The positive association between countries’ scientific and technological knowledge and their level of development is a well-known fact. Although all individuals have an important potential for their society, this knowledge that will lead a society to a high civilization level is only possible by individuals who are able to execute high-level cognitive functioning, who can realize and offer various explanations for solving problems, and who can excel in different fields. Watters and Diezmann (2003) stated that people with special talent have a great role in the development of countries. In the Special Education Services Regulation (MONE, 2016), the concept of special talent is considered together with giftedness and an individual with special talent—or gifted individual—is defined as “an individual who performs at a higher level than his/her peers in intelligence, creativity, art, sports, leadership, or special academic fields” (MONE, 2016, p. 2). Similarly, Hoh (2008) and Davis and Rimm (2004) consider gifted individuals who are noticeable in the society in terms of cognitive, affective, and psychomotor learning domains as individuals who can display superior analytical skills and have high-level thinking skills such as abstract, logical, creative, and critical thinking. These characteristics point to requirement of challenging and enriched teaching in gifted education.

The fact that science is a challenging field that contains abstract and complex elements is a critical feature that causes science to attract highly gifted individuals’ attention (Tomlinson, 2005). Moreover, science contributes to practical skills and includes activities that require the use of mind and hands, which make this area more attractive for gifted individuals (Morris et al., 2019). The nature of science requires making generalizations with the help of predictions using scientific data to provide an enriched and open-ended learning process, which is a necessity for educating gifted individuals. The reflection of these processes in formal science education is a desirable circumstance for those individuals’ education. The ability to provide rich, open-ended, and challenging learning processes in science courses is related to how much intellectual risk is allowed in the teaching environment, how positive the learning environment perception is, and how much motivational support is provided (Beghetto, 2009; Glynn et al., 2005; Marlowe and Page, 2005). Therefore, this study aimed to examine the associations among constructive learning environment perception (CLEP), motivation to learn science (MLS), intellectual risk-taking (IRT), and science achievement (SA).
However, there exist a limited number of studies focusing on IRT skills. More specifically, IRT skills affect achievement in science. Gifted students tend to be successful through animations. According to the results, animations were implemented in the classroom, asking questions to their friends or teachers, making explanations about a topic discussed in the classroom, offering solutions to a problem that has not been tried before, and being able to use the previously learned knowledge in new activities (Beghetto, 2009; House, 2002; Neihart, 1999). Specifically, gifted individuals have a very high potential in showing many behaviors related to IRT (Akdağ and Köksal, 2017). Rainwater and Wittner (2016) stated that gifted students’ high success in cognitive areas including SA, understanding, and problem solving is a result of their IRT behaviors. In addition, Soares et al. (2016) stated that gifted students focus on questioning the phenomenon related to a scientific problem within the scope of IRT rather than focusing on a single solution to solve the problem. Çakır and Yaman (2015) found that IRT skills and cognitive awareness levels of ordinary students were higher than average and there was a positive significant relationship between IRT skills, metacognitive awareness levels, and SA. Similar results were found for ordinary students by Özbay (2016). On the other hand, Akdağ and Köksal (2017) compared gifted students with the others in terms of IRT skills and found no difference between them. In another study, Daşcı and Yaman (2014) examined the change in IRT skills of gifted students with time. The results revealed that students’ skills decreased as their grade levels got higher. The researchers specifically found that by eight grade there is a straight decrease in students’ IRT skills. Akdağ et al. (2016) found a similar result for gifted students. These two studies may be considered as evidence of the need to examine the variables that might be related to the decrease in students’ IRT skills. In another study, Akkaya (2016) attempted to increase fourth grade students’ IRT skills through animations. According to the results, animations were found to be effective in increasing their skills as well as their achievement in science. Gifted students tend to be successful in science due to their IRT skills. More specifically, IRT skills predict students’ achievement in science (Beghetto, 2009). However, there exist a limited number of studies focusing on relationship of this variable with other important variables in science education. Therefore, examining the association between IRT behavior and other critical variables is critical to provide enriched learning processes for gifted students.

**Motivation in science learning**

Students’ ability to take intellectual risks to a certain extent is effective in increasing their motivation and SA (Beghetto, 2009; House, 2002). Pintrich (2003) defines motivation as a structure that includes intrinsic or extrinsic factors that affect the stimulation, maintenance, and control of a behavior. In addition, motivation is the most critical factor that affects SA of students who continues learning activities and that ensures the continuation of their success (Glynn et al., 2005; Guay et al., 2010; Martin, 2001; Pintrich, 2003; Schulze and Lemmer, 2017). Matthews and McBee (2007) also reveal that one of the main reasons in academic failure is lack of motivation, especially for gifted students. The motivation of a gifted student is a complex structure that is affected by content, materials, curriculum, teaching methods, and teacher’s and student’s personal characteristics (Lee and Brophy, 1996). In other words, the motivation of those individuals while learning science is influenced by IRT behaviors and key components of teaching (Lee and Brophy, 1996; Matthews and McBee, 2007).

There are studies focusing on gifted students and their MLS. Kilani and Emir (2009) designed a science program for gifted students and examined its effectiveness and its effect on students’ motivation. The results showed that the program positively affected gifted students’ motivation. Alkan and Bayri (2017) reported that many studies found evidence showing the positive association between students’ motivation and SA. However, in a study, Köksal (2013b) examined the association between motivation and students’ IQ level, logical thinking, and critical thinking and found no association between students’ motivation and the other variables. This revealed that there was an important gap between motivation and cognitive learning in terms of gifted students. Therefore, it is critical to examine the association between MLS and IRT skills and their effects on SA to plan enriched science curriculum and teaching for gifted students.

**Learning environment in science education**

Learning environment is another factor that affects students’ MLS and IRT behaviors (Nickerson, 1999; Yılmaz and Cavas, 2007). Active learning environments in which students are encouraged to use the skills and processes related to inquiry in science learning and to search about their own thoughts positively affect students’ MLS (Lang et al., 2005; Velayutham et al., 2012; Watters and Diezmann, 2003). Learning environments where scientific inquiries and research are carried out and problems are solved by using higher order thinking skills are constructive learning environments (Marlowe and Page, 2005). Students in these learning environments may easily share their comments or guesses with the others by thinking deeply about a problem even if they cannot predict the results in the solution of the problem.
These behaviors constitute the content of IRT, which indicates a relationship between IRT and CLEP as Nickerson (1990) stated. Rita and Martin-Dunlop (2011) examined students’ preferences in terms of learning environment and found that students preferred a more constructive learning environment than the one they were actually experiencing. In addition, the researchers revealed the positive and significant association between students’ academic achievement and CLEP. Similar results were found in another study conducted by Pramathevan and Fraser (2019). Moreover, CLEP has a significant effect in supporting students’ motivation for learning (Yılmaz and Çavaş, 2007). However, students’ general perspectives reveal that this environment are not well-established in schools of Turkey (Baş, 2012), which requires more research to fill the gap in the literature. In these schools, gifted students are also taking courses, but there is no study focusing on the association of their learning environment with motivation and IRT.

Based on the literature, it is concluded that the IRT, MLS, and CLEP variables are effective in increasing the classroom activities of gifted students and the quality of the teaching process. Therefore, it is critical to determine the association among those variables. The findings of this study will provide an evidence-based outline for enriched classroom activities.

**METHODOLOGY**

In this study, a correlational research approach was employed in order to examine the association among variables and to predict possible results. To provide a better understanding of the source of variation in the data set, a path analysis, a special technique of multiple regression analysis, was performed in the study (Daşdağ et al., 2006). Based on the literature review, a model was designed to represent the possible associations among the variables and is depicted in Figure 1.

In the path diagram, causal relationships between variables are shown with one-way arrows. Accordingly, when Figure 1 is examined, it is seen that the CLEP and MLS variables are direct and indirect predictors of the IRT variable and all these variables together are predictors of SA in science. In addition, since CLEP and MLS are related to each other, this association is expressed by a two-way arrow. In the path analysis technique, explanatory and predictive correlations between variables form a complex structure. Therefore, instead of separating variables as dependent and independent variables in path diagrams, predicted variables are generally classified as endogenous variables and predictive variables as exogenous variables (Bayram, 2013). As seen from the figure, CLEP and MLS are the exogenous variables of the hypothesis and IRT and SA are endogenous variables.

**Participants**

Participants of the current study were students of the Science and Art Centers (SAC). SAC are institutions that provide after-school programs for gifted students. Students need to pass two different stages of identification to be admitted to the program. The stages are prepared to identify gifted students and the students should have IQ scores higher than cutoff determined by the ministry. Students who are enrolled in SAC complete the following sessions throughout the academic year: Adaptation, support training, realization of individual abilities, development of special skills, and project training and management (MONE, 2016). There is an obligation to attend all sessions and students who complete the program are provided a certificate of completion.

The population of the study consists of eight-grade gifted students (Mean age = 14) in Turkey. Geographically, Turkey is divided into seven regions. For this particular study, one big city and one small city were identified from each region, which makes 14 cities in total. Those cities are Malatya, Elazığ, Gaziantep, Adıyaman, Adana, Osmaniye, İzmir, AFyon, Bursa, Bilecik, Samsun, Tokat, Sivas, and Ankara. There is at least one SAC in each city. Before conducting the study, necessary permissions were obtained from the Ministry of Education. Moreover, informed consent forms were also signed by the parents of the participants. The participants of the study consist of the SAC students (n = 132, females = 46 and males = 86) from those cities. A total of 133 students were recruited for the study. Due to missing values and invariance, one case was dropped out from the path analysis, which left 132 cases. Sample size is critical for validity and reliability of analysis.

**Data Collection Tools**

In this study, five different data collection tools were used: A demographic information form, the IRT in science scale (IRT-S), constructive learning environment scale, motivational strategies learning questionnaire, and an academic achievement test. The demographic information form included questions related to gender, age, grade level, and the SAC location the participants registered. The details of the other data collection tools are provided below.

**The IRT-S**

The scale was developed by Beghetto (2009) and adopted to Turkish by Yaman and Köksal (2014). It has six items under only one factor. Each item was measured with a five-point Likert type and responses ranged from not true (1) to very true (5). There was no reverse-coded item in the scale. An example item is “During science, I like doing new things even if I am not very good at them.” The good fit and acceptable fit indices (Büyükoztürk et al., 2016; İlhan and Çetin, 2014) and the fit indices calculated for the scale with one-factor in this study are provided in Table 1.

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![Figure 1: Initial hypothesized model for the relationship among science constructive learning environment perception, motivation to learn science, intellectual risk-taking, and academic achievement](image-url)
According to the results, while \( x^2/df \) value (1.764) was considered as good fit, the comparative fit index (CFI) (0.88) and goodness of fit index (GFI) values (0.95) were at acceptable level. The Cronbach’s alpha value was calculated as 0.68.

**The Constructivist Learning Environments Scale (CLES)**

The scale developed by Taylor and Fraser (1991) aims to measure students’ perceptions related to the frequency of occurrence of critical constructivist learning environment. Although the original scale had 30 items, it was revised by Johnson and McClure (2004) and reduced to 20 items. The items in the revised version of the scale were designed as a five-point Likert type ranging from never (1) to always (5). The scale has five factors including personal relevance, uncertainty, critical voice, shared control, and student negotiation. The Cronbach’s alpha coefficient was calculated as 0.91. An example item is “Students talk with other students about how to solve problems.” The fit indices obtained from the confirmatory factor analysis in this study, good and acceptable fit indices are provided in Table 2.

According to the table, \( x^2/df \) (1.15) and CFI (0.96) were at acceptable level. Although, GFI value was not among the acceptable values in Table 2, they were close to this level based on a study conducted by Büyüköztürk et al. (2004). Hence, we ignored the GFI value due to the importance of the variable in the study.

**The Motivational Strategies for Learning Questionnaire (MSLQ)**

To examine students’ motivational level in science courses, the MSLQ was used. The questionnaire was developed by Pintrich et al. (1991) and translated to Turkish by Sungur (2004). It consisted of two sections: motivation (31 items) and learning strategies (51 items). An example item is “It is important for me to learn scientific concepts.” For this particular study, only four sub-dimensions of the motivation section were used: Intrinsic goal orientation, task value, control of learning beliefs, and self-efficacy for learning and performance. In this study, only intrinsic goal orientation was used to measure motivation toward learning science. By this way, we reduced number of questions applied to the student. The Cronbach’s alpha coefficient was calculated as 0.83. The fit indices obtained from the confirmatory factor analysis; in this study, good and acceptable fit indices are provided in Table 3.

The table presents the good and acceptable fit indices (İlhan and Çetin, 2014) and the fit indices of the scale. The values for \( x^2/df \), CFI, and GFI were at the good fit range. Therefore, the scale perfectly fits the previously determined one-factor structure.

**Academic Achievement Test for Science (AAT-S)**

Another data collection tool is the AAT-S, which is a 45-item test developed by Aşut (2013). The items in the test were designed by taking into account the 6th, 7th, and 8th grade science curriculum. The descriptive results are presented in Table 4. The reliability level of the test was determined based on the KR-20 internal consistency coefficient, which was found to be 0.81. This value is an acceptable value according to Esin (2014). The average item difficulty index (0.64) and average discrimination index (0.33) values revealed that the test is at medium difficulty level (Pehlivan-Tuğc and Kutlu, 2014). An example item is provided below.

**Question:** Which of the following is not included in the “productive livings” category?

a. Human
b. Sunflower
c. Cyanobacteria
d. Blue-green algae.

<p>| Table 1: The fit index values for the scale based on the confirmatory factor analysis |
|--------------------------------------|------------------|-------------------|-----------------------------|</p>
<table>
<thead>
<tr>
<th><strong>Fit indexes</strong></th>
<th><strong>Good fit values</strong></th>
<th><strong>Acceptable values</strong></th>
<th><strong>One-factor structure values</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td>( x^2/df )</td>
<td>( 0 \leq x^2/df \leq 2 )</td>
<td>( 2 \leq x^2/df \leq 3 )</td>
<td>1.764</td>
</tr>
<tr>
<td>CFI</td>
<td>0.97 ≤ CFI ≤ 1</td>
<td>0.80 ≤ CFI ≤ 0.97</td>
<td>0.88</td>
</tr>
<tr>
<td>GFI</td>
<td>0.95 ≤ GFI ≤ 1</td>
<td>0.90 ≤ GFI ≤ 0.95</td>
<td>0.95</td>
</tr>
</tbody>
</table>

GFI: Goodness of fit index, CFI: Comparative fit index

<p>| Table 2: The fit index values for the scale based on the confirmatory factor analysis |
|--------------------------------------|------------------|-------------------|-----------------------------|</p>
<table>
<thead>
<tr>
<th><strong>Fit indexes</strong></th>
<th><strong>Good fit values</strong></th>
<th><strong>Acceptable values</strong></th>
<th><strong>Five-factor structure values</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td>( x^2/df )</td>
<td>( 0 \leq x^2/df \leq 2 )</td>
<td>( 2 \leq x^2/df \leq 3 )</td>
<td>1.15</td>
</tr>
<tr>
<td>CFI</td>
<td>0.97 ≤ CFI ≤ 1</td>
<td>0.95 ≤ CFI ≤ 0.97</td>
<td>0.96</td>
</tr>
<tr>
<td>GFI</td>
<td>0.95 ≤ GFI ≤ 1</td>
<td>0.90 ≤ GFI ≤ 0.95</td>
<td>0.78</td>
</tr>
</tbody>
</table>

GFI: Goodness of fit index, CFI: Comparative fit index

<p>| Table 3: The fit index values for the scale based on the confirmatory factor analysis |
|--------------------------------------|------------------|-------------------|-----------------------------|</p>
<table>
<thead>
<tr>
<th><strong>Fit indexes</strong></th>
<th><strong>Good fit values</strong></th>
<th><strong>Acceptable values</strong></th>
<th><strong>One-factor structure values</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td>( x^2/df )</td>
<td>( 0 \leq x^2/df \leq 2 )</td>
<td>( 2 \leq x^2/df \leq 3 )</td>
<td>0.6</td>
</tr>
<tr>
<td>CFI</td>
<td>0.97 ≤ CFI ≤ 1</td>
<td>0.95 ≤ CFI ≤ 0.97</td>
<td>0.99</td>
</tr>
<tr>
<td>GFI</td>
<td>0.95 ≤ GFI ≤ 1</td>
<td>0.90 ≤ GFI ≤ 0.95</td>
<td>0.99</td>
</tr>
</tbody>
</table>

GFI: Goodness of fit index, CFI: Comparative fit index

| Table 4: Descriptive statistics and item analysis results of the test |
|------------------------|-----------------------------|
| **Statistics**         | **Values**                 |
| Number of items        | 45                          |
| Number of participants | 63                          |
| Average Score          | 29.25                       |
| Standard Deviation     | 6.39                        |
| Variance               | 40.87                       |
| Minimum score          | 12                          |
| Maximum score          | 39                          |
| Average difficulty index| 0.64                       |
| Average discrimination index | 0.33                   |
| KR-20 (Alpha) Reliability | 0.81                     |
Data Analysis
To reveal the causal relationships between variables, a path analysis was conducted. Before the path analysis, validity and reliability analyses were conducted for each data collection tool and the results were provided above. Then, average scores were calculated for each of them and screened to determine whether the data were normally distributed. According to the skewness and kurtosis values, the data were normally distributed. Therefore, parametric tests were run for analysis.

Before testing the initial model, the direction and degree of relationships between variables were determined using Pearson product-moment correlation coefficients. In addition, multivariate normality was examined by calculating Mardia (1970) multivariate kurtosis coefficient values. The result revealed that the data were suitable for the path analysis. The IBM SPSS AMOS version 21 was used to calculate the path coefficients. The direct, indirect, and total effects of the variables on each other were determined with the calculated path coefficients. The compatibility of the initial model with the data was examined by calculating the fit indices. The differences between the expected and observed data matrices were examined and the concordance between matrices was revealed. The results are provided below.

Findings
The findings of the study are discussed under two sub-headings as descriptive findings and inferential findings.

Descriptive findings
The descriptive results related to IRT-S, CLES, MSLQ, and AAT-S of eight grade-gifted students are provided in Table 5.

According to the findings, students had above average score for AAT-S, IRT-S, CLES, and MSLQ. The skewness values ranged between –0.96 and 0.12 and the kurtosis values ranged between 0.167 and 2.4. These values revealed that the data were normally distributed (Tabachnick and Fidell, 2013).

INFERENTIAL RESULTS
To identify the direct and indirect effect of SA, IRT, CLEP, and MLS path analysis was performed. Before discussing the path analysis results, it is critical to examine the one-way correlations among those variables. Table 6 provides the correlation matrix of all the factors.

According to the matrix, significant correlations were found between CLEP and MLS, CLEP and IRT, MLS and IRT, and IRT and SA. Among them, the one between IRT and SA was negative. There was no correlation between CLEP and SA and MLS and SA. The descriptive findings and the correlation matrix values revealed that the data were suitable for path analysis except for the association between CLEP and SA. The hypothesized model is provided in Figure 2.

Figure 2, it is seen that MLS and CLEP directly influence IRT, and IRT directly influence SA. In addition, MLS and CLEP indirectly influence SA through IRT. CLEP and MLS are the exogenous variables of the model, IRT and SA are the endogenous variables. In this model presented in Figure 2, the variables in the rectangle area are the observed variables of the model; therefore, each has a standardized error term (e1, e2, e3, and e4), which are represented in the model with a one-way arrow as the exogenous unobserved variable. To test the model, path analysis was performed and the results are provided in Figure 3.

Table 5: Descriptive results

<table>
<thead>
<tr>
<th>Values</th>
<th>AAT-S</th>
<th>IRT-S</th>
<th>CLEP</th>
<th>MSLQ</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of participants</td>
<td>132</td>
<td>132</td>
<td>132</td>
<td>132</td>
</tr>
<tr>
<td>Mean</td>
<td>30.06/45</td>
<td>3.74/5</td>
<td>3.66/5</td>
<td>4.28/6</td>
</tr>
<tr>
<td>Standard deviation</td>
<td>5.8</td>
<td>0.74</td>
<td>0.65</td>
<td>0.51</td>
</tr>
<tr>
<td>Minimum</td>
<td>13</td>
<td>1.2</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>Maximum</td>
<td>39</td>
<td>5.00</td>
<td>4.80</td>
<td>6.4</td>
</tr>
<tr>
<td>Skewness</td>
<td>–0.78</td>
<td>–0.96</td>
<td>–0.71</td>
<td>0.11</td>
</tr>
<tr>
<td>Kurtosis</td>
<td>–4.52</td>
<td>3.33</td>
<td>0.55</td>
<td>3.67</td>
</tr>
<tr>
<td>Standard error for skewness</td>
<td>–0.52</td>
<td>3.91</td>
<td>0.51</td>
<td>3.67</td>
</tr>
<tr>
<td>Standard error for kurtosis</td>
<td>0.39</td>
<td>2.81</td>
<td>3.27</td>
<td>5.81</td>
</tr>
</tbody>
</table>

Table 6: Correlation coefficient (Pearson r) Matrix

<table>
<thead>
<tr>
<th>Variables</th>
<th>CLEP</th>
<th>MLS</th>
<th>IRT</th>
<th>SA</th>
</tr>
</thead>
<tbody>
<tr>
<td>CLEP</td>
<td>-</td>
<td>0.33*</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td>MLS</td>
<td>0.33*</td>
<td>-</td>
<td>0.57*</td>
<td>-</td>
</tr>
<tr>
<td>IRT</td>
<td>0.57*</td>
<td>0.29*</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>AA</td>
<td>0.01</td>
<td>0.02</td>
<td>–0.04*</td>
<td>-</td>
</tr>
</tbody>
</table>

*<0.05, CLEP: Constructive learning environment perception, IRT: Intellectual risk-taking, SA: Science achievement, MLS: Motivation to learn science.
According to Suhr (2008), inter-variable path coefficients are divided into three categories: small effect if path coefficients $<0.1$, medium effect if between 0.1 and 0.5, and large effect if $>0.5$. According to Figure 3, there was a large effect between CLEP and IRT ($β = 0.53$), a medium effect between MLS and IRT ($β = 0.17$), IRT and SA ($β = 0.15$), and CLEP and MLS ($β = 0.23$), and a small effect between CLEP and SA ($β = -0.01$) and MLS and SA ($β = -0.02$). In addition, the fit indices of the model were not acceptable. To improve the model fit, the model was modified by removing the paths between CLEP and SA and MLS and SA (Figure 4).

According to the modified model, although CLEP and MLS did not directly influence SA, they indirectly influenced it through IRT. Furthermore, IRT directly influenced SA, which revealed the association between them. There was a correlation between CLEP and MLS. In order to test the model, fit indices ($χ^2$/df ratio, GFI, root-mean-square error of approximation [RMSEA], adjusted goodness-of-fit index [AGFI], and CFI) need to be calculated through path analysis (Bentler and Yuan, 1999; Pedhazur, 1997). The good and acceptable fit indices and the fit indices of the model are provided in Table 7 (Baumgartner and Homburg, 1996; Byrne, 2010; Kline, 2011; Meyers et al., 2016; Schermelleh-Engel and Moosbrugger, 2003). The $χ^2$/df ratio of the modified model (0.04) was among the good fit criteria.

Accordingly, there was no significant difference between the observed covariance matrix and the estimated covariance matrix; therefore, the model was found to be fit. Another fit index value in Table 7 is the GFI and it shows to what extent the developed model measures the covariance matrix of the sample (Çokluk et al., 2012). The value above 0.95 proves the perfect fit of the model. The GFI value for the modified model was calculated as 0.99, which shows that the model was the perfect fit. Furthermore, the adjusted GFI was found to be 0.99, which is in the perfect fit range. The CFI examines the model fit by examining the disagreement between the data and the hypothesized model (Çokluk et al., 2012). According to the table, the CFI value verified that the model was perfect fit. The RMSEA value compares the observed covariance matrix with the hypothesized covariance matrix. Unlike GFI and AGFI, this value is expected to be close to zero with the 90% confidence interval (Schumacker and Lomax, 2010). According to the findings, the RMSEA value of 0.00 verified the perfect fit.

According to the findings, the modified model fit perfectly with the data. However, it is critical to examine the direct and indirect effects among variables. These relationships between variables were addressed based on the path coefficients and variance rates. The path coefficient values are given in Figure 5.

According to the results provided in Figure 5, the explanatory relationships between variables were at a significant level ($<0.05$). In other words, a cause-effect relationship was established between variables. The path coefficient values between the variables and the explained variance ratios were $β = 0.39$, $R^2 = 0.15$ between CLEP and MLS, $β = 0.53$, $R^2 = 0.28$ between CLEP and IRT, $β = 0.17$, $R^2 = 0.02$ between MLS and IRT, and $β = 0.14$, $R^2 = 0.01$ between IRT and SA.

