Effect of the Predict-Observe-Explain (POE) Strategy on Achievement in Science Education: A Meta-Analysis Study

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Abstract: Predict-observe-explain (POE) is a strategy that has been used in science education for several decades. It is important to determine how effective this strategy is, especially when used in the constructing scientific concepts. In this study, the effect of the POE strategy on students' science achievement was examined in a meta-analysis. Databases were searched using specific keywords and 35 studies (6 theses and 29 articles) that met the inclusion criteria were found. Hedges' g and the random effects model were used to calculate effect sizes. As a result, the average effect size (g=0.979, 95% CI: 0.771-1.188, p<0.001) was found to be high. The POE strategy was found to increase students' science achievement with a high-level effect. The differences between studies were determined by analog to the ANOVA and meta-regression moderator analyses. While the effect of POE on students' science achievement differed significantly according to the moderators of field of science and level of education, it did not differ significantly according to the moderator of implementation type. The moderators of implementation time and year of publication were similarly not found to be significant predictors of the effect of POE on science achievement. Based on these results, suggestions are made for researchers and practitioners.

Keywords: Predict-observe-explain, science education, academic achievement, meta-analysis

Fen Eğitiminde Tahmin-Gözlem-Açıkla Stratejisinin Akademik Başarıya Etkisi: Bir Meta Analiz Çalışması

Öz: Tahmin gözlem açıklama (TGA) uzun bir süredir fen eğitiminde kullanılan bir stratejidir. Özellikle fen alanındaki kavramların yapılandırıldığı deneylerde kullanılan bu stratejinin ne kadar etkili olduğunun belirlenmesi önemlidir. Bu araştırmada TGA stratejisinin öğrencilerin fen başarısı üzerindeki etkisi metaanaliz yöntemiyle incelenmiştir. Belirlenen anahtar kelimeler ile veri tabanlarında çeşitli taramalar gerçekleştirilmiş, dâhil edilme kriterlerini sağlayan 35 çalışma (6 tez, 29 makale) elde edilmiştir. Etki büyüklüklerinin hesaplanmasında Hedges's g ve analizlerde rastgele etkiler modeli tercih edilmiştir. Analiz sonucunda ortalama etki büyüklüğü (g=0.979, 95% CI:0.771-1.188, p<0.001) sınıflamaya göre yüksek düzeyde etkilemektedir. Çalışmalar arasındaki farklılıklar analog ANOVA ve meta-regresyon analizleri ile belirlenmiştir. TGA stratejisinin öğrencilerin fen başarısı üzerindeki etkisi bilim alanı ve öğretim kademesi moderatörlerine göre anlamlı farklılıklar gösterirken, uygulanma şekli moderatörüne göre anlamlı bir farklılıklar gösterirken, uygulanma şekli moderatörüne göre anlamlı bir farklılıklar gösterirken, uygulanma şekli moderatörüne göre anlamlı bir aşarısı üzerindeki etkisi bilim alanı ve öğretim kademesi uzerindeki etkisinin anlamlı birer yordayıcıları olmadığı ortaya koyulmuştur. Bu sonuçlardan yola çıkarak araştırmacı ve uygulayıcılara bazı önerilerde bulunulmuştur.

Anahtar kelimeler: Tahmin-gözlem-açıkla, fen eğitimi, akademik başarı, meta-analiz

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Introduction

Following the transition from the classical approach to the constructivist approach in science education, student-centered learning methods and techniques have come to the fore. In constructivist learning, the new information to be taught is built onto the existing knowledge of the students (Matthews, 2002). Constructing concepts interactively in a social context is particularly important in the teaching of science. In classroom environments where students are active, expressing their ideas and comparing their own ideas with those of their peers enables them to construct knowledge (Kearney, 2004). Predict-observe-explain (POE) is one of the studentcentered strategies under the umbrella of inquiry-based learning, which is in harmony with social constructivism and allows students to become responsible for their own learning. The POE strategy is a further development of a technique implemented at the University of Pittsburgh in the 1980s, known as demonstration, observation, and explanation (DOE) (Kearney, 2004). POE, which is based on classical research procedures, essentially includes the steps of forming a hypothesis, expressing the justifications for that hypothesis, collecting data to test the hypothesis, and discussing the results (Kearney, 2004; White, 1988). In POE, students first make a justified prediction before an experiment or activity to be carried out, and then they make observations about the experiment or activity. Finally, they explain whether there is a difference between their predictions and observations (White, 1988; White & Gunstone, 1992). Closed-ended "cookbook" laboratory experiments, in which students are given the steps to be followed and the aim is to verify information, do not facilitate in-depth learning and only allow students to perform the assigned task (Hofstein & Lunetta, 2004). In contrast to closed-ended experiments, POE is particularly used in experiments in which scientific concepts are constructed (Bulunuz & Bulunuz, 2017; Erdem Özcan & Uyanık, 2022).

The POE strategy, has been utilized in science education for several decades (Gunstone & White, 1981; White & Gunstone, 1992). Many experimental studies have been conducted on the effects of the POE strategy on science achievement. Experimental studies have been carried out particularly intensively in recent years. However, these studies differ from each other due to the different independent variables they include and their results do not seem to be consistent with each other (e.g., Çalış & Özkan, 2022; Tetik, 2019). Conducting a meta-analysis study by combining the results of those previous experimental studies will help in understanding the general results. In the literature, there have been two meta-analysis studies of the POE strategy to date. In the metaanalysis study performed by Gustina et al. (2023), 70 studies conducted in the fields of science and mathematics between 2013 and 2022 were included in the analysis. The researchers found that POE is more effective on learning outcomes than critical thinking skills, that it is highly effective at the senior high school level while it has rather low effects at the elementary school level, that it is more effective in mathematics than in science, and that the results of studies conducted between 2013 and 2017 were more effective than those of studies conducted between 2018 and 2022. In the second meta-analysis, Nurshafara (2022) investigated the effects of POE on physics learning outcomes. The results showed that POE was effective in the cognitive and affective domains at the junior and senior high school levels in studies conducted in Sumatra, Indonesia. Meta-analysis studies on the POE strategy (Gustina et al., 2023; Nurshafara, 2022) have indicated the need for more comprehensive meta-analysis. In recent years, especially in studies conducted on the effects of the POE strategy on academic achievement in science education, some variables have been common across different studies. In these studies, the POE strategy has been implemented alone or together with the support of technology; in different fields of science such as physics, chemistry,

biology, astronomy, and geology; at various educational levels such as primary school, middle school, high school, and university; and for periods of time ranging from a week to a semester. The existence of these differences among studies is one of the most important reasons for conducting the present meta-analysis study. While expanding further on the results of studies conducted to date, it is important to determine the effectiveness of the POE strategy, which is particularly used when scientific concepts are being constructed, for science achievement and to analyze its effects while taking into account different moderators such as field of science, implementation type, level of education, implementation time, and publication year. Examining the effects of these moderating variables on academic achievement may help produce a general perspective while providing ideas for researchers, teachers, curriculum developers, and policymakers. Furthermore, this meta-analysis may contribute to the development of new course designs and the improvement of current courses. In this respect, answers were sought to the following questions in the present study:

(1) What is the effect of the POE strategy on students' science achievement?

(2) Does the effect of the POE strategy on students' science achievement differ significantly according to the field of science, implementation type and level of education?

(3) Are the implementation time and year of publication significant predictors of the effect of the POE strategy on students' science achievement?

POE Strategy

The POE strategy consists of three stages: prediction, observation, and explanation. In the introductory part of a lesson plan, students' attention is drawn to the topic to be covered, they are motivated, and their prior knowledge is identified. This is also the case in the first stage of POE, namely prediction (P). In this stage, students are asked to make justified predictions about the topic to be explained, an experiment, or an event (White & Gunstone, 1992). Students' ideas, beliefs, misconceptions, and incomplete or erroneous knowledge can be identified while collecting their justified predictions (White & Gunstone, 1992). What is important in the prediction stage is justification. Making predictions without justification is similar to not showing the connections in concept maps. In such a case, "showing that you understand," which is the most important feature of POE, is not taken into account. It is important to ask open-ended questions appropriate to the level of the student in this stage, where justifications can be expressed (Kearney, 2004). Encouraging students to document their predictions by writing them down will also improve their written communication skills (Atasoy, 2004; Coe, 1993; Kearney, 2002, 2004; Ross & Munby, 1991). In the prediction stage, students can be asked open-ended questions that do not limit their thoughts or choices (Atasoy, 2004). Özçelik (2019) supported the prediction stage with concept cartoons reflecting more than one opinion. In the observation (O) stage, students conduct individual or group research or experiments about their predictions. Since the observation stage requires research, a lack of material may hinder the process (Kearney, 2004). However, the observation stage can be supported by instructional technologies such as animation, simulation, and video. Students can take notes about their observations and record data. Students' justified predictions may be correct or incorrect (White & Gunstone, 1992; Tao & Gunstone, 1997). Since observations are theory-laden, it is important to remember that students' observations are affected by their knowledge and the predictions they make (Kearney, 2002; Liew & Treagust, 1998; White & Gunstone, 1992; Yaman, Ayas, & Çalık, 2019). If there are contradictions between students' predictions and observations, they are expected to resolve them in the explanation (E) stage, where they have the opportunity to revise their ideas. Based on the students' explanations, the teacher can

conduct diagnostic and formative assessments. In this sense, the POE strategy also provides an opportunity for teachers to get to know their students (Kearney, 2002, 2004; Yaman, Ayas, & Çalık, 2019).

Moderator Variables

POE strategy has been widely implemented in science fields such as physics (Akpınar, 2014; Chen et al., 2020; Chen, 2022), chemistry (Acar Şeşen, 2013; Coştu, 2008; Yaman & Ayas, 2015), biology (Lucas et al., 2022), geology (James et al., 2022) and astronomy (Hsiao et al., 2017). In addition, it has also been used in the fields of mathematics (Yang & Chen, 2023) and psychology (Guerrero et al., 2022). In the process of implementing the POE strategy, some field of science such as physics, chemistry, and biology allow for direct observations and applications, while astronomy and earth science allow for indirect observations and applications. Therefore, determining in which field of science the use of the POE strategy is more effective on academic achievement has been seen as an important moderator variable.

In the literature, the POE strategy has been used both on its own (Erdem Özcan & Uyanık, 2022; Furgani et al., 2018; Karamustafaoğlu & Mamlok-Naaman, 2015) and with the support of technology (Akpınar, 2014; Chen et al., 2020; Fateen, 2020; Nyirahabimana et al., 2023; Yaman & Avas, 2015). With the POE strategy, technology can be used as a replacement for difficult, costly, time-consuming, and dangerous experiments (Kearney, 2004). All or some of the stages of the POE strategy were supported by technology in previous studies. In that way, it was aimed to make abstract and complex scientific concepts concrete and comprehensible. For instance, Tüysüz and Özdemir (2024) implemented the traditional POE strategy in one of two groups of 9th graders taking a physics course and the POE strategy supported by computer simulations in the other group. The groups had similar outcomes for the prediction and observation stages. However, in the classroom where traditional POE was implemented, the teacher made explanations about the observations using visual elements such as drawings, pictures, and analogies, while with the POE strategy supported by computer simulations, the teacher acted as a guide in the explanation stage, enabling students to benefit from the visual and interactive features of computer simulations. It was concluded that the POE strategy enriched with simulations was more effective than traditional POE in increasing student achievement. Yaman and Ayas (2015) incorporated digital video clips, simulations, and animations into the POE strategy for lessons on acids and bases at the high school level. They used animations and simulations to eliminate students' misconceptions. Macroscopic events were observed via video clips and microscopic events were observed via simulations. Akpinar (2014) investigated the effects of animation-supported POE and normal instruction on the learning of concepts in teaching about electricity in elementary school. For both approaches, experiments were conducted on the subject of static electricity. Subsequently, in the experimental group, dynamic and interactive animations were implemented together with the POE strategy. The prediction and observation stages of POE were supported with animations. The results suggested that the students in the experimental group, which was taught using POE supported by interactive animations, learned concepts related to electricity better. Chen et al. (2020) concluded that supporting game-based learning with POE in middle school physics contributed positively to the students' conceptual understanding. In the implementation process, the observation stage of the POE strategy was supported with games. As previously mentioned, there are studies in the literature confirming that the POE strategy is more effective when it is supported by technology. Therefore, the type of POE implementation (only POE or technology-supported POE) was also considered as

an important variable that could affect academic achievement and was accordingly selected as one of the moderating variables.

POE strategy has been used at the preschool (Hsu et al., 2011), elementary school (Erdem Özcan & Uyanık, 2022), middle school (Chen et al., 2020; Chen, 2022), high school (Çinici et al., 2011; Yaman & Ayas, 2015), and university (James et al., 2022; Lucas et al., 2022) levels, especially in the teaching and learning of scientific concepts. In one of the previously conducted meta-analysis studies, it was concluded that the POE strategy had a very limited effect at the elementary school level (Gustina et al., 2023), and in another meta-analysis, the POE strategy was shown to be effective at the middle school and high school levels (Nurshafara, 2022). Although researchers have stated that the POE strategy is suitable for elementary school students (Erdem & Özcan Uyanık, 2022; Palmer, 1995), the differences in the effects of this strategy according to education level remain to be clarified. Therefore, level of education was considered as a moderating variable in the present meta-analysis study.

In experimental studies where the POE strategy was used, implementation time ranged from one week (e.g., Karpin et al., 2014) to a period of 12 weeks (Ayvacı & Durmuş, 2016). Increasing the implementation time may increase academic achievement (White, 1988). Therefore, intervention time was considered as one of the moderating variables. Another moderator examined within the scope of this research is the publication year. The widespread use of the constructivist approach and technology in education after the 2000s, and distance education during the COVID-19 period have made the predictive value of the publication year on the effect of the POE strategy on academic achievement a matter of curiosity. However, since the moderators of intervention time and publication year were considered as continuous variables in the present study, no categorizations were performed for these variables while analyzing the effect of the POE strategy on academic achievement (Hedges & Olkin, 1985).

Method

In this study, the effect of POE on students' science achievement was analyzed by metaanalysis. Meta-analysis is a method that combines statistical findings obtained with different calculations (Borenstein et al., 2021). In the selection and analysis of the studies, a 3-step procedure was followed, similar to the methods suggested by Card (2012), Field and Gillett (2010), and Glass et al. (1981). These steps were the selection of studies, the application of a coding strategy, and data analysis.

Selection of Studies

The search for relevant studies was carried out in two stages. In the first stage, the keywords "predict," "observe," and "explain" were used, and in the second stage, the keywords "prediction," "observation," and "explanation" were used. The Web of Science (WoS), Scopus, and Education Resources Information Center (ERIC) databases were utilized to search for articles and the ProQuest Dissertations & Theses Global (PQDT) database was used to search for theses.

The criteria applied in selecting the studies included in the meta-analysis were as follows: 1) studies related to POE and scientific achievement, 2) open access full-text studies published between 2003 and 2023, 3) studies published in English or Turkish, 4) studies published in thesis or article format, 5) studies that included POE implementation, and 6) studies that had sufficient data for the calculation of effect size.

Due to the different characteristics of the databases, a standardized search was not performed and different Boolean operators were used in the process. The searching was completed on January 10, 2024. The keywords identified in the databases and the way in which the searching was conducted according to the inclusion criteria are explained in detail below:

• In the search section of the PQDT database, keywords were entered with the NOFT code (predict AND observe AND explain) in the first stage and with NOFT(prediction AND observation AND explanation) in the second stage, ensuring that they were found in all studies anywhere except the full text. All studies were filtered as full text, year range 2003-2023, and publication languages English and Turkish. As a result of these search, a total of 411 studies were identified, including 117 in the first stage and 294 in the second stage.

• In the simple search section of the WoS database, keywords were entered as predict AND observe AND explain in the first stage and prediction AND observation AND explanation in the second stage; while searching in the WoS Core Collection, editions SCI-E, SSCI, AHCI, and ESCI, the search field was specified as topic. All studies were filtered as publication years 2003-2023; document type article; WoS categories Education-Educational Research, Education-Scientific Disciplines, and Education-Special; and publication languages English and Turkish. As a result of these searches, a total of 225 studies were identified, including 190 in the first stage and 35 in the second stage.

• In the advanced search section of the Scopus database, keywords were entered with the TITLE-ABS-KEY code (predict AND observe AND explain) in the first stage and the TITLE-ABS-KEY code (prediction AND observation AND explanation) in the second stage. All studies were filtered as year range 2003-2023, subject area Social Sciences, document type article, and publication languages English and Turkish. As a result of these searches, a total of 189 studies were identified, including 101 in the first stage and 88 in the second stage.

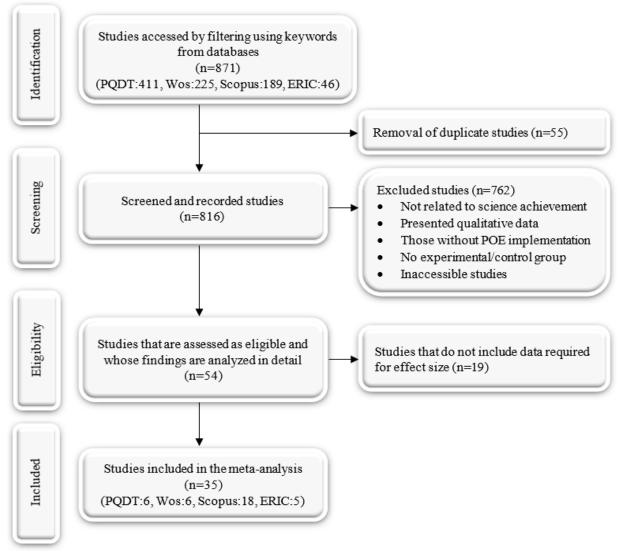
• In the simple search section of the ERIC database, keywords were entered as predict AND observe AND explain in the first stage and prediction AND observation AND explanation in the second stage. All studies were filtered as full text, publication dates 2003-2023, and publication type journal articles. As a result of these searches, a total of 46 studies were identified, including 30 in the first stage and 16 in the second stage.

The reason for conducting the searches using only the keywords "predict," "prediction," "observe," "observation," "explain," and "explanation," excluding words related to science achievement, was to obtain as many studies as possible related to the research problem and avoid overlooking any relevant studies. During the searches, publications produced from conference papers, project reports, and theses were frequently encountered. Due to their lack of data, conference papers and project reports, publications produced from theses, and studies that did not meet any of the inclusion criteria were excluded. In cases where the same study was found in more than one database, the study was taken from the first database in which it was found and subsequent appearances of it were ignored. A detailed description of the review process is critical in a meta-analysis for the transparency, reliability, and reproducibility of the research (Moher et al., 2009; Sarkis-Onofre et al., 2021).

The review process of the present study was conducted in accordance with the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) guidelines (Page et al., 2021). The PRISMA flowchart is shown in Figure 1.

Figure 1

Screening Process PRISMA Flowchart



As can be seen in Figure 2, a total of 438 studies (PQDT: 117, WoS: 190, Scopus: 101, ERIC: 30) were found at the end of the first stage and 433 studies (PQDT: 294, WoS: 35, Scopus: 88, ERIC: 16) were found at the end of the second stage of searches conducted with the keywords, limitations, and inclusion criteria described above, totaling 871 studies. After the removal of 55 duplicate studies from different databases, 816 studies were examined according to their titles, abstracts, and methods. It was decided to exclude studies that were not related to science achievement, that presented qualitative data, that did not include POE implementation, that lacked experimental/control groups, or that were inaccessible. After excluding 762 studies, the remaining 54 studies were evaluated as suitable for the meta-analysis and their findings were examined in detail for effect size. It was decided to exclude 19 studies that did not have sufficient data for the calculation of effect size. Finally, a total of 35 studies, including 6 from PQDT, 6 from WoS, 18 from Scopus, and 5 from ERIC, were included in the meta-analysis.

Coding Strategy

A coding form was created by the author to be used in the coding process. This form included the following information about the studies: 1) study number, 2) author(s) of the study, 3) publication year of the study, 4) title of the study, 5) field of science for which POE was implemented, 6) implementation type of POE, 7) level of education, 8) implementation time of POE, and 9) sample numbers, arithmetic averages, standard deviations, and other data for the calculation of effect size.

The coding was carried out by two coders, one of whom was an expert in meta-analysis and the other the author herself. The coders first read the titles and abstracts of the studies that had been obtained and then read the contents of the studies. Each of the studies that were deemed appropriate by the coders were recorded separately on their computers using the coding form. The recorded studies were compared from time to time, and inconsistencies, if any, were resolved through a mutual exchange of views. The coding was continued until full agreement was achieved between the coders. Cohen's kappa coefficient was used to determine intercoder reliability. There was no disagreement between the coders until the last stage of the review process, so Cohen's kappa was not calculated. In the last stage, 4 of the 54 studies whose findings were analyzed had disagreement and 50 had consensus. The coders re-evaluated whether the studies for which there was disagreement had the necessary data for the calculation of effect size. It was decided to exclude 2 of the 4 studies for which there was disagreement due to missing data and to include those of Özçelik (2019) and Hsiao et al. (2017) in the meta-analysis. Cohen's kappa coefficient was calculated according to the studies with agreement and disagreement and found to be 0.839. This value is greater than 0.80, indicating a high level of agreement between the coders (Cohen et al., 2002).

Since 3 of the 35 studies included multiple experimental groups, it was decided to code the studies of Cinici et al. (2011) and Özçelik (2019) as two independent studies and that of Hsiao et al. (2017) as three independent studies. The characteristics of the studies included in the metaanalysis are given in Table 1. As can be seen in Table 1, the publication years of the studies ranged between 2008 and 2023 despite the broader range limit used in the inclusion criteria. It can be concluded that such studies have intensified in recent years. The studies were categorized according to the fields of astronomy, physics, chemistry, biology, chemistry-physics, and geology. In some studies, POE was applied together with digital games (Chen et al., 2020; Hsu et al., 2011), simulations (Chen et al., 2013; Fateen, 2020), animations (Akpınar, 2014), multimedia technologies (Nyirahabimana et al., 2023), computers (Karslı Baydere, 2021; Yaman & Ayas, 2015), and websites, videos, and flash animations (Hsiao et al., 2017). For this reason, the studies were categorized into two groups according to the implementation type of POE: studies in which POE was used alone and studies in which it was supported by technology. The samples of the studies were grouped according to the level of education: preschool, elementary school (grades 1-4), middle school (grades 5-8), high school (grades 9-12), and university (UNESCO, 2012). The implementation time of POE was determined in weeks as stated in the studies. However, in the studies of Bulunuz and Bulunuz (2017), Chen (2022), Coștu (2008), Furqani et al. (2018), Harman and Yenikalaycı (2022), Hsu et al. (2011), James et al. (2022), Lucas et al. (2022), and Zhao et al. (2021), the implementation time was not clearly stated in weeks. As can be seen in Table 1, that information was accordingly left blank during coding. Although not specified in the table, the sample sizes of the studies varied between 12 and 365. The total sample size of the combined studies was 2840.

Table 1

Characteristics of the Studies Included in the Meta-Analysis

| ID | Author(s), year | Field of science | Implementation type of POE | Level of education | Implementation time (week) |
|----|---------------------------------------|---------------------|----------------------------|--------------------|----------------------------|
| 1 | Akpınar, 2014 | Physics | Technology supported POE | Primary school | 10 |
| 2 | Ayvacı & Durmuş, 2016 | Physics | POE | University | 12 |
| 3 | Banawi et al., 2019 | Chemistry | POE | University | 4 |
| 4 | Bilen & Köse, 2012 | Biology | POE | University | 4 |
| 5 | Bolat & Karamustafaoğlu, 2021 | Physics | POE | Middle school | 3 |
| 6 | Bulunuz & Bulunuz, 2017 | Physics | POE | High school | |
| 7 | Chen et al., 2013 | Physics | Technology supported POE | University | 1 |
| 8 | Chen et al., 2020 | Physics | Technology supported POE | Middle school | 2 |
| 9 | Chen, 2022 | Physics | Technology supported POE | Middle school | |
| 10 | Coștu, 2008 | Chemistry | POE | High school | |
| 11 | Coștu, 2021 | Chemistry & physics | POE | University | 9 |
| 12 | Çalış & Özkan, 2022 | Physics | POE | Middle school | 6 |
| 13 | Çıngıl Barış, 2022 | Biology | POE | University | 6 |
| 14 | Çinici et al., 2011_1 | Biology | POE | High school | 4 |
| 15 | Çinici et al., 2011_2 | Biology | POE | High school | 4 |
| 16 | Erdem Özcan & Uyanık, 2022 | Chemistry | POE | Primary school | 8 |
| 17 | Fateen, 2020 | Physics | Technology supported POE | University | 1 |
| 18 | Furqani et al., 2018 | Physics | POE | Middle school | |
| 19 | Harman & Yenikalaycı, 2022 | Physics | POE | University | |
| 20 | Hsiao et al., 2017_1 | Astronomy | POE | Primary school | 5 |
| 21 | Hsiao et al., 2017_2 | Astronomy | Technology supported POE | Primary school | 5 |
| 22 | Hsiao et al., 2017_3 | Astronomy | POE | Primary school | 5 |
| 23 | Hsu et al., 2011 | Physics | Technology supported POE | Preschool | |
| 24 | James et al., 2022 | Geology | Technology supported POE | University | |
| 25 | Kahraman, 2023 | Chemistry | Technology supported POE | University | 4 |
| 26 | Karamustafaoğlu & Mamlok-Naaman, 2015 | Chemistry | POE | University | 3 |
| 27 | Karpin et al., 2014 | Chemistry | POE | High school | 1 |
| 28 | Karslı Baydere, 2021 | Chemistry | Technology supported POE | Middle school | 2 |
| 29 | Kibirige et al., 2014 | Chemistry | POE | High school | 5 |
| 30 | Lucas et al., 2022 | Biology | Technology supported POE | University | |
| 31 | Nyirahabimana et al., 2023 | Physics | Technology supported POE | University | 6 |
| 32 | Özçelik, 2019_1 | Physics | POE | Middle school | 4 |
| 33 | Özçelik, 2019_2 | Physics | POE | Middle school | 4 |
| 34 | Özkan, 2022 | Physics | POE | Middle school | 5 |
| 35 | Tetik, 2019 | Chemistry | POE | High school | 4 |
| 36 | Tokay, 2022 | Chemistry | POE | High school | 4 |
| 37 | Wiyarsi et al., 2021 | Chemistry | POE | High school | 6 |
| 38 | Yaman & Ayas, 2015 | Chemistry | Technology supported POE | High school | 5 |
| 39 | Zhao et al., 2021 | Physics | POE | Middle school | |

Data Analysis

The free trial version of the Comprehensive Meta-Analysis (CMA) program was used for publication bias control, calculation of effect size values, heterogeneity testing, and moderator analyses (Borenstein et al., 2014). This program was preferred because it can easily combine and calculate data in different categories obtained from different studies.

Publication bias is the outcome of significant results being more likely to be published than statistically insignificant findings (Dickersin, 2005). This causes the calculated average effect size to be larger than the actual value (Card, 2012). Due to this problem, it is necessary to determine whether the studies included in a meta-analysis are impacted by publication bias or not. The methods used in this study for identifying publication bias were as follows: 1) the funnel plot method, 2) classic fail-safe N method, 3) Orwin's fail-safe N method, 4) Egger's regression test, and 5) Duval and Tweedie's trim-and-fill method.

When there is a study or studies with a sample size smaller than 20 among the papers included in a meta-analysis, Hedges' g can be used instead of Cohen's d in the calculation of effect sizes (Borenstein et al., 2021). In the analysis conducted for the present study, Hedges' g was preferred due to the presence of studies with sample sizes smaller than 20 (Bolat & Karamustafaoglu, 2021; Furqani et al., 2018; Yaman & Ayas, 2015). There are many classifications in the literature for effect sizes and levels. In this study, Thalheimer and Cook's (2002) classification was used. The ranges and levels of Hedges' g in this classification are as follows: $0.00 \le g < 0.15$, ignored; $0.15 \le g < 0.40$, low; $0.40 \le g < 0.75$, medium; $0.75 \le g < 1.10$, high; $1.10 \le g < 1.45$, very high; $1.45 \le g$, excellent.

Two models, the fixed effect model and the random effects model, are used to combine the effect sizes obtained from different studies. In meta-analyses, the average effect size will vary according to the selected model. The average effect size is calculated according to the random effects model if the effect sizes of the studies show a heterogeneous distribution and according to the fixed effect model if they do not (Cooper et al., 2009). In high-quality meta-analysis studies, the model being used is specified in advance (Borenstein et al., 2021). Furthermore, it is recommended to use the random effects model for meta-analysis studies (Field & Gillett, 2010). In this study, the average effect size was calculated according to the random effects model for reasons such as the populations of the studies not being the same and the data collection tools being different. This was also supported by heterogeneity test analysis results (Konstantopoulos & Hedges, 2019). Cochran's Q test was used to determine the difference between effect sizes and the I2 statistic was used to determine the level. A Cochran Q value greater than the value in the chisquare table (critical value) and a p significance value less than 0.05 indicate that effect sizes are heterogeneously distributed (Hedges & Olkin, 1985). An I2 statistic value greater than 50% indicates a sufficient level of heterogeneity, while a value greater than 75% indicates a high level of heterogeneity (Higgins et al., 2003). When there is a heterogeneous distribution among effect sizes, possible causes are identified through moderators (Rodriguez et al., 2023). In this study, possible reasons for heterogeneity were sought through both categorical and continuous moderators. Analog to the ANOVA (Hedges, 1982) was conducted for the categorical moderators of field of science, implementation type, and level of education, while meta-regression (Hedges & Olkin, 1985) was applied for the continuous moderators of publication year and implementation time.

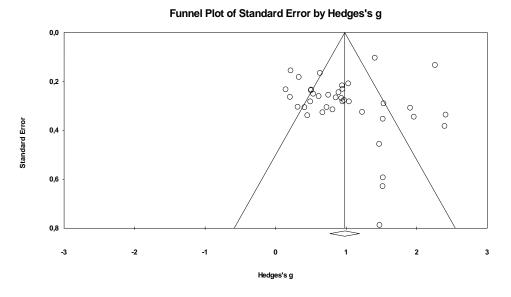
Findings

Publication Bias

Publication bias was investigated using the funnel plot, classic fail-safe N, Orwin's failsafe N, Egger's regression, and Duval and Tweedie's trim-and-fill methods. Figure 2 shows the distribution of effect sizes in a funnel plot.

Figure 2

Funnel Plot of Sstandard Error by Hedges's g



It can be seen in Figure 2 that most of the effect sizes are located within and on the upper side of the funnel plot. It can be argued that studies on the right or left side disrupt the symmetrical structure. The subjective interpretation of funnel plots necessitates additional methods to determine publication bias (Borenstein et al., 2021). Table 2 shows the publication bias findings based on the classic fail-safe N method.

Table 2

| z-value for observed studies | 23.362 |
|---|--------|
| p-value for observed studies | 0.000* |
| Alpha | 0.050 |
| Tails | 2 |
| z for alpha | 1.959 |
| Number of observed studies | 39 |
| Number of missing studies that would bring p-value to > alpha | 5503 |
| *p<0.001 | |

In classic fail-safe N analysis, the number of missing studies that need to be included in the analysis for the significance to change is specified. This number, referred to as the fail-safe N

(FSN) value, is 5503, as can be seen in Table 2. This number is greater than 10 more than 5 times the number of available studies (n=39), or 205. This can be interpreted as signifying a low probability of publication bias (Rosenthal, 1991). Similarly, Orwin's FSN value was found to be 3859. This value indicates the number of missing studies needed to reduce Hedges' g below a specified value (0.01). The fact that this number is also large and that it would be difficult to reach 3859 papers confirms the conclusion that the likelihood of publication bias is low. Table 3 shows the findings of Egger's regression test for publication bias.

Table 3

| Intercept-0.651Standard error1.145 |
|------------------------------------|
| Standard error 1.145 |
| |
| 95% lower limit (2-tailed) -2.973 |
| 95% upper limit (2-tailed) 1.669 |
| t-value 0.568 |
| df 37 |
| p-value (1-tailed) 0.286 |
| p-value (2-tailed) 0.572 |

Egger's Regression Intercept Ffindings of Publication Bias

Based on Egger's regression analysis, the null hypothesis of "there is no publication bias" was tested. As can be seen in Table 3, the 1-tailed p value (suggested) is 0.286 and the 2-tailed p value is 0.572. The result is not significant (t=0.568, p>0.05). Accordingly, it can be said that there is no publication bias (Egger et al., 1997). Table 4 shows the publication bias findings for Duval and Tweedie's trim-and-fill method (Duval & Tweedie, 2000).

Table 4

Duval and Tweedie's Trim and Fill Method Publication Bias Findings

| | Studies trimmed | Point estimate | Lower limit | Upper limit | Q value |
|-----------------|-----------------|----------------|-------------|-------------|---------|
| Observed values | | 0.979 | 0.771 | 1.188 | 259.418 |
| Adjusted values | 3 | 1.053 | 0.845 | 1.262 | 314.025 |

As shown in Table 4, the average effect size value calculated according to the random effects model is 0.979, and the average effect size value calculated with the 3 missing studies as described above is 1.053. The difference between the observed and adjusted average effect sizes (0.074) is less than 20%, indicating that there is no publication bias (Vevea et al., 2019).

Average Effect Size and Heterogeneity Test

For each of the included studies, Hedges' g, standard error, variance, 95% confidence interval lower limit and upper limit, z, and p values and the distribution of effect sizes are shown in Figure 3.

Figure 3

Forest Plot of Studies Included in Meta-Analysis

| Study name | Statistics for each study | | | | | each study Hedges's g and 95%Cl | | | | |
|-------------------------------|---------------------------|-------------------|----------|----------------|----------------|---------------------------------|---------|-----------------------|--|--|
| | Hedges's g | Standard error | Variance | Lower limit | Upper limit | Z-Value | p-Value | | | |
| Akpinar, 2014 | 0.972 | 0,277 | 0,077 | 0,429 | 1,515 | 3,509 | 0,000 | | | |
| Ayvaci & Durmus, 2016 | 0.726 | 0.306 | 0.094 | 0.126 | 1.326 | 2,371 | 0,018 | | | |
| Banawi et al., 2019 | 0.895 | 0,245 | 0.060 | 0,414 | 1,375 | 3,649 | 0,000 | | | |
| Bilen & Kose, 2012 | 0,506 | 0,235 | 0,055 | 0,046 | 0,966 | 2,158 | 0,031 | | | |
| Bolat & Karamustafaoglu, 2021 | 1,471 | 0,456 | 0,208 | 0,576 | 2,365 | 3,222 | 0,001 | | | |
| Bulunuz & Bulunuz, 2017 | 0.951 | 0,231 | 0.054 | 0.497 | 1,404 | 4,109 | 0,000 | | | |
| Calis & Ozkan, 2022 | 0,144 | 0,233 | 0.055 | -0,314 | 0.602 | 0,617 | 0,537 | | | |
| Chen et al., 2013 | 0,452 | 0,339 | 0,115 | -0,213 | 1,117 | 1,332 | 0,183 | | | |
| Chen et al., 2020 | 0.331 | 0,183 | 0.033 | -0.027 | 0.690 | 1,812 | 0,070 | | | |
| Chen, 2022 | 0.206 | 0.264 | 0.070 | -0,312 | 0.724 | 0.779 | 0,436 | | | |
| Cingil Baris, 2022 | 0,536 | 0,251 | 0,063 | 0,043 | 1,029 | 2,131 | 0,033 | | | |
| Cinici et al., 2011_1 | 0,934 | 0,269 | 0,072 | 0,408 | 1,461 | 3,476 | 0,001 | | | |
| Cinici et al., 2011_2 | 0.613 | 0.261 | 0.068 | 0.101 | 1.124 | 2.349 | 0.019 | | | |
| Costu, 2008 | 1,961 | 0,345 | 0,119 | 1,284 | 2,638 | 5,679 | 0,000 | | | |
| Costu, 2021 | 2,398 | 0,383 | 0.147 | 1,647 | 3,148 | 6,261 | 0,000 | | | |
| Erdem Ozcan & Uyanik, 2022 | 1,910 | 0,309 | 0,095 | 1,305 | 2,516 | 6,187 | 0,000 | | | |
| Fateen, 2020 | 1.032 | 0,209 | 0.044 | 0,622 | 1,442 | 4,935 | 0,000 | | | |
| Furgani et al., 2018 | 1,520 | 0.629 | 0.396 | 0,287 | 2,753 | 2,415 | 0,016 | | | |
| Harman & Yenikalayci, 2022 | 1,523 | 0,593 | 0,352 | 0,360 | 2,686 | 2,566 | 0,010 | | | |
| Hsiao et al., 2017_1 | 0,412 | | 0,094 | -0,189 | 1,012 | 1,344 | 0,179 | | | |
| Hsiao et al., 2017 2 | 0.950 | 0.283 | 0.080 | 0.396 | 1,504 | 3,360 | 0,001 | | | |
| Hsiao et al., 2017 3 | 0.855 | 0,267 | 0.071 | 0,332 | 1,377 | 3,206 | 0,001 | | | |
| Hsu et al., 2011 | 0,494 | 0,283 | 0,080 | -0,061 | 1,048 | 1,746 | 0,081 | | | |
| James et al., 2022 | 0,215 | 0,156 | 0,024 | -0,091 | 0,521 | 1,380 | 0,168 | | | |
| Kahraman, 2023 | 1,043 | 0,283 | 0,080 | 0,488 | 1,597 | 3,687 | 0,000 | | | |
| Karamustafaoglu & Mamlok-Naa | man, 201521 | 0,354 | 0,125 | 0,828 | 2,214 | 4,303 | 0,000 | | | |
| Karpin et al., 2014 | 1,229 | 0,326 | 0,106 | 0,590 | 1,867 | 3,771 | 0,000 | | | |
| Karsli Baydere, 2021 | 0,667 | 0,327 | 0,107 | 0,026 | 1,308 | 2,038 | 0,042 | | | |
| Kibirige et al., 2014 | 0,944 | 0,217 | 0,047 | 0,518 | 1,370 | 4,346 | 0,000 | | | |
| Lucas et al., 2022 | 1,409 | 0,104 | 0,011 | 1,205 | 1,613 | 13,510 | 0,000 | | | |
| Nyirahabimana et al., 2023 | 2,262 | 0,134 | 0,018 | 1,999 | 2,524 | 16,893 | 0,000 | | | |
| Ozcelik, 2019_1 | 0,315 | 0,305 | 0,093 | -0,283 | 0,912 | 1,033 | 0,302 | | | |
| Ozcelik, 2019_2 | 0,810 | 0,315 | 0,099 | 0,192 | 1,428 | 2,568 | 0,010 | | | |
| Ozkan, 2022 | 0,504 | 0,237 | 0,056 | 0,039 | 0,968 | 2,125 | 0,034 | | | |
| Tetik, 2019 | 2,413 | 0,337 | 0,113 | 1,753 | 3,073 | 7,164 | 0,000 | | | |
| Tokay, 2022 | 1,533 | 0,291 | 0,085 | 0,963 | 2,103 | 5,273 | 0,000 | | | |
| Wiyarsi et al., 2021 | 0,749 | 0,256 | 0,065 | 0,248 | 1,250 | 2,928 | 0,003 | | | |
| Yaman & Ayas, 2015 | 1,476 | 0,788 | 0,621 | -0,068 | 3,020 | 1,873 | 0,061 | | | |
| Zhao et al., 2021 | 0,631 | 0,166 | 0,028 | 0,305 | 0,957 | 3,791 | 0,000 | | | |
| | 0,979 | 0,106 | 0,011 | 0,771 | 1,188 | 9,205 | 0,000 | | | |
| | | | | | | | | -4,00 -2,00 0,00 2,00 | | |

As seen in Figure 3, the Hedges' g values obtained from the studies vary between 0.144 and 2.413. It was found that the Hedges' g value was highest for the study by Tetik (2019) at 2.413 and smallest for the study by Çalış and Özkan (2022) at 0.144. The effect size p values of the studies conducted by Çalış and Özkan (2022), Chen et al. (2020), Chen (2022), Hsiao et al. (2017), Hsu et al. (2011), James et al. (2022), Özçelik (2019), and Yaman and Ayas (2015) are not significant (p>0.05). All effect sizes are positive with different reliability intervals. According to Thalheimer and Cook's (2002) classification, there are 1 (2.6%) ignored, 4 (10.3%) low, 11 (28.2%) medium, 10 (25.6%) high, 3 (7.7%) very high, and 10 (25.6%) excellent effect sizes. The average effect size, related values, and heterogeneity test findings of the studies according to fixed and random effects models are given in Table 5.

Table 5

Average Effect Size and Relevant Values of the Studies Included in the Meta-Aanalysis, as well as Heterogeneity Test Findings

| | | Point | Standard | 95% CI | | | | | | | |
|----------|----|----------|----------|----------------|----------------|--------|--------|---------|----|--------|----------------|
| Model | k | estimate | error | Lower limit | Upper limit | Z | р | Q | df | р | I ² |
| Fixed | 39 | 0.999 | 0.039 | 0.923 | 1.075 | 25.725 | 0.000* | 259.418 | 38 | 0.000* | 85.352 |
| Random | 39 | 0.979 | 0.106 | 0.771 | 1.188 | 9.205 | 0.000* | - | | | |
| *p<0.001 | | | | | | | | | | | |

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As seen in Table 5, according to the fixed effect model, the average effect size is 0.999, the standard error is 0.039, the lower limit of the 95% confidence interval is 0.923, and the upper limit is 1.075 (z=25.725, p<0.001). According to the random effects model, the average effect size is 0.979, the standard error is 0.106, the lower limit of the 95% confidence interval is 0.771, and the upper limit is 1.188 (z=9.205, p<0.001). For both models, the average effect size value is at a high level according to Thalheimer and Cook's (2002) classification. This finding shows that the POE strategy has a high-level effect on students' science achievement according to both models.

When the heterogeneity test findings were analyzed, it was found that the p value was 0.000 and the Q value was 259.418. It is statistically significant that the calculated p value is less than 0.05 and the Q value is greater than the chi-square table value (53.383) at 38 degrees of freedom and a significance level of 0.05 (Q> χ^2 , p<0.05). Accordingly, the effect sizes show a heterogeneous distribution. The value of the I² statistic (85.352) is greater than 75%, which indicates a high level of heterogeneity (Higgins et al., 2003). It can be stated that approximately 85.35% of the variance between effect sizes is due to real differences in effect sizes and 14.65% is due to sampling error (Huedo-Medina et al., 2006; Li et al., 2021). Moderator analyses need to be conducted to explain such a high level of variation (Lipsey & Wilson, 2001).

Moderator Analyses

Analog to the ANOVA was conducted in line with the following question: "Does the effect of the POE strategy on students' science achievement differ significantly according to the field of science, implementation type, and level of education?" Analog to the ANOVA findings related to the moderators of field of science, implementation type of POE, and level of education are given in Table 6.

Table 6

| | | | _ | | 95% | 6 CI | | | | | |
|-------------------------------|-----------------------------|----|-------|-------|----------------|----------------|-------|--------|----------------|----|--------|
| Variables | Category | k | Ē | SE | Lower limit | Upper limit | Z | р | Q _B | df | р |
| | Astronomy | 3 | 0.760 | 0.164 | 0.439 | 1.081 | 4.638 | 0.000* | | | |
| | Biology | 5 | 0.825 | 0.231 | 0.372 | 1.278 | 3.570 | 0.000* | | | 0.000* |
| | Chemistry | 12 | 1.327 | 0.157 | 1.019 | 1.635 | 8.448 | 0.000* | | | |
| Field of science | Chemistry & physics | 1 | 2.398 | 0.383 | 1.647 | 3.148 | 6.261 | 0.000* | 42.466 | 5 | |
| | Geology | 1 | 0.215 | 0.156 | -0.091 | 0.521 | 1.380 | 0.168 | | | |
| | Physics | 17 | 0.815 | 0.187 | 0.448 | 1.183 | 4.353 | 0.000* | | | |
| I | POE | 26 | 1.019 | 0.111 | 0.802 | 1.236 | 9.188 | 0.000* | | | |
| Implementation type of POE | Technology supported POE | 13 | 0.875 | 0.215 | 0.453 | 1.296 | 4.068 | 0.000* | 0.056 | 1 | 0.551 |
| | High school | 10 | 1.229 | 0.172 | 0.891 | 1.566 | 7.143 | 0.000* | | | |
| Level of | Middle school | 10 | 0.515 | 0.105 | 0.308 | 0.721 | 4.878 | 0.000* | | | |
| education | Preschool | 1 | 0.494 | 0.283 | -0.061 | 1.048 | 1.746 | 0.081 | 17.497 | 4 | 0.002 |
| euucauoii | Primary school | 5 | 1.015 | 0.229 | 0.565 | 1.464 | 4.425 | 0.000* | | | |
| | University | 13 | 1.100 | 0.211 | 0.686 | 1.513 | 5.209 | 0.000* | | | |

Analog to the ANOVA Moderator Analysis Findings

*p<0.001

Table 6 shows that the p value is 0.000 and the Q_B value is 42.466. The fact that the p value is less than 0.05 and the Q_B value is greater than the critical value (11.070) indicates that there is a significant difference between the groups (Q_B> χ^2 , p<0.05). According to these results, the effect of

the POE strategy on students' science achievement differs significantly with respect to the moderator of field of science. The average effect size values are 0.760, 0.825, 1.327, 2.398, 0.215, and 0.815 for astronomy, biology, chemistry, chemistry-physics, geology, and physics, respectively. According to Thalheimer and Cook's (2002) classification, the POE strategy had a low effect on students' science achievement in geology; a high effect in astronomy, biology, and physics; a very high effect in chemistry; and an excellent effect in chemistry-physics. When the heterogeneity test findings for the implementation type of POE was analyzed, the p value was 0.551 and the Q_B value was 0.056. The fact that the p value is greater than 0.05 and the Q_B value is less than the critical value (3.841) indicates that there is no significant difference between the groups $(Q_B < \chi^2, p > 0.05)$. Accordingly, it can be concluded that different ways of implementing POE affected students' science achievement at similar levels. The average effect size values for POE and technology-supported POE were 1.019 and 0.875, respectively. Although there was no significant moderator, POE and technology-supported POE affected students' science achievement at a high level according to the classification. When the heterogeneity test findings for level of education were analyzed, it was found that the p value was 0.002 and the Q_B value was 17.497. The fact that the p value is less than 0.05 and the Q_B value is greater than the critical value (9.487) indicates that there is a significant difference between the groups ($Q_B > \chi^2$, p<0.05). Accordingly, the effect of the POE strategy on students' science achievement differs significantly according to the moderator of education level. The average effect size values for high school, middle school, preschool, elementary school, and university are 1.229, 0.515, 0.494, 1.015, and 1.100, respectively. The POE strategy had a very high effect on students' science achievement in high school and university, a high effect in elementary school, and a medium effect in middle school and preschool.

Meta-regression analyses were conducted in line with the following question: "Is the implementation time and year of publication a significant predictor of the effect of the POE strategy on students' science achievement?" The meta-regression analysis findings are given in Table 7.

Table 7

| Variable | k | В | SE | 95% | 6 CI | Z | р | Tau ² | R ² |
|-----------------------------|----|-------|-------|----------------|----------------|-------|-------|------------------|----------------|
| variable | K | D | 52 | Lower limit | Upper limit | | | | K |
| Intervention time (week) | 30 | 0.044 | 0.044 | -0.042 | 0.130 | 1.000 | 0.318 | 0.268 | 0.040 |
| Publication year | 39 | 0.008 | 0.023 | -0.038 | 0.055 | 0.350 | 0.725 | 0.277 | 0.005 |

Meta-Regression Moderator Analysis Findings

Before starting the meta-regression analyses, linear regression assumptions were checked separately using IBM SPSS Statistics 26 (IBM Corp., 2019). First, the graphs of the variables were checked to see whether they showed a linear relationship, and then the normality values were checked to see whether the data were normally distributed (Delen & Şen, 2023). It was seen that the graphs showed linear relationships. When the kurtosis and skewness values of the variables were analyzed separately, it was seen that these values were between -2 and +2, thus meeting the

normality assumptions (George & Mallery, 2010). Analyses regarding the moderators of implementation time and publication year were conducted separately with the available data, and the maximum likelihood method was used in the analysis according to the random effects model. As can be seen in Table 7, implementation time and year of publication were not significant predictors of the effect of the POE strategy on students' science achievement (p>0.05). In other words, it can be argued that the effect of POE on students' science achievement was not affected by changes in implementation time and publication year. Approximately 4% of the variance related to the effect of the POE strategy on students' science achievement is explained by the moderator of implementation time and approximately 5% by the moderator of publication year.

Discussion

In this study, which aimed to examine the effect of the POE strategy on achievement in science education by meta-analysis method, 39 independent effect size values were obtained from 35 individual studies. The analysis results of the specified methods were examined separately for possible publication bias. According to the findings, it can be concluded that there was no publication bias that could affect the validity of the analysis. In other words, publication bias was not an important factor affecting the average effect size. Although the p values of 8 studies included in the meta-analysis were not significant (Calış & Özkan, 2022; Chen et al., 2020; Chen, 2022; Hsiao et al., 2017; Hsu et al., 2011; James et al., 2022; Özçelik, 2019; Yaman & Ayas, 2015), the average effect size was calculated as 0.979 (Hedges' g) according to the random effects model. This is a high-level value according to Thalheimer and Cook's (2002) classification. In other words, the POE strategy has a high-level effect on students' science achievement. This finding is consistent with those of previous meta-analyses examining the effect of the POE strategy on learning outcomes (Gustina et al., 2023; Nurshafara, 2022). Gustina et al. (2023) found that POE had a medium-level effect on students' science and math learning outcomes, while Nurshafara (2022) concluded that POE had a great effect (0.995) on physics learning outcomes. The differences of the present study from the previous meta-analyses are that it included only studies conducted in the field of science, addressed different moderators, and utilized international databases. In the first stage of the POE strategy, students are motivated by drawing their attention to the experiment or activity to be carried out. Since a connection is established between students' prior knowledge and the information to be learned, the information is learned through structuring and meaningful learning is thus provided. With the POE strategy, students have opportunities for discussion while learning science concepts (Coştu, 2008, 2021). Since the stages of POE are clear and understandable, it is easy for teachers to implement it. For this reason, from the first studies conducted using the POE strategy (Gunstone & White, 1981; White & Gunstone, 1992) to the most recent research (Tüysüz & Özdemir, 2024), it has been consistently observed that POE contributes positively to cognitive domains such as student achievement, conceptual change, and knowledge level.

As a result of this meta-analysis, it was seen that the moderator of science field impacted the effect of the POE strategy on students' science achievement. Studies in geology affected students' science achievement at a low level; those in astronomy, biology, and physics had highlevel effects; those in chemistry had very high-level effects; and those in chemistry-physics affected science achievement at an excellent level. The fact that activities and experiments carried out in the fields of physics, chemistry, and biology can be done in a short time and can be repeated may make POE more effective in increasing student achievement in those fields. Experiments and

activities in the field of geology are generally based on observation and investigation. Therefore, assuming that students will be less active in the process of experiments or activities in the field of geology, it can be suggested that the implementation of the POE strategy in the field of geology is less effective in increasing student achievement. In addition, the fact that most of the concepts are theoretical and include events and phenomena that have already occurred or will occur over a long period of time may cause difficulties in teaching the concepts and information in the field of geology. In fact, only one study in the field of geology (James et al., 2022) was included in this meta-analysis, and that study was supported by simulations and video demonstrations. Similarly, there was only one study in the field of chemistry-physics (Coştu, 2021) included in this meta-analysis. When those studies were excluded from the meta-analysis, the difference between the studies remained similar. Therefore, there is a need for more studies in astronomy, geology, chemistry-physics, and other combinations of different fields of science. The studies included in this meta-analysis were mostly conducted in the fields of physics, chemistry, and biology, which are the main branches of science. In this respect, it can be suggested that studies in other science fields are needed for future meta-analysis studies.

In 13 of the studies included in this meta-analysis, the POE strategy was supported by various technologies such as animations, simulations, videos, and digital games. For this reason, the implementation type of POE (POE or technology-supported POE) was determined as a moderator variable in affecting students' science achievement. POE and technology-supported POE affected students' science achievement at similarly high levels. Accordingly, it can be argued that POE is a powerful strategy in influencing students' science achievement on its own, without being supported by technology. Since the POE strategy is compatible with social constructivism (Kearney, 2004), it provides cognitive, affective, and psychomotor gains such as attracting interest, providing motivation and active participation, and learning by structuring knowledge. Although the use of technology may have made a difference in the POE stages by preventing mediocre teaching, it may not have changed the learning outcomes. It can be suggested that conducting a hands-on experiment in a computer environment and trying to draw students' attention with a video or animation instead of by asking questions would have equal effects. In other words, the goal is the same, but the means are different. This finding of the present study is incompatible with the results of Tüysüz and Özdemir (2024), who concluded that the implementation of the POE strategy enriched with simulations was more effective than POE alone in increasing student achievement. However, the results of studies in the literature comparing the effects of technology and hands-on experiments on student achievement showed that computerized experiments involving animation and simulation were as effective as hands-on experiments (Evangelou & Kotsis, 2019; Zacharia & Olympiou, 2011). This inconsistency in the literature may be due to differences in the implementation of the POE strategy. Furthermore, this finding indicates the need for more research on the impact of implementing the POE strategy alone and with the support of technology.

The studies included in this meta-analysis were also examined according to the level of education. The effect of POE on students' science achievement differs significantly according to this moderator. The POE strategy had a very high effect on students' science achievement in high school and university, a high effect in elementary school, and a medium effect in middle school and preschool. There was only one study conducted at the preschool level (Hsu et al., 2011) included in this meta-analysis. When that study was excluded from the meta-analysis, the difference between the studies remained similar. This finding of the present study is similar to the results of Gustina et al. (2023) and Nurshafara (2022) regarding the effectiveness of the POE

strategy at high school level. However, it is inconsistent with the finding of Gustina et al. (2023) that POE has a very low effect on student achievement at the level of elementary school. This may be due to the fact that Gustina et al. (2023) included studies in the fields of mathematics and natural sciences in their meta-analysis. In the present meta-analysis, only studies in the field of science education were included. In the present meta-analysis, implementation of the POE strategy in elementary school was found to be highly effective. Erdem and Özcan Uyanık (2022) stated that implementation of POE at the elementary school level is fun and intriguing and provides permanent learning. Since students in elementary school tend to explain their thoughts through verbal communication, Palmer (1995) argued that it is more effective to apply the POE strategy at this level rather than in middle or high school.

The results of this study also showed that the effect of the POE strategy on students' science achievement was not affected by changes in the implementation time or publication year. Although it can be seen that these studies have increased in recent years, the publication years of the studies included in this meta-analysis varied between 2008 and 2023. In this regard, Gustina et al. (2023) concluded that the POE strategy was effective on critical thinking and learning outcome in studies conducted between 2013-2017 at a medium level and in studies conducted between 2018-2022 at a low level. The researchers stated that this situation is likely to be due to COVID-19. The implementation time of the studies varied between 1 and 12 weeks, with an average of 4.8 weeks. According to the results of this research, it can be claimed that there is no difference between short-term and long-term implementation of the POE strategy for student achievement.

Limitations and Suggestions for Future Studies

The first limitation of this meta-analysis is that only studies in Turkish and English were examined. Therefore, in future meta-analysis studies, the effect of the POE strategy on science achievement could be examined with the review of studies published in different languages by a team of expert researchers. The moderator variables of this study constitute another limitation. The moderator variables were limited to field of science, level of education, implementation type, implementation time, and year of publication. In this context, meta-analysis studies on the POE strategy could be conducted using different moderators. The year range of the included studies and the databases searched are other limitations of this meta-analysis study.

Within the scope of this study, the effect of the POE strategy on learning outcomes such as conceptual understanding, academic achievement, and conceptual change was addressed as science achievement. Accordingly, it was found that the effect of POE on science achievement was high. Based on this result, the POE strategy could be integrated into science lessons in order to understand and construct concepts in the field of science correctly. Experimental research on the effectiveness of the POE strategy in the fields of geology and astronomy at preschool and elementary school levels should be conducted, since there is a limited number of experimental studies in geology and astronomy at those levels. The POE strategy could be implemented in long-term research while combining different science fields. Meta-analysis studies could be conducted on the effects of POE on different cognitive and affective outcomes. Qualitative studies on the POE strategy could be systematically examined and meta-synthesis studies could be conducted. In particular, meta-synthesis studies that explain in detail how POE can be implemented in classrooms would be useful for teachers and academics who are practitioners in the field.

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Geniş Özet

Problem Durumu

Fen eğitiminde klasik yaklaşımdan yapılandırmacı yaklaşıma geçişle birlikte öğrenci merkezli öğretim yöntem ve teknikleri ön plana çıkmıştır. Araştırma-sorgulamaya dayalı öğrenme çatısı altında yer alan, sosyal yapılandırmacılık ile uyumlu olan, öğrencilerin kendi öğrenmelerinden sorumlu olduğu tahmin gözlem açıklama (TGA) öğrenci merkezli stratejilerden biridir. Klasik araştırma basamaklarına dayanan TGA hipotez oluşturma, hipotezin gerekçelerini ifade etme, hipotezi test etmek için veri toplama ve sonuçları tartışma adımlarını içermektedir (Kearney, 2004; White, 1988). TGA'da öğrenciler gerçekleştirilecek bir deney veya etkinlik öncesinde gerekçeli tahminde bulunurlar, daha sonra deney veya etkinlikle ilgili gözlem yaparlar. Son olarak tahmin ve gözlemleri arasında bir fark olup olmadığını açıklarlar (White, 1988; White & Gunstone, 1992).

TGA, uzun süredir fen eğitiminde kullanılmaktadır (Gunstone & White, 1981; White & Gunstone, 1992). TGA stratejisinin fen başarısına etkisini konu alan pek çok deneysel araştırma yapılmıştır. Bu deneysel araştırma sonuçlarının birleştirilerek bir meta-analiz çalışmasının yapılması genel sonucu görmeye yardımcı olacaktır. Literatürde TGA stratejisi ile ilgili şu ana kadar yapılan iki adet meta-analiz çalışmasına rastlanmıştır (Gustina vd., 2023; Nurshafara, 2022). Mevcut meta-analiz çalışmalarından farklı olarak, özellikle fen kavramların yapılandırıldığı deneylerde kullanılan TGA stratejisinin, fen başarısı üzerinde ne kadar etkili olduğunun belirlenmesi; bilim alanı, uygulanma şekli, öğretim kademesi, uygulama süresi ve yayın yılı gibi farklı moderatörlerin hesaba katılarak analiz yapılması önem teşkil etmektedir. Bu doğrultuda bu meta-analiz çalışmasında aşağıdaki problemlere yanıt aranmıştır:

(1) TGA stratejisinin öğrencilerin fen başarısı üzerindeki etkisi nedir?

(2) TGA stratejisinin öğrencilerin fen başarısı üzerindeki etkisi bilim alanına, uygulanma şekline ve öğretim kademesine göre anlamlı farklılık göstermekte midir?

(3) Uygulama süresi ve yayın yılı, TGA stratejisinin öğrenci fen başarısı üzerindeki etkisinin anlamlı bir yordayıcısı mıdır?

Yöntem

Bu araştırmada, TGA stratejisinin öğrencilerin fen başarısı üzerindeki etkisi meta-analiz yöntemiyle incelenmiştir. Çalışmaların seçimi ve analizinde Card (2012), Field ve Gillett (2010), Glass vd. (1981) tarafından önerilen 3 aşamalı prosedür izlenmiştir: (1) Çalışmaların seçimi, (2) kodlama stratejisi, (3) veri analizi.

Taramalar "predict", "observe" ve "explain" anahtar kelimeleri kullanılarak gerçekleştirilmiştir. Makalelerin taranmasında eğitimle ilgili kaliteli ve popüler bazı dergileri kapsadıkları için "Web of Science (WoS)", "Scopus" ve "Education Resources Information Center (ERIC)" veri tabanları; tezlerin taranmasında "ProQuest Dissertations & Theses Global (PQDT)"

veri tabanı tercih edilmiştir. Meta-analize dahil edilen çalışmaların belirlenmesinde uygulanan kriterler şunlardır: (1) TGA ve fen başarısı ile ilgili çalışmalar, (2) 2003-2023 yılları arasında yayımlanmış erişime açık tam metinli çalışmalar, (3) İngilizce veya Türkçe yayımlanmış çalışmalar, (4) çalışmaların tez veya hakemli dergilerde makale türünde yayımlanması, (5) çalışmaların TGA uygulaması içermesi, (6) çalışmaların, etki büyüklüğünün hesaplanması için yeterli veriye sahip olması. Bu kriterlere uygun olarak meta-analize PQDT'den 6, WoS'tan 6, Scopus'tan 18 ve ERIC'ten 5 olmak üzere toplam 35 çalışma dahil edilmiştir. Bu çalışmalardan 39 bağımsız etki büyüklüğü değeri elde edilmiştir. Yayın yanlılığı kontrolü, etki büyüklüğü değerlerinin hesaplanması, heterojenlik testi ve moderatör analizlerinin tamamında Comprehensive Meta-Analysis (CMA) programının ücretsiz deneme sürümü kullanılmıştır.

Bulgular

Analiz sonuçları, TGA stratejisinin öğrencilerin fen başarısı üzerinde yüksek düzey etkiye sahip olduğunu ortaya koymuştur. Bilim alanı moderatörüne göre TGA stratejisi öğrencilerin fen başarısını anlamlı olarak farklılaştırmıştır. Ortalama etki büyüklüğü değerleri astronomi, biyoloji, kimya, kimya-fizik, yerbilimi ve fizik bilim alanları için sırasıyla 0.760, 0.825, 1.327, 2.398, 0.215 ve 0.815'tir. TGA stratejisi yerbilimi bilim alanında öğrencilerinin fen başarısını Thalheimer ve Cook'un (2002) sınıflamasına göre düşük düzeyde; astronomi, biyoloji ve fizik alanlarında yüksek düzeyde, kimya alanında çok yüksek düzeyde, kimya-fizik alanında mükemmel düzeyde etkilemistir. TGA'nın tek basına ve teknoloji ile desteklenerek uvgulanma seklinin öğrencilerin fen başarısını benzer ve yüksek düzeyde etkilediği söylenebilir. Ortalama etki büyüklüğü değerleri TGA ve teknoloji destekli TGA için sırasıyla 1.019 ve 0.875'tir. Öğretim kademesi moderatörüne göre TGA stratejisi öğrencilerin fen başarısını anlamlı olarak farklılaştırmıştır. Ortalama etki büyüklüğü değerleri lise, ortaokul, okul öncesi, ilkokul ve üniversite için sırasıyla 1.229, 0.515, 0.494, 1.015 ve 1.100'dür. TGA stratejisi lise ve üniversite öğretim kademesinde öğrencilerin fen başarısını çok yüksek düzeyde, ilkokulda yüksek düzeyde, ortaokul ve okul öncesi kademelerinde orta düzeyde etkilemiştir. Meta-regresyon analizleri uygulama süresi ve yayın yılının, TGA stratejisinin öğrenci fen başarısı üzerindeki etkisinin anlamlı birer yordayıcısı olmadığını göstermiştir (p>0.05).

Sonuç ve Tartışma

Araştırma sonucunda TGA stratejisinin öğrencilerin fen başarısı üzerinde yüksek düzeyde bir etkive sahip olduğu görülmüstür. Bu bulgu TGA stratejisinin öğrenme cıktıları üzerindeki etkisini inceleyen önceki meta-analiz çalışmaları ile uyumludur (Gustina vd., 2023; Nurshafara, 2022). Bilim alanı moderatörüne göre TGA stratejisi öğrencilerin fen başarısını anlamlı olarak farklılaştırmıştır. Fizik, kimya ve biyoloji alanlarında gerçekleştirilen etkinlik ve deneylerin kısa sürede yapılabilmesi, tekrar edilebilir olması TGA stratejisini öğrenci başarısını artırmada daha etkili kılmış olabilir. Yerbilimi alanında yapılan deney ve etkinlikler genelde gözlem ve incelemeye dayalıdır. Dolayısıyla yerbilimi alanında yapılan deney ve etkinlik sürecinde öğrencilerin daha az aktif olacağı düşüncesinden yola çıkarak, yerbilimi alanında TGA stratejisinin uygulanmasının öğrenci başarısını artırmada daha az etkili olduğu iddia edilebilir. Araştırma sonucunda TGA ve teknoloji destekli TGA'nın öğrencilerin fen başarısını yüksek ve benzer düzeyde etkilediği görülmüştür. Buradan TGA'nın öğrencilerin fen başarısını etkilemede teknoloji ile desteklenmeden tek başına uygulanmasının da etkili olduğu iddia edilebilir. Bu meta-analiz çalışmasında TGA stratejisinin ilkokul kademesinde uygulanmasının yüksek düzeyde etkili olduğu görülmüştür. Erdem ve Özcan Uyanık (2022), TGA'nın ilkokul kademesinde uygulanmasının

eğlenceli ve merak uyandırıcı olduğunu, kalıcı öğrenmeyi sağladığını ifade etmişlerdir. İlkokul kademesindeki öğrenciler sözlü iletişim kurarak düşüncelerini açıklama eğiliminde oldukları için Palmer (1995) TGA stratejisinin ortaokul ve lise kademelerinde uygulanmasından ziyade ilkokul kademesinde uygulanmasının daha etkili olduğunu öne sürmüştür. Ayrıca, TGA stratejisinin öğrenci fen başarısı üzerindeki etkisinin uygulama süresi ve yayın yılı değişkenlerinden etkilenmediği görülmüştür. Meta-analize dahil edilen çalışmaların son yıllarda yoğunlaştığı görülmekle birlikte yayın yılları 2008-2023 arasında değişmektedir. Çalışmaların uygulama süresi 1-12 hafta arasında değişmekle birlikte ortalaması 4.8 haftadır.