conferences and Computers and Information in Engineering Conference, Las Vegas, Nevada.
- Helms, M., Vattam, S. S., & Goel, A. K. (2009). Biologically inspired design: process and products. *Design Studies*, 30(5), 606–622.
- Hernandez, N. V., Schmidt, L. C., Kremer, G. O., & Lin, C.-Y. (2014). An empirical study of the effectiveness of selected cognitive aids on multiple design tasks. In J. S. Gero (Ed.), *Design Computing and Cognition '12* (pp. 227–246). Netherlands: Springer.
- Hernandez, N. V., Schmidt, L. C., & Okudan, G. E. (2013). Systematic ideation effectiveness study of TRIZ. *Journal of Mechanical Design*, 135(10), 101009.
- Ilevbare, I. M., Probert, D., & Phaal, R. (2013). A review of TRIZ, and its benefits and challenges in practice. *Technovation*, *33*(2), 30–37.
- Jansson, D. G., & Smith, S. M. (1991). Design fixation. *Design Studies*, 12(1), 3–11.
- Jensen, D. D. (2012, June 10–13). AC 2012-3797: Evaluating Ideation using the Publications Popular Science, Popular Mechanics and Make in Coordination with a New Patent Search Tool and the 6-3-5 Method. Paper presented at the American Society of Engineering Education Annual Meeting (ASEE), San Antonio, TX.
- Jensen, D. J., Weaver, J., Wood, K. L., Linsey, J. S., & Wood, J. (2009). Techniques to Enhance Concept Generation and Develop Creativity. Paper presented at the American Society of Engineering Education Annual Meeting, Austin, TX.

- Jin, Y., & Benami, O. (2010). Creative patterns and stimulation in conceptual design. Artificial Intelligence for Engineering Design, Analysis and Manufacturing, 24(02), 191–209.
- Klein, G. (1993). A recognition primed decision (RPD) model of rapid decision making. In J. O. G. Klein, R. Calderwood, & C. E. Zsambok (Eds.), *Decision Making in Action: Models and Methods* (pp. 205–218). Cambridge, MA: MIT Press.
- Kolodner, J. (1993). Case-based Reasoning. San Francisco, CA: Morgan Kaufmann Publishers Inc.
- Kolodner, J. L. (1997). Educational implications of analogy: a view from casebased reasoning. *American Psychologist*, 52(1), 57.
- Krippendorff, K. (2005). *The Semantic Turn: A New Foundation for Design*. Boca Raton, FL: CRC Press.
- Lawson, B. R. (1979). Cognitive strategies in architectural design. *Ergonomics*, 22(1), 59-68.
- Lawson, B. (1980). How Designers Think: The Design Process Demystified. London, UK: Architectural.
- Lawson, B. (2012). What Designers Know. London, UK: Routledge.
- Linsey, J. S. (2007). Design-by-analogy and Representation in Innovative Engineering Concept Generation. (PhD Dissertation), Austin, Texas: University of Texas.
- Linsey, J. S., Markman, A. B., & Wood, K. L. (2012). Design by analogy: a study of the WordTree Method for problem re-representation. *Journal of Mechanical Design*, 134(4), 041009.
- Linsey, J. S., Tseng, I., Fu, K., Cagan, J., Wood, K. L., & Schunn, C. D. (2010). A study of design fixation, its mitigation and perception in engineering design faculty. *Journal of Mechanical Design*, 132(4), 1–12.
- Liu, Y. C., Bligh, T., & Chakrabarti, A. (2003). Towards an 'ideal' approach for concept generation. *Design Studies*, 24(4), 341–355.
- MacLean, A., Young, R. M., Bellotti, V. M. E., & Moran, T. P. (1991). Questions, options, and criteria: elements of design space analysis. *Human-computer Interaction*, 6(3–4), 201–250.
- Matthews, P., Wallace, K., & Blessing, L. (2000). Design heuristics extraction. Artificial Intelligence in Design'00. Netherlands: Springer 435–453.
- Nisbett, R. E., & Ross, L. (1980). Human Inference: Strategies, and Shortcomings of Social Judgment. Englewood Cliffs, NJ: Prentice-Hall.
- Ogot, M., & Okudan, G. E. (2007). Systematic creativity methods in engineering education: a learning styles perspective. *International Journal of Engineering Education*, 22(3), 566–576.
- Orloff, M. A. (2003). *Inventive Thinking through TRIZ: A Practical Guide*. Berlin, Germany: Springer.
- Perez, A., Linsey, J. S., Tsenn, J., & Glier, M. (2011, August 28–31). Identifying Product Scaling Principles: A Step towards Enhancing Biomimetic Design. Paper presented at the ASME 2011 International Mechanical Engineering Congress and Exposition, Washington, DC.
- Perkins, D. (1997). Creativity's camel: the role of analogy in invention. In T. Ward, S. Smith, & J. Vaid (Eds.), *Creative Thought* (pp. 523–528). Washington, DC: American Psychological Association.
- Purcell, A. T., & Gero, J. S. (1996). Design and other types of fixation. *Design Studies*, 17(4), 363–383.
- Qian, L., & Gero, J. S. (1996). Function-behavior-structure paths and their role in analogy-based design. Artificial Intelligence for Engineering, Design, Analysis and Manufacturing, 10(04), 289-312.

- Riel, A. J. (1996). Object-oriented Design Heuristics. Reading, Massachusetts: Addison-Wesley Professional.
- Saunders, M. N., Seepersad, C. C., & Hölttä-Otto, K. (2011). The characteristics of innovative mechanical products. *Journal of Mechanical Design*, *133*. 021009–021001-021009.
- Savransky, S. D. (2000). Engineering of Creativity: Introduction to TRIZ Methodology of Inventive Problem Solving. Boca Raton: CRC Press.
- Shah, J. J., Vargas-Hernandez, N., & Smith, S. M. (2003). Metrics for measuring ideation effectiveness. *Design Studies*, 24, 111–134.
- Simon, H. A. (1981). *The Sciences of the Artificial* (2nd ed.). Cambridge, MA: MIT Press.
- Singh, V., Skiles, S. M., Krager, J. E., Wood, K., Jensen, D., & Sierakowski, R. (2009). Innovations in design through transformation: a fundamental study of transformation principles. *Journal of Mechanical Design*, 131(8). 081010-081011-081018.
- Singh, V., Walther, B., Krager, J. E., Putnam, N., Koraishy, B., Wood, K. L., et al. (2007, September 4–7). *Design for Transformation: Theory, Method and Application*. Paper presented at the ASME 2007 International Design Engineering Technical Conferences and Computers and Information in Engineering Conference, Las Vegas, Nevada.
- Sio, U. N., Kotovsky, K., & Cagan, J. (2015). Fixation or Inspiration? A Metaanalytic Review of the Role of Examples on Design Processes. *Design Studies*, 39(C), 70–99.
- Skiles, S. M., Singh, V., Krager, J. E., Seepersad, C. C., Wood, K. L., & Jensen, D. (2006, September 10–13). Adapted Concept Generation and Computational Techniques for the Application of a Transformer Design Theory. Paper presented at the ASME 2006 International Design Engineering Technical Conferences and Computers and Information in Engineering Conference, Philedelphia, PA.
- Smith, S. M. (1995). Getting into and out of mental ruts: a theory of fixation, incubation, and insight. In R. J. Sternberg, & J. E. Davidson (Eds.), *The Nature* of *Insight*. Cambridge: MIT Press.
- Smith, G. F. (1998). Idea-generation techniques: a formulary of active ingredients. *The Journal of Creative Behavior*, *32*(2), 107–134.
- Smith, E. M. (2003). From Russia with TRIZ. Mechanical Engineering, 125(3), D18.
- Visser, W. (1996). Two functions of analogical reasoning in design: a cognitivepsychology approach. *Design Studies*, 17(4), 417–434.
- Viswanathan, V. K., & Linsey, J. S. (2013). Design fixation and its mitigation. Journal of Mechanical Design, 135(5), 051008.
- Weaver, J. M., Kuhr, R., Wang, D., Crawford, R. H., Wood, K. L., Jensen, D., et al. (2009, August 30–September 2). *Increasing Innovation in Multi-function Systems: Evaluation and Experimentation of Two Ideation Methods for Design*. Paper presented at the ASME 2009 International Design Engineering Technical Conferences and Computers and Information in Engineering Conference, San Diego, CA.
- Weaver, J., Wood, K., Crawford, R. H., & Jensen, D. (2010). Transformation design theory: a meta-analogical framework. *Journal of Computing and Information Science in Engineering*, 10(3), 031012.
- Weaver, J. M., Wood, K. L., & Jensen, D. (2008). Transformation Facilitators: A Quantitative Analysis of Reconfigurable Products and their Characteristics. Paper presented at the ASME International Design Engineering Technical Conferences and Computers and Information in Engineering Conference, Brooklyn, NY.

Design heuristics for idea generation

- White, C., Wood, K., & Jensen, D. (2012). From brainstorming to C-sketch to principles of historical innovators: ideation techniques to enhance student creativity. *Journal of STEM Education*, *13*(5), 12–25.
- Yilmaz, S., Daly, S. R., Seifert, C. M., & Gonzalez, R. (2015). How do designers generate new ideas? Design Heuristics across two disciplines. *Design Science*, 1, 1–29.
- Yilmaz, S., & Seifert, C. M. (2011). Creativity through design heuristics: a case study of expert product design. *Design Studies*, 32(4), 384-415.
- Yilmaz, S., Seifert, C. M., Daly, S. R., & Gonzalez, R. (2016). Design Heuristics in innovative products. *Journal of Mechanical Design*, 138(7). http:// dx.doi.org/10.1115/1.4032219. 071102-071102-12.
- Youmans, R. J., & Arciszewski, T. (2014). Design fixation: classifications and modern methods of prevention. Artificial Intelligence for Engineering Design, Analysis and Manufacturing, 28(02), 129–137.