The path coefficient value between CLEP and MLS was 0.39 ($<0.05$ and $t = 2.5$), which shows a linear relationship between the variables. In other words, students with a high perception of science learning environment had high intrinsic MLS.

The path coefficient value between CLEP and IRT was 0.53 ($<0.05$ and $t = 7.35$). One unit of change in the standard deviation of CLEP directly created a change of 0.53 in the standard deviation of IRT. According to Suhr (2008), there was a high correlation between the CLEP and IRT. As a result, students who had positive perception about constructive learning environments tended to take intellectual risks while learning science.

The path coefficient value between MLS and IRT was 0.17 ($<0.05$ and $t = 2.3$). MLS directly created a change of 0.17 in the standard deviation of IRT, which shows a medium level association between these two variables (Suhr, 2008). This result reveals that students with high level of intrinsic motivation took a moderate level of intellectual risks in science lessons.

The path coefficient value between IRT and SA was 0.14 ($<0.05$ and $t = 1.6$). IRT created a change of 0.14 in the standard deviation of IRT. Suhr (2008) considers this change as a medium level association. This result may be interpreted that IRT behavior in science lessons affects students’ SA in science at a moderate level.

One of the advantages of path analysis compared with the other regression analysis is that path analysis provides information about both direct and indirect effects among variables.

![Figure 4: The modified model](image-url)

### Table 7: Fit index values for the modified model

<table>
<thead>
<tr>
<th>Fit Indexes</th>
<th>Good fit</th>
<th>Acceptable</th>
<th>Modified fit indices</th>
</tr>
</thead>
<tbody>
<tr>
<td>$χ^2$/SD</td>
<td>$0 ≤ χ^2$/SD $&lt; 2$</td>
<td>$2 ≤ χ^2$/SD $≤ 3$</td>
<td>0.04</td>
</tr>
<tr>
<td>AGFI</td>
<td>$0.90 ≤ AGFI ≤ 1.00$</td>
<td>$0.85 ≤ AGFI ≤ 0.90$</td>
<td>0.99</td>
</tr>
<tr>
<td>GFI</td>
<td>$0.95 ≤ GFI ≤ 1.00$</td>
<td>$0.90 ≤ GFI ≤ 0.95$</td>
<td>0.99</td>
</tr>
<tr>
<td>CFI</td>
<td>$0.95 ≤ CFI ≤ 1.00$</td>
<td>$0.90 ≤ CFI ≤ 0.95$</td>
<td>0.99</td>
</tr>
<tr>
<td>RMSEA</td>
<td>$0.00 ≤ RMSEA ≤ 0.05$</td>
<td>$0.05 ≤ RMSEA ≤ 0.08$</td>
<td>0.00</td>
</tr>
</tbody>
</table>

GFI: Goodness of fit index, CFI: Comparative fit index, AGFI: Adjusted goodness-of-fit index, RMSEA: Root-mean-square error of approximation.
This advantage enables researchers to compare the direct and indirect effects and to find overall effect (Asher, 1983; Olobatuyi, 1992). As discussed previously, the path coefficients in the model are the direct effects. On the other hand, indirect effect is the effect level of the external or internal explanatory variable on the variable explained through another variable. The indirect effect value is calculated as the multiplication of a path coefficient between the explanatory variable and the mediating variable with a path coefficient between the mediating variable and the explained variable. The overall effect of the explanatory variable on the explained variable is the sum of the direct and indirect effects (Çokluk et al., 2012). The direct, indirect, and overall effects of the variables in the model are provided in Table 8.

According to the table, CLEP had an indirect effect (0.07) on SA through IRT. In other words, one unit change in the standard deviation of CLEP creates a change of 0.07 in the standard deviation of SA through IRT. Similarly, MLS had an indirect effect (0.02) on SA through IRT.

According to the total effect values between the variables, the exogenous variable CLEP caused a total change of 0.53 units on IRT and 0.07 units on SA and the exogenous variable MLS caused a total change of 0.17 units on IRT and 0.02 units on SA. Finally, IRT created a change of 0.14 units on SA. Based on the findings, the final model is provided in Figure 6.

Considering all findings, although the hypothesized model created based on the literature was not fully confirmed, it was largely supported by the data. The results of the path analysis revealed that students’ CLEP positively affects their IRT behaviors in learning science and SA. A similar influence was observed for students’ MLS. In addition, IRT behaviors predicted students’ SA in science.

**DISCUSSIONS, CONCLUSION, AND RECOMMENDATIONS**

This study examined eight-grade gifted students’ MLS, perceptions about learning environments, IRT behaviors, and academic achievement in science. The descriptive results revealed that the participants had above average SA in science, IRT level, CLEP, and intrinsic MLS. In addition, according to the path analysis results, while gifted students’ CLEP and MLS predicted their IRT behaviors and SA in science, their IRT behaviors predicted their SA. The findings of this study on motivation supported the literature on the gifted students.

In the study, it was found that eight-grade gifted students had high intrinsic MLS. Intrinsic motivation is a critical element of creativity and persistent learning (Ryan and Stiller, 1991). Specifically, gifted students are highly motivated to complete a duty from the beginning to the end. Renzulli (1976) describes being gifted as having SA, creativity, and motivation above average. Similarly, Davashigil (2004) discusses advanced intellectual ability, special talents in various fields, sensitivity, creativity, and high motivation as characteristics that distinguish gifted students from their peers. According to the studies, gifted students are more motivated to learn compared with their peers (Davis and Rimm, 1989; Köksal, 2013a; Yang et al., 2014).

Another finding of the study is related to gifted students’ CLEP. The participants considered the learning environment as full of constructive activities, where the responsibility was not only on the classroom teacher but also on the students and where active participation of all learners takes place. In a similar study, Rita and Martin-Dunlop (2011) found that while students prefer more constructive learning environment than their current learning environments, gifted students’ perception about their current learning environments was more positive than the other students’ perceptions. This finding may be related to the structure of science lessons. Science lessons require scientific
process skills, engineering design skills, and life skills to be actively used in teaching processes (MONE, 2016). Gifted students have the potential to perform high-level abilities in science (Soares, 2016). In other words, gifted students can manage to complete many active learning activities from designing and conducting an experiment to designing artifacts to solve high-level problems. In the literature, there exist studies reporting that students with or without gifted diagnosis generally regard science learning environments as constructive learning environment (Eroğlu et al., 2015; Kim et al., 2000; Özkal et al., 2009). On the other hand, high-level motivation, which is included in the definition of giftedness (Renzulli, 1986), may trigger positive perceptions about learning environment. There is a high correlation between gifted students’ CLEP and their motivation (Akkanat and Gökdere, 2018; Lüftenegger et al., 2015). In this study, the data were collected from gifted students registered to SAC. Since learning activities in SAC are based on project-based learning, problems solving, and learning by experimentation, which are the elements of constructive learning, students are exposed to constructive learning in these centers (MONE, 2016).

In the study, it was found that eight-grade gifted students’ level of IRT behavior was above average. More specifically, gifted students tend to share their thoughts in science even though they are not sure about correctness of their thoughts, gain new knowledge even though they may make mistakes, and talk about their solutions even though other students’ may have negative thoughts about themselves. Science courses require students to take intellectual risks. Actually, students have to uncover some information they do not know through experiments, discuss their thoughts related to a scientific problem without any concern, and ask questions and defend themselves about an association between the phenomena they observe and scientific concepts that they cannot make sense. According to Soares (2016), gifted students take intellectual risks by focusing on various solutions while solving a scientific problem and questioning the phenomenon related to the problem. National studies also reveal that gifted students’ IRT levels are above average (Akdağ et al., 2016; Akdağ and Köksal, 2016). The emphasis on enrichment, challenge, and independent study in teaching in SAC may also be an important explanation for this situation (MONE, 2016).

Another critical finding of the study is that gifted students’ SA levels were above average. A similar result was found in a study conducted by Ertekin (2017). This result may be related to gifted students’ ability to employ various cognitive skills special to science including establishing association between the scientific concepts and observed phenomenon, understanding scientific models and theories, and creating patterns related to scientific data without unknown connections (Gilbert and Newberry, 2007). According to Gould et al. (2003), gifted students can understand scientific concepts in depth by knowing different facts in science compared to their peers. In short, it may be concluded that eight grade-gifted students are successful in science courses.

One of the inferential results of the study is that eight-grade gifted students’ CLEP directly influenced their IRT behaviors and indirectly influence SA via IRT behaviors. In other words, the fact that gifted students consider science-learning environments as constructive learning environment positively affects their IRT behaviors. According to Byrnes (1998), IRT is a critical factor in active learning. More specifically, constructive learning environments are based on the idea that students may make mistakes; however, those mistakes should not be underestimated or should not be avoided (Jonassen et al., 1999). Furthermore, constructive learning environments require students to produce knowledge by themselves rather than getting it directly from instructors, which means that students need to reach knowledge by using the scientific process skills in science education. Students’ production of knowledge is based on the principle of making an effort to reach information even they are not sure of the result. In the literature, there is no study directly examining the association between IRT behaviors and science-learning environments. Constructivist learning approach is related with problem-based learning, active learning, and cooperative teaching strategies (Wilson, 1996). Weingrad (1998) stated that students take intellectual risks in learning environments in which they use reasoning processes, which shed light on the relationship between IRT and constructive learning environments. In a similar study, Köksal and Köseoğlu (2019) put strong emphasis on conditions of taking intellectual risks. Çakır (2017) also revealed that students who were exposed to constructive learning environments showed more IRT behaviors compared with their peers who were exposed to traditional learning environments. In short, conducting science activities with gifted students in constructive environments will positively affect their IRT behavior.

There was no direct association between students’ CLEP and their SA in science. However, some studies revealed positive effect of learning environment perceptions on SA in science for students with and without gifted diagnosis (Gautam, 2020; Ural and Bümen, 2016; Rita and Martin-Dunlop, 2011). On the other hand, Cairns and Areepattammil (2017) found a negative effect of inquiry-based learning, which is based on constructivist approach, on students’ SA in science. Furthermore, Cairns (2019) conducted a study using PISA 2015 dataset and found that students who constantly conduct science experiments had lower SA compared with students who occasionally conduct science experiments. According to these results, it may be concluded that CLEP does not always affect students’ SA positively. On the other hand, in this study, it was found that students’ perceptions indirectly influenced students’ achievement through their IRT behaviors. In inquiry-based learning environments, students’ freedom to ask questions and answer them is considered as a critical factor that increases the quality of teaching (Sadah and Zion, 2009; Windschitl, 2002). For example, in the directed inquiry approach, teachers provide the problem to students and students have full responsibility to find out the solution (Windschitl,
students’ intrinsic MLS had a high correlation with their SA, the significance level of this association is quite low for male students. Furthermore, Yenice et al. (2012) found that middle school students’ MLS decreases as their grade level increases. In this particular study, the participants consisted of eight-grade gifted students and the number of male students in the sample group was approximately twice as high as the number of female students. This limitation may be the reason for the insignificant association between motivation and SA. On the other hand, there was an indirect effect of motivation on SA through IRT behaviors, which implies that IRT is a motivational factor in science learning. Therefore, to increase eight-grade gifted students’ MLS, their IRT behaviors must be taken into account.

Another critical finding of the current study is the association between IRT behaviors and SA. Students’ IRT behaviors positively predicted their SA in science. This means that when gifted students are willing to answer questions even though they do not know the answer, design and conduct experiments even though they do not know the results, design new and original products without worrying about criticisms from their classmates, and share their thoughts about others’ products and argumentations, their SA in science increases. This argumentation is supported by the other studies as well. Özbay (2016) examined the association among middle school students’ SA in science, scientific epistemological beliefs, and IRT behaviors and found that IRT behaviors positively predicted students’ SA. In another study, Çakır and Yaman (2015) investigated middle school students’ IRT skills, metacognitive awareness, and SA and found a moderate level association between students’ IRT skills and SA in science. Deveci and Aydin (2018) found similar results. In addition, Tay et al. (2009) examined gifted students’ problem solving skills and IRT behaviors. They revealed that problem solving skills, which is critical for SA in science (Durgun and Önder, 2019), had a positive association with IRT behaviors.

Due to its nature, science is a field that contain scientific process skills including observations, hypothesis, and determining variables and conducting experiments. Undoubtedly, science has the power to develop countries’ economy by providing technological innovations that facilitate human life and making important inventions with the effort to understand nature, people, and the universe. Due to their high-level thinking skills, gifted individuals are key people for the development of a country since they have the potential to undertake very important innovations and inventions in the field of science. Therefore, it is critical to determine the factors that affect those students’ SA and the associations among those factors. In this context, this study aimed to examine eight grade-gifted students’ MLS, perceptions about learning environment, IRT behaviors, and academic achievement through path analysis. The previous studies were conducted by recruiting participants other than gifted students and performing basic level analysis with a few variables. Compared to other studies, this study differs by using path analysis and considering the predictability of four variables on a model basis. The findings revealed
positive associations among gifted students’ MLS, IRT behaviors, constructivist learning environment perceptions, and academic achievement in science. More specifically, gifted students’ MLS and CLEP directly affect their IRT behaviors and indirectly affect their SA. Based on the findings, it is concluded that eight-grade gifted students should be exposed to science learning environments in which they feel free to share their thoughts, active learning takes place, and activities that increase students’ intrinsic motivation are employed. This may advance gifted students’ IRT skills and, as a result, increase their SA in science. In the literature, there is no study that examines the explanatory relationships among gifted students’ CLEP, MLS, IRT skills, and SA in science at the same time. In addition, this study provides data for meta-analysis studies and observation-based research in the future. Since the eighth grade is the grade level in which decreases are observed in students’ motivation and attitude levels, the findings of this study also contribute to researchers focusing on the discussed problems. Moreover, the study also provides suggestions for effective and cognitive components that should be taken into account from the design of science activities carried out in SACs to the implementation and evaluation process.

LIMITATIONS

Considering the findings, limitations of the study, and difficulties encountered during the research, the following suggestions are offered. The first limitation is related to the variables. Although the ability to take intellectual risks for science learning is an important variable that affects SA, there may exist other affective and cognitive variables that may be related to this variable. In the future studies, it is critical to develop models that explain the association among those variables. The second limitation is that the participants of this study consisted of only eight grade-gifted students. The associations among the variables need to be examined for the other gifted students from different age groups and the students without gifted diagnosis. Another limitation is that this study was designed based on quantitative research method. Future research must be considering employing other data collection techniques, which are part of qualitative or mixed research method. The next limitation is related to motivation and its other dimensions. Intrinsic motivation was taken into account as a dimension in this particular study. Motivation for learning science includes many different dimensions such as internal control. More research is needed by considering other motivational factors as well. Another limitation is that a hypothetical model was created in order to examine the relationship among science learning environment perception, motivation for learning science, IRT, and academic achievement in science variables. In the future, experimental studies may be conducted on the effects of the mentioned variables on IRT behaviors and SA. Finally, in the study, a direct relationship was not found between the intrinsic motivation toward science and the science learning environment perception variables and the SA in science variable. Therefore, this study must be replicated with a larger sample group.

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