About the Match of Trainees' Learning Styles to Their Tutors' Instructional Strategies and the Correlation of This Match with the Trainees' Academic Achievements

Hagit Krisher¹ & Nissim Sabag²

¹ Support Center for Students with Special Needs, ORT Braude College, Karmiel, Israel

² Department of Electrical and Electronic Engineering, ORT Braude College, Karmiel, Israel

Correspondence: Hagit Krisher, Support Center for Students with Special Needs, ORT Braude College, P.O. Box 78, Karmiel 21982, Israel. Tel: 972-4-990-1909.

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Abstract

The match between students' learning styles (LS) and their teachers' instructional strategies (IS) and the correlation of this LS-IS match with students' academic achievements was studied in earlier research. However, there is no report of research where one-on-one education is implemented. Moreover, there are no references relating the match of a trainees' LS to their tutors' IS and the correlation of this LS-IS match with the trainees' achievements. Accordingly, the current paper presents a study designed to investigate the correlation between a trainees' achievements and the LS-IS match. Two different methods were used to measure the LS-IS match. First, calculating the correlation between trainees' LS and tutors' IS (LS-IS correlation); second, calculating the LS-IS distance. Forty-two trainees with learning disabilities were paired with 39 tutors (three tutors had two trainees each) during the 2016 academic year. Thus, 42 pairs of trainees and tutors worked to help the trainees achieve better academic grades. The Felder-Soloman Index of Learning Styles (ILS) was used to measure the trainees' preferred LS and the tutors' preferred IS. In the first method, the LS-IS correlations were correlated with the trainees' grades; then, in the second method, the LS-IS distances were correlated with trainees' grades. If the LS-IS match influences the trainees' achievements, significant positive correlations in the first method and significant negative correlation in the second method must appear. However, the results show no significant correlation (positive or negative, accordingly) between the LS-IS match and students' achievements at the end of the first semester of 2016. A replication of the above study was made in the second semester of 2016 and similar results were obtained.

Keywords: learning styles, instructional strategies, learning disabilities, tutor

1. Introduction

During the last two decades, one of the leading engineering colleges in Israel has been running a special program to help students with learning disabilities. Among other means, the program provides peer tutoring that was suggested by many researchers as a useful tool to improve academic achievements (Bowman-Perrott, Davis, Vannest, Williams, Greenwood, & Parker, 2013; Okilwa & Shelby, 2010). Students with learning disabilities are paired (as trainees) with excellent students who are strictly selected and trained to serve as tutors, and work together during two semesters to improve the trainees' academic achievements.

Much has been written on the relationship between the match of learning styles to instructional strategies (*LS-IS match*) and students' achievements (Safe, 2008; Tella, Tella, & Adeniyi, 2009; Gilakjani, 2012). Therefore, the authors of the current paper thought it interesting to investigate the relationship between the match of trainees' learning styles to the tutors' instructional strategies (*LS-IS match*) and the trainees' academic achievements. Accordingly, the current study designed to investigate whether a good *LS-IS match* influences trainees' achievements.

The paper is organized as follows: Section 2 provides a theoretical review on peer tutoring; definitions of learning styles and models of learning styles followed by the Felder–Soloman Index of Learning Styles (*ILS*) and its validity and reliability; and learning styles and academic achievement. Section 3 presents the research question, the research

method, and the research setting. In Section 4, the findings are described. In Sections 5 and 6, the findings are further discussed, conclusions presented, and future research directions suggested.

2. Literature Review

2.1 Peer Tutoring

The positive effects of peer tutoring have been demonstrated across much research over the last 40 years. The success of peer tutoring for both tutors and tutees is likely from incorporating instructional features such as frequent opportunities to respond, increased time on task, and regular and immediate feedback. Each of these components is empirically linked to increased academic achievement (Bowman-Perrott, Davis, Vannest, Williams, