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## Using Interactive Nutrition Modules to Increase Critical Thinking Skills in College Courses

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**USING INTERACTIVE NUTRITION MODULES TO INCREASE CRITICAL THINKING SKILLS IN COLLEGE COURSES**  
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<b>Abstract:</b>	<p><b>Objective:</b> To understand how the addition of an evidence-based framework to an online nutrition module influences college students' critical thinking decision making (CT-DM).</p> <p><b>Design:</b> Students were individually randomized into an intervention group or a control group. The nutrition modules focused on two topics related to different types of eating behavior. Students completed a CT-DM activity to generate a score.</p> <p><b>Participants:</b> College students, between 18-24 years old, recruited from introductory nutrition and agriculture science courses at two universities.</p> <p><b>Intervention:</b> Intervention and control received two nutrition modules. The intervention added a CT-DM framework that framed the topic as a problem, incorporated activities, and provided scaffolding.</p> <p><b>Main Outcome Measure(s):</b> CT-DM was scored using a validated rubric to assess the use of critical thinking skills when making a food-related decision. Green eating and critical thinking disposition were measured.</p> <p><b>Analysis:</b> Hierarchical linear regression and t-tests were used to assess outcomes.</p> <p><b>Results:</b> 431 students participated (intervention=203; control=228). After controlling for university, the intervention group scored significantly higher on CT-DM (18.1±7.6) compared to the control (15.4±8.4); F (3,428) =14.58, p&lt;.001.</p> <p><b>Conclusions and Implications:</b> The results show that an evidence-based framework using nutrition topics encourages CT-DM skills. Future nutrition higher-education interventions should use frameworks to enhance student learning.</p>

**Cover letter:**

This manuscript is being submitted as a *Research Article*. The manuscript was written specifically for the Scholarship of Learning special issue. This research focuses on how evidence-based instructional practices can be used to encourage critical decision making skills in introductory nutrition and agriculture courses. The manuscript has not been and will not be submitted elsewhere for publication. The evidence-based instructional practices that were operationalized and used in this research were described and evaluated in a separate manuscript that is currently under review, however, no overlapping outcomes between the manuscripts were used. All authors have reviewed and approved the complete manuscript including tables and the figure. The complete page count is 18 pages and 5,212 words.

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## TITLE PAGE

### USING INTERACTIVE NUTRITION MODULES TO INCREASE CRITICAL THINKING SKILLS IN COLLEGE COURSES

#### *Research Article*

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## **Abstract**

**Objective:** To understand how the addition of an evidence-based framework to an online nutrition module influences college students' critical thinking decision making (CT-DM).

**Design:** Students were individually randomized into an intervention group or a control group. The nutrition modules focused on two topics related to different types of eating behavior. Students completed a CT-DM activity to generate a score.

**Participants:** College students, between 18-24 years old, recruited from introductory nutrition and agriculture science courses at two universities.

**Intervention:** Intervention and control received two nutrition modules. The intervention added a CT-DM framework that framed the topic as a problem, incorporated activities, and provided scaffolding.

**Main Outcome Measure(s):** CT-DM was scored using a validated rubric to assess the use of critical thinking skills when making a food-related decision. Green eating and critical thinking disposition were measured.

**Analysis:** Hierarchical linear regression and t-tests were used to assess outcomes.

**Results:** 431 students participated (intervention=203; control=228). After controlling for university, the intervention group scored significantly higher on CT-DM ( $18.1 \pm 7.6$ ) compared to the control ( $15.4 \pm 8.4$ );  $F(3,428) = 14.58, p < .001$ .

**Conclusions and Implications:** The results show that an evidence-based framework using nutrition topics encourages CT-DM skills. Future nutrition higher-education interventions should use frameworks to enhance student learning.

(Word Count: 200)

**Key Words:** Critical Thinking, Decision Making, STEM Education, Problem-Based Learning

# 1 USING INTERACTIVE NUTRITION MODULES TO INCREASE CRITICAL 2 THINKING SKILLS IN COLLEGE COURSES

## 3 4 5 INTRODUCTION 6 7

8 Twenty first century skills that include critical thinking, have been identified by  
9 employers of college graduates as more important than academic success in making hiring  
10 decisions.<sup>1-3</sup> This is, in part, related to the increasing pace of scientific discovery and advances in  
11 technology that requires critical thinking skills. There is evidence that individuals with high level  
12 critical thinking skills make better decisions, such as taking less unnecessary risks, than those  
13 with less developed critical thinking skills.<sup>1,4</sup> However, teaching critical thinking skills in  
14 introductory college courses is challenging, particularly in science, technology, engineering and  
15 math (STEM) disciplines such as nutrition and animal sciences where there is a high focus on  
16 memorizing critical information.<sup>5,6</sup> There also have been few research studies focusing on the  
17 science of teaching and learning addressing critical thinking within STEM courses.

18 Critical thinking skills include the ability to apply standards, seek out information,  
19 problem solve, transform knowledge, predict consequences of decisions, be creative, practice  
20 logical reasoning, and evaluate evidence when faced with a problem or question.<sup>7-9</sup> Because  
21 critical thinking is such a broad construct and includes so many skills, it is difficult to measure  
22 for educational outcomes.<sup>10</sup> Narrowing the construct of critical thinking to critical thinking  
23 decision making (CT-DM), defined by having skills in problem solving, logical reasoning, and  
24 evaluating evidence when making decisions<sup>9</sup>, allows for realistic measurement methods and  
25 facilitates curriculum development.

26 A learner-centered curriculum fosters CT-DM by allowing students to connect thoughts  
27 and ideas through reflection of what they already know, investigation of new knowledge, and  
28 explicit skill development.<sup>11, 12</sup> Learner-centered curricula can be operationalized through the  
29 theory of social constructivism using problem-based learning (PBL) activities.<sup>13, 14</sup> Social  
30 constructivism theory postulates that students learn by connecting concepts to previous  
31 knowledge and experiences<sup>15, 16</sup>, while PBL provides students with opportunities to assess  
32 complex problems using a variety of resources, and develop their own strategies for addressing  
33 these problems.<sup>17, 18</sup> However, incorporating PBL into introductory STEM classes is difficult due  
34 to discipline and accreditation demands for course content. One solution to overcome these time  
35 constraints has been to utilize out-of-class, online PBL activities to enhance skill development  
36 and motivate students to engage in learning activities. Additionally, students have expressed  
37 frustration with PBL activities, particularly in introductory courses, because of a lack of  
38 framework to successfully complete the tasks,<sup>19</sup> which underscores the importance of  
39 intentionally designing activities and using topics that motivate students to engage in learning.

40 In addition to pedagogical strategies, personal factors such as attitude towards critical  
41 thinking or interest in the PBL topic may affect student engagement with the curriculum.<sup>20, 21</sup>  
42 Some students are naturally more open to using critical thinking skills, which has been assessed  
43 in previous research by the critical thinking disposition scale (CTD)<sup>22</sup>, which may influence the  
44 outcome of a CT-DM intervention. Also, researchers have shown that personal interest in a topic  
45 motivates students to participate, express their point of view and engage in a learning activity.<sup>23</sup>  
46 For example, nutrition and food choice topics are ideal for motivating students to engage in  
47 learning because the majority can connect with those scenarios as they are making eating and  
48 food choice decisions daily. One way of assessing personal interest in a topic is by measuring

49 self-reported behaviors pertaining to the topic. Both variables, CTD and personal interest in a  
50 topic, need to be measured to evaluate the role they play in explaining CT-DM.

51 Thus, the primary objective was to determine if the addition of a contextual framework,  
52 defined in this study as the critical thinking decision making framework (CTDM-F), to a two-  
53 module, online PBL curriculum influences college students' CT-DM skills. The second objective  
54 was to explore the relationship between the mediating factors of CTD and personal interest in a  
55 topic on CT-DM skills. The primary hypothesis was that undergraduate students exposed to the  
56 CTDM-F would have a significantly higher CT-DM score when compared to the students in the  
57 control group and that CTD and interest in the topic would be mediators of CT-DM score.

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## METHODS

### 62 **Research Design**

63 This study was a randomized controlled trial, where students from two geographically  
64 diverse universities (University of Rhode Island and University of North Texas) were recruited  
65 from introductory level nutrition and animal science courses, and individually randomized into to  
66 a CTDM-F group or a control group. The study was approved at both universities by their  
67 respective Institutional Review Boards. Students were provided with a link to access the online  
68 consent to participate in the research study and sign-up for the modules. Students were then  
69 randomized into either the CTDM-F group or control group by the computer. Students were then  
70 directed to an online pretest, assessing key mediating variables such as inclination to use critical  
71 thinking (measured by CTD), prior interest in the topic of sustainable eating (measured by green  
72 eating behavior (GE)) and demographic data. They were then immediately directed to the first  
73 module: pros and cons of animal protein vs. non-animal protein food choices. One week later,



74 they were notified that the second module was available: pros and cons of organic food vs. non-  
75 organic food choices. The post-test immediately followed module 2, which was similar to the  
76 pretest.

77 The topics for the modules, animal vs. plant-based sources of protein and organic vs.  
78 non-organic food, were chosen because both topics can be argued from multiple perspectives and  
79 relate to authentic scenarios that college students may face when making food purchasing or  
80 eating decisions. As higher education institutions move towards providing greater education on  
81 sustainable practices and sustainably produced food choices<sup>24</sup>, college students will be faced  
82 with these decision making scenarios. Previous research also supports that college students are  
83 interested in topics related to environmentally conscious eating and find these topics motivating  
84 for learning.<sup>25</sup>

#### 85 **CT-DM Module and Control Module**

86 Both modules were easily accessible to students from various platforms via internet  
87 connections and took about 15 minutes to complete. Module 1 began with an interactive “quiz”  
88 to determine what type of learner the student was, a video discussing the importance of critical  
89 thinking, and two videos addressing both sides of the specific topic area (animal protein vs. non-  
90 animal protein foods). After watching the topic video, the student was asked to make a decision  
91 about which side they agreed with and then were prompted to write a brief response explaining  
92 why they made that decision. Module 2 had the same format as module 1 and then was followed  
93 by the post-test.

#### 94 **Critical Thinking Decision-Making Framework**

95 The CTDM-F was designed using the social constructivism model with the aim to bring  
96 about a conceptual change in student thinking by having the student construct their own

97 conclusions when presented with information.<sup>15, 16</sup> Scaffolding, or structuring the learning  
98 process to help the learner more towards independence, was also a key component in the CTDM-  
99 F.<sup>15, 16</sup> The three specific strategies to operationalize an online delivery of social constructivism  
100 with an emphasis on scaffolding, were as follows: 1) topic was introduced as an authentic  
101 problem<sup>26</sup>, 2) an input scaffold in the form of a t-chart was provided to help organize information  
102 from the two sources with alternative views<sup>27</sup>, and 3) an output scaffold was provided to help  
103 frame the argument separate from decision making<sup>28</sup>, in the form of a “mind-map”. The “mind-  
104 map” was an activity where students structured their response with their decision and reasoning  
105 using a drag-and-drop activity. For module 1, the CTDM-F provided more scaffolding for the  
106 decision-making activity by using a closed exercise format with fill-in-the-blanks structured  
107 responses. For module 2, the scaffolding in the decision-making activity was removed by  
108 providing only a blank text box for recording the decision-making response.

109 **Control Module.** The control group was exposed to the same videos but did not have the topic  
110 introduced as a problem, did not receive the input scaffold and did not have an output scaffold.  
111 They were provided with a blank decision-making text box in both modules. The differences in  
112 the layout of module 1 and module 2 between groups is described in Table 1.

113 [TABLE 1]

#### 114 **Measurements**

115 **Primary Outcome: Critical Thinking Decision-Making Score.** To calculate the CT-DM score,  
116 a previously developed and tested rubric<sup>29</sup> was used to score the decision-making activity at the  
117 end of module 2, when both groups received the text box. Scores ranged from 0-30, evaluating  
118 the extent of CT-DM, with 0 representing a non-response/failed to provide a text response  
119 addressing the task, and 30 indicating a text response that addressed the three following

120 constructs: 1) ability to make a decision (0/10 points), 2) evidence to support the decision (0-15  
121 points), and 3) ability to see the other side's point of view (0/5 points). The responses were  
122 scored using a computer-assisted scoring system that guided trained researchers through the  
123 response criteria generating a total score. The passing rate for training was set at an interrater  
124 reliability (IRR) score of  $\geq .80$ , which has been used in previous research as an acceptable  
125 agreement score.<sup>30</sup> All responses were duplicate scored by research assistants (n=4) who  
126 participated in training and were evaluated for reliability based on matching rate. Scores that did  
127 not match (18%) were then scored by a senior researcher to determine a final score.

128 **Critical Thinking Disposition.** Disposition to use critical thinking operationalized as CTD was  
129 measured using a 5-item subscale from the Motivated Strategies for Learning Questionnaire  
130 (MSLQ).<sup>22</sup> The subscale was designed and validated to measure college students' thinking  
131 strategies to apply knowledge and critically evaluate situations and found to have good internal  
132 reliability with a Cronbach's alpha ( $\alpha$ ) value of .72.<sup>31</sup> The composite score is an indicator of  
133 students' inclination to use CT during a decision-making process.<sup>22</sup> The five items were  
134 measured using a 5-point Likert scale, ranging from (1) strongly disagree (5) strongly agree.

135 **Green Eating Behavior.** Prior interest in the topic, operationalized as GE behavior, was  
136 measured using a validated 7-item survey ( $\alpha = .81$ ), which assessed the frequency of choosing  
137 sustainably produced food.<sup>24</sup> The items included behaviors related to purchasing foods locally  
138 grown, shopping at farmer's markets, buying organic, and purchasing free-range animal proteins.  
139 Items were measured on a 5-point Likert scale ranging from (1) barely ever to never to (5)  
140 almost always, a higher score indicating a greater frequency of choosing sustainably produced  
141 foods.

142 **Data Analysis**

143 Data were analyzed using SPSS 24 (IBM, Armonk, NY, 2016). Descriptive variables  
144 were analyzed for normal distribution using values of -2 to +2 for skewness and kurtosis.<sup>32</sup>  
145 Demographic variables were analyzed using means ( $\pm$  standard deviation) and frequency (%).  
146 Independent t-tests were used to analyze baseline differences between the CTDM-F and the  
147 control groups, and between universities. Analysis of covariance (ANCOVA) was conducted at  
148 baseline to control for significant differences between groups. A Pearson's Chi-square test of  
149 independence analysis was used to analyze categorical variables. To assess differences in CT-  
150 DM score between groups, a two-stage hierarchical multiple regression was performed with CT-  
151 DM score as the dependent variable. University was entered at stage one of the regression to  
152 control for differences at baseline. Group (control vs CTDM-F) was entered at stage two to  
153 assess differences between groups on CT-DM score after controlling for university. ANCOVA  
154 was also used to assess differences in scores and components of the CDM score between groups.  
155 To evaluate change in CTD score and GE behavior from baseline to post-intervention, repeated  
156 measures ANCOVA was conducted. For all the analyses significance level was set at  $p < 0.05$ . An  
157 exploratory structural equation model (SEM) using path analysis was also performed to explore  
158 the amount of variation that was accounted for in CT-DM score by group, CTD, and GE  
159 behavior. For the SEM outcomes, ideal macro-level fit indices parameters include:  $\chi^2 p > .05$ ,  
160  $\chi^2/df$  ratio  $< 4$ , CFI  $> .90$ , RMSEA  $< .08$ , and for  $\chi^2$  difference, a larger value equates to better  
161 fit.<sup>32</sup>

162

163

## RESULTS

164

165 **Sample**

166 A total of 440 students randomly assigned to either the control (n=230) or the CTDM-F  
167 group (n=210). Participants were excluded if they did not complete baseline demographic questions  
168 and if they did not complete both module 1 and module 2 (n=9). Students' mean age was 19.4 ( $\pm$ 1.4)  
169 years old and 73.8% were female. All descriptive variables were normally distributed. Reported  
170 major was grouped into three categories: 1) Arts and Humanities: social sciences, arts, and  
171 undecided (47.5%); 2) STEM: science, technology, engineering, agriculture, and math (22.2%);  
172 and 3) STEM-Health: nutrition, kinesiology, nursing, pre-med (30.3%) to examine differences.

173 Significant baseline differences between universities were found for all categorical and  
174 descriptive variables. Based on these findings, university was controlled for in the analyses using  
175 ANCOVA. Baseline comparisons between groups, after controlling for university, are shown in  
176 Table 2. At baseline, the control group initially had a significantly higher pre-CTD score ( $p < .05$ ),  
177 but after controlling for university, the difference was no longer significant. No other differences  
178 between the CTDM-F group and control group were found.

179  
180 [TABLE 2]  
181

## 182 **CT-DM Score**

183 The hierarchical multiple regression revealed that at step 1, university contributed  
184 significantly to the regression model,  $F(2,428) = 19.11, p < .001$  and accounted for 8.0% of the  
185 variation in CT-DM-score. Introducing the grouping variable (CTDM-F group vs. control group)  
186 explained an additional 3.0% of the variation in CT-DM score and the change in  $R^2$  was  
187 significant,  $F(1,428) = 25.45, p < .001$ . The results show that after controlling for university, the  
188 grouping variable explained a significant proportion of the variation in CT-DM scores,  
189 demonstrating that the CTDM-F group had a significantly higher CT-DM score than the control  
190 group.

191 Group differences in CT-DM score components were assessed using ANCOVA and are  
192 presented in Table 3. Results show a greater percentage of the intervention group made a  
193 decision and used significantly more evidence-based reasons to support their decision when  
194 compared to the control. However, there were no differences between groups in recognizing the  
195 other side's point of view.

196 After controlling for university using repeated measures ANCOVA, there were no  
197 between group differences in CTD or GE over time. However, there was a significant within-  
198 group change in CTD for both the CTDM-F group from a mean baseline value of 3.52 ( $\pm$ .6) to a  
199 post-intervention mean of 3.63 ( $\pm$ .6),  $p < .01$  and the control group from a baseline mean of 3.54  
200 ( $\pm$ .6) to a post-intervention mean of 3.69 ( $\pm$ .6),  $p < .001$ . There were no significant within group  
201 changes for GE behavior.

202 For the exploratory SEM path analysis, three model versions were hypothesized and  
203 tested using EQS software<sup>33</sup>: direct, predictive, and mediational models. Findings revealed that  
204 compared to a direct model with only a single predictor from group to CT-DM score, and a  
205 mediational model with CTD and GE behavior as mediators, a third prediction model with paths  
206 from three predictors (group, CTD, and GE behavior) fit best.

207 Macro-level fit indices showed that the  $\chi^2$ , df, CFI, and RMSEA were all in a near-  
208 optimal range for the selected prediction model (Table 4). In contrast, fit indices for the direct  
209 effect and mediational pathway indicated that these models were not adequately describing the  
210 data. Standardized maximum likelihood parameter estimates for the prediction model path  
211 coefficients are shown in Figure 1, along with  $R^2$  values. The results indicate that group and  
212 interest, (i.e., GE behavior) were significant predictors of CT-DM score, but CTD score was not  
213 a significant predictor of CT-DM score.

214

215

## DISCUSSION

216

217         The purpose of this study was to evaluate the effectiveness of a CTDM-F to facilitate CT-  
218 DM skills in large introductory STEM classes. Findings showed that the CTDM-F, provided as  
219 an out-of-class activity, was successful at encouraging students to use more CT-DM skills when  
220 compared to the control group. Specifically, The CTDM-F group was better at making a decision  
221 and supporting that decision with a greater number of evidence-based reasons. These findings  
222 were similar to other interventions that were implemented in semester-long courses<sup>34</sup>,  
223 highlighting that online modules with an appropriate framework to facilitate learning can be used  
224 as supplemental instruction to support higher order thinking skills. This study demonstrated the  
225 importance of incorporating an instructional framework to intentionally teach skills associated  
226 with CT-DM as well as encourage the significance of skill development for using critical  
227 thinking skills during a decision-making scenario.

228         The success of the online interactive modules in facilitating CT-DM can be attributed to  
229 grounding the curriculum within the social constructivism theory, using PBL, and scaffolding the  
230 learning to help students move towards independence in their CT-DM skills. The findings in this  
231 study are reinforced by Perry et al.<sup>35</sup>, who implemented similar methods using PBL activities  
232 throughout a semester-long course in a large introductory class. Their findings showed that  
233 students increased their critical thinking skills from the beginning of the semester compared to  
234 the end when exposed to competing viewpoints of topics related to sustainability. Developing  
235 critical thinking skills is essential because both educators and employers have expressed the need  
236 for students to graduate from higher education with strong critical thinking skills.<sup>36, 37</sup> However,

237 developing these skills require use of curricula that implement evidence-based instructional  
238 practices based on research in the science of teaching and learning.

239         Interestingly, while the experimental group made a decision more often and used more  
240 evidence to support their decision, the CTDM-F was not successful at having students recognize  
241 the differing perspective. One reason for this may be because a discussion activity was not  
242 included, which has shown to encourage recognition of other perspectives and to identify the  
243 other side's point of view.<sup>38</sup> Extending the framework by adding an additional module with more  
244 scaffolding and a discussion activity would help students move towards greater independence in  
245 their CT-DM response.

246         The current study found no significant differences between groups in change in attitudes  
247 toward critical thinking measured by CTD; both groups increased scores from baseline to post.  
248 The increase in the CTD scores over time can be explained by both groups being exposed to the  
249 same components of PBL instruction such as the critical thinking video, the competing narratives  
250 in both modules and the decision-making activity.<sup>39,40</sup> Additionally, there were no changes in  
251 GE behavior from pre-test to post-test or between groups. This suggests that even though the  
252 modules provided pros and cons of different environmental eating positions, additional  
253 intervention components such as goal setting may be needed to lead to behavior change.<sup>25</sup>

254         The path analysis showed that the CTDM-F and GE behavior were significant,  
255 independent predictors of CT-DM score, while CTD was not. This provides additional support  
256 that the framework encouraged the use of CT-DM skills. The path analysis showed that a  
257 prediction model fit better than a mediation or null model. This indicates that interest, or  
258 frequency of GE behavior, is an important variable to consider when predicting CT-DM score,  
259 but it was not a mediating variable, so it did not strengthen or weaken the CT-DM score.



260 Researchers have found that when students express more interest in a topic, they have a  
261 greater motivation to engage in critical thinking activities during low stakes assessment (e.g.,  
262 extra credit).<sup>41, 42</sup> Bruna et al.<sup>21</sup> found that when PBL was used to teach metabolism using  
263 nutrition and health articles, instead of a traditional lecture format, students reported more  
264 motivation to participate, engaged more in self-directed learning, and reported that the learning  
265 process encouraged them to use critical thinking skills. The authors contributed the positive  
266 outcomes of the study to the use of authentic education material where students were solving  
267 real-world problems. This supports why GE behavior was a significant predictor of CT-DM  
268 score. While GE may have led to a higher CT-DM score for those reporting higher GE behavior,  
269 it may have reduced CT-DM scores for those students who were not as involved in sustainable  
270 eating behavior. Nevertheless, the importance of interest in a topic was reinforced by the path  
271 analysis findings.

272 The path analysis also revealed that CTD score did not account for a significant amount  
273 of variation in CT-DM score. This is conflicting with other studies that have found that CTD is  
274 an indicator of critical thinking skills<sup>43</sup>, however the main outcome for this study measured CT-  
275 DM skills rather than the larger construct of critical thinking, which is a possible explanation as  
276 to why CTD was not a significant predictor of the outcome. Furthermore, having a natural  
277 inclination to be a critical thinker does not equate to having strong critical thinking skills.<sup>31</sup> For  
278 example, students may have recognized the need to use critical thinking in their courses (i.e.,  
279 measured by CTD score) but may not have progressed to the stage of implementing those  
280 strategies (i.e., measured by CT-DM score). Overall, the CT-DM framework intervention was  
281 not sufficient to cause a change in students' CTD when compared to the control.

282           The previous discussion brings forth a limitation. This study did not measure the broader  
283 construct of critical thinking, which limits the ability to completely understand how the CTDM-F  
284 impacted students' ability to think critically. Despite the limitation, this intervention used a  
285 rigorous study design and introduced a novel approach that facilitated CT-DM skills by using an  
286 online interactive framework that can be easily administered online or in a classroom setting and  
287 be manipulated to fit multiple topics within the STEM fields and beyond.

288

### 289                           **IMPLICATIONS FOR FUTURE RESEARCH AND PRACTICE**

290

291           This research supports the importance of including evidence-based instructional practices  
292 such as social constructivism theory and PBL when developing curricula, along with using topics  
293 that students can relate to such as examining sustainable food system practices and making food  
294 choices. Through the science of teaching and learning educators have the opportunity to develop  
295 curricula that encourages students to consider and evaluate facts, which is imperative for STEM  
296 students to master as they begin their careers. As future professionals, this generation faces  
297 future problems that are unpredictable and while information is more abundant than ever, the use  
298 of the internet and technology requires information seekers to be critical of the facts they  
299 consider when making complex decisions.

300           To further enhance the science of teaching and learning, the CTDM-F should be tested  
301 and administered in classroom settings to examine how a framework can be used to develop  
302 critical thinking skills and help students make informed decisions. Future implications include  
303 understanding how the CTDM-F can work to overcome pre-existing biases or emotional  
304 reasoning during a decision making scenario, along with measuring the broader construct of

305 critical thinking using validated measures.<sup>10, 44</sup> This would provide a more complete  
306 understanding of how critical thinking impacts students' decisions. As evident by this study,  
307 using a framework within two online interactive modules was found to be a facilitator of CT-DM  
308 and students were more likely to recognize and use facts when making a simulated decision  
309 about food choices.

## 310 REFERENCES

- 313 1. Butler HA. Halpern Critical Thinking Assessment Predicts Real- World Outcomes of  
314 Critical Thinking. *Applied Cognitive Psychology*. 2012;26(5):721-9.
- 315 2. Desai MS, Berger BD, Higgs R. Critical thinking skills for business school graduates as  
316 demanded by employers: a strategic perspective and recommendations. *Academy of*  
317 *Educational Leadership Journal*. 2016;20(1).
- 318 3. Dwyer CP, Hogan MJ, Stewart I. An integrated critical thinking framework for the 21st  
319 century. *Thinking Skills and Creativity*. 2014;12:43-52.
- 320 4. Franco A, Costa PS, Almeida LS. Do critical thinkers drink too much alcohol, forget to  
321 do class assignments, or cheat on exams? Using a critical thinking measure to predict  
322 college students' real-world outcomes. *Psychological Studies*. 2017:1-10.
- 323 5. Knight JK, Wood WB. Teaching more by lecturing less. *Cell Biol. Educ*. 2005;4(4):298-  
324 310.
- 325 6. Gasiewski JA, Eagan MK, Garcia GA, et al. From gatekeeping to engagement: A  
326 multicontextual, mixed method study of student academic engagement in introductory  
327 STEM courses. *Research in higher education*. 2012;53(2):229-61.
- 328 7. Duron R, Limbach B, Waugh W. Critical thinking framework for any discipline.  
329 *International Journal of Teaching and Learning in Higher Education*. 2006;17(2):160-6.
- 330 8. Rudd R, Baker M, Hoover T. Undergraduate agriculture student learning styles and  
331 critical thinking abilities: Is there a relationship? *Journal of agricultural education*.  
332 2000;41(3):2-12.
- 333 9. Scheffer BK, Rubenfeld MG. A consensus statement on critical thinking in nursing.  
334 *Journal of Nursing Education*. 2000;39(8):352-9.
- 335 10. Liu OL, Frankel L, Roohr KC. Assessing Critical Thinking in Higher Education: Current  
336 State and Directions for Next- Generation Assessment. *ETS Research Report Series*.  
337 2014;2014(1):1-23.
- 338 11. Brown KL. From teacher-centered to learner-centered curriculum: Improving learning in  
339 diverse classrooms. *Education*. 2003;124(1):49.
- 340 12. Huba ME, Freed JE. Learner centered assessment on college campuses: Shifting the  
341 focus from teaching to learning. Boston, MA: Allyn and Bacon, 2000.
- 342 13. Hendry GD, Frommer M, Walker RA. Constructivism and problem- based learning.  
343 *Journal of further and higher education*. 1999;23(3):369-71.

- 344 14. Mughal F, Zafar A. Experiential learning from a constructivist perspective:  
345 Reconceptualizing the Kolbian cycle. *International Journal of Learning and*  
346 *Development*. 2011;1(2):27-37.
- 347 15. Taylor PC. Constructivism: value added. *International handbook of science education*.  
348 1998;2:1111-23.
- 349 16. Vygotsky LS. *Thought and Language*: MIT Press, Massachusetts Institute of Technology  
350 and John Wiley and Sons, 1962.
- 351 17. Hannafin MJ, Hill JR, Land SM. Student-centered learning and interactive multimedia:  
352 Status, issues, and implication. *Contemporary Education*. 1997;68(2):94.
- 353 18. Yew EH, Karen G. Problem-based learning: an overview of its process and impact on  
354 learning. *Health Professions Education*. 2016;2(2):75-9.
- 355 19. Kim K, Sharma P, Land SM, et al. Effects of Active Learning on Enhancing Student  
356 Critical Thinking in an Undergraduate General Science Course. *Innovative Higher*  
357 *Education*. 2013;38(3):223-35.
- 358 20. Kek M, Yih Chyn A, Huijser H. The power of problem- based learning in developing  
359 critical thinking skills: preparing students for tomorrow's digital futures in today's  
360 classrooms. *Higher Education Research & Development*. 2011;30(3):329-41.
- 361 21. Bruna CE, Valenzuela NA, Bruna DV, et al. Learning Metabolism by Problem- Based  
362 Learning Through the Analysis of Health or Nutrition Articles from the Web in  
363 Biochemistry. *Journal of Food Science Education*. 2019;18(2):37-44.
- 364 22. Pintrich PR, Smith DAF, García T, et al. Reliability and predictive validity of the  
365 Motivated Strategies for Learning Questionnaire (MSLQ). *Educational and*  
366 *psychological measurement*. 1993;53(3):801-13.
- 367 23. Trosset C, Weisler S. Using longitudinal assessment data to improve retention and  
368 student experiences. *New Directions for Institutional Research*. 2010;2010(S2):79-88.
- 369 24. Weller KE, Greene GW, Redding CA, et al. Development and validation of green eating  
370 behaviors, stage of change, decisional balance, and self-efficacy scales in college  
371 students. *J Nutr Educ Behav*. 2014;46(5):324-33.
- 372 25. Monroe JT, Lofgren I, Sartini BL, et al. The Green Eating Project: Web-based  
373 intervention to promote environmentally conscious eating behaviors in United States  
374 university students. *Public Health Nutrition*. 2015;18(13):2368-78.
- 375 26. Chaillé C. *Constructivism across the curriculum in early childhood classrooms: Big ideas*  
376 *as inspiration*: Allyn & Bacon, 2008.
- 377 27. Brand-Gruwel S, Wopereis I, Vermetten Y. Information problem solving by experts and  
378 novices: Analysis of a complex cognitive skill. *Computers in Human Behavior*.  
379 2005;21(3):487-508.
- 380 28. Wheeler LA, & Collins, S. K. . The influence of concept mapping on critical thinking in  
381 baccalaureate nursing students. *J Prof Nurs*. 2003;19(6):339-46.
- 382 29. Rosen Y, Tager M. Making Student Thinking Visible through a Concept Map in  
383 Computer-Based Assessment of Critical Thinking. *Journal of Educational Computing*  
384 *Research*. 2014;50(2):249-70.
- 385 30. Horacek TM, Erdman MB, Byrd-Bredbenner C, et al. Assessment of the dining  
386 environment on and near the campuses of fifteen post-secondary institutions. *Public*  
387 *Health Nutr*. 2013;16(7):1186-96.

- 388 31. Stupnisky RH, Renaud RD, Daniels LM, et al. The interrelation of first-year college  
389 students' critical thinking disposition, perceived academic control, and academic  
390 achievement. *Research in Higher Education*. 2008;49(6):513-30.
- 391 32. Harlow LL. *The essence of multivariate thinking: Basic themes and methods*: Routledge,  
392 2014.
- 393 33. Bentler PM. *EQS Structural Equations Program Manual*. Encino, CA: Multivariate  
394 Software Inc., 2006.
- 395 34. Şendağ S, Ferhan Odabaşı H. Effects of an online problem based learning course on  
396 content knowledge acquisition and critical thinking skills. 2009:132-41.
- 397 35. Perry DK, Burgess MH, Sowell BF, et al. Using Competing Narratives to Increase  
398 Critical Thinking Abilities. *NACTA Journal*. 2017;61(1):41.
- 399 36. Accreditation Council for Education in Nutrition and Dietetics, Commission on Dietetic  
400 Registration, Council on Future Practice, Education Committee, Nutrition & Dietetics  
401 Educators and Preceptors DPG. 2013.
- 402 37. Easterly T, Warner AJ, Myers BE, et al. Skills Students Need in the Real World:  
403 Competencies Desired by Agricultural and Natural Resources Industry Leaders. *Journal*  
404 *of Agricultural Education*. 2017;58(4):225-39.
- 405 38. Guiller J, Durndell A, Ross A. Peer interaction and critical thinking: Face-to-face or  
406 online discussion? *Learning and instruction*. 2008;18(2):187-200.
- 407 39. Abrami PC, Bernard RM, Borokhovski E, et al. Instructional interventions affecting  
408 critical thinking skills and dispositions: A stage 1 meta-analysis. *Review of Educational*  
409 *Research*. 2008;78(4):1102-34.
- 410 40. Dochy F, Segers M, Van den Bossche P, et al. Effects of problem-based learning: A  
411 meta-analysis. *Learning and instruction*. 2003;13(5):533-68.
- 412 41. Liu OL, Bridgeman B, Adler RM. Measuring learning outcomes in higher education:  
413 Motivation matters. *Educational Researcher*. 2012;41(9):352-62.
- 414 42. Weiler A. Information-seeking behavior in generation y students: Motivation, critical  
415 thinking, and learning theory. *J Acad Libr* 2005;31(46):53.
- 416 43. Profetto-McGrath J. The relationship of critical thinking skills and critical thinking  
417 dispositions of baccalaureate nursing students. *J. Adv. Nurs*. 2003;43(6):569-77.
- 418 44. Abrami PC, Bernard RM, Borokhovski E, et al. Strategies for teaching students to think  
419 critically: A meta-analysis. *Review of Educational Research*. 2015;85(2):275-314.
- 420

**Table 1: Differences in Module 1 and 2 Design Between Critical Thinking Decision Making Framework (CTDM-F) Group and Control Group**

	<b>CTDM-F Group</b>	<b>Control Group</b>
<b>Module 1:</b>		
<b>Animal/Plant Protein</b>		
Topic	Introduced in the form of a problem	Introduced as the topic
Support	T-Chart Mind-Map	None
Decision-Making Activity	Structured fill in the blanks	Text box
<b>Module 2:</b>		
<b>Organic/Non-Organic</b>		
Topic	Introduced in the form of a problem	Introduced as the topic
Support	T-Chart Mind-Map	None
Decision-Making Activity	Text box	Text box

<b>Table 2: Baseline Demographics by Critical Thinking Decision-Making Framework (CTDM-F) Group and Control Group</b>			
<b>Variable</b>	<b>CTDM-F (n=203) mean (std)</b>	<b>Control (n=228) mean (std)</b>	<b>F-Value</b>
<b>Age</b>	19.3 (1.3)	19.4 (1.5)	1.82
<b>Pre-CTD<sup>a</sup></b>	3.51 (.6)	3.54 (.6)	.50
<b>Pre-GE Behavior<sup>b</sup></b>	2.72 (.7)	2.76 (.8)	.29
<b>Variable</b>	<b>CTDM-F (n=203) %</b>	<b>Control (n=228) %</b>	<b>Pearson Chi-Square</b>
<b>Gender</b>			.03
<b>Male</b>	26.6	26.0	
<b>Female</b>	73.4	74.0	
<b>Major<sup>c</sup></b>			.85
<b>Arts and Humanities</b>	42.4	42.4	
<b>STEM</b>	23.6	20.5	
<b>STEM-Health</b>	28.6	31.4	
<b>Undecided</b>	5.4	5.7	
<b>University<sup>d</sup></b>			1.06
<b>University 1</b>	50.7	55.7	
<b>University 2</b>	49.3	44.3	

<sup>a</sup>CTD= Critical Thinking Disposition, Motivated Strategies for Learning Questionnaire (MSLQ), mean score of five items measured using a 5-point Likert scale: (1) strongly disagree (5) strongly agree.  
<sup>b</sup>Pre-GE Behavior= Pre-Green Eating Behavior, 7-item survey, 5-point Likert scale ranging from (1) barely ever to never to (5) almost always.  
<sup>c</sup>Arts and Humanities= social sciences, arts, and undecided; STEM Majors= Science, Technology, Engineering, and Math; STEM-Health Majors= nutrition, kinesiology, nursing, pre-med  
<sup>d</sup>University 1= University of Rhode Island; University 2= University of North Texas  
Continuous variables assessed using ANCOVA, controlling for university  
Categorical variables assessed using Pearson Chi-Square

<b>Table 3: Differences in Critical Thinking Decision-Making Framework (CTDM-F) Group Responses and Control Group Responses in Module 2 After Controlling for University</b>				
<b>Variable</b>		<b>CTDM-F (n=203)</b>	<b>Control (n=228)</b>	<b>F-Value</b>
<b>CT-DM Score<sup>a</sup> mean (std)</b>		18.1 (7.6)	15.4 (8.4)	14.58**
<b>Number of Evidence-Based Reasons<sup>b</sup> mean (std)</b>		1.7 (1.1)	1.3 (1.0)	15.21**
<b>Variable</b>		<b>CTDM-F (n=203)</b>	<b>Control (n=228)</b>	<b>Pearson Chi-Square</b>
<b>Made a decision (%)</b>				
	Yes	91.1	80.3	
	No	8.9	19.7	10.17**
<b>Reason (% reported)</b>				
	Health	39.5	31.7	3.93
	Animal	2.4	1.7	2.04
	Price	12.2	9.1	1.94
	Economic	4.4	4.3	.00
	Environment	50.7	47.0	.62
	Other	3.4	4.4	.25
<b>Identified other perspective (%)</b>				
	Yes	14.3	18.4	
	No	85.7	81.6	1.34

<sup>a</sup>Critical Thinking Decision Making (CT-DM) Score: range from 0-30, 0= no CT-DM and 30=high CTDM

<sup>b</sup>Number of evidence-based reasons used to support decision: range from 0 to 3, higher value = more reasons used to support decision

Continuous variables assessed using ANCOVA, controlling for university

Categorical variables assessed using Pearson Chi-Square

\*p<.05

\*\*p<.01



<b>Table 4: Structural Equation Modeling Macro-Level Fit Indices for Direct, Predictive, and Mediation Pathways Explaining Critical Thinking Decision Making Score</b>						
<b>Pathway</b>	<b>Macro-Level Fit Indices<sup>d</sup></b>					
	<b>X<sup>2</sup></b>	<b>df</b>	<b>X<sup>2</sup>/df ratio</b>	<b>CFI</b>	<b>RMSEA</b>	<b>X<sup>2</sup> difference</b>
Direct <sup>a</sup>	21.82*	5	4.36	.302	.093	---
Mediation <sup>b</sup>	21.72*	2	10.86	.181	.159	.10
Predictive <sup>c</sup>	14.65*	3	4.88	.516	.100	7.07

<sup>a</sup>Direct pathway from group to critical thinking decision making (CT-DM) score  
<sup>b</sup>Mediation pathway with group as independent variable and GE behavior and CTD as mediating variables explaining CT-DM score  
<sup>c</sup>Predictive pathway with group, green eating (GE) behavior, and critical thinking disposition (CTD) predicting CT-DM score  
<sup>d</sup>Ideal macro-level fit indices parameters: X<sup>2</sup>p > 0.05, X<sup>2</sup>/df ratio < 4, CFI > 0.90, RMSEA < 0.08, X<sup>2</sup> difference= larger value equates to better fit  
 \*p < .001

Figure 1: Prediction Model Pathway with Parameter Estimates for Critical Thinking Disposition Score, Group, and Green-Eating Behavior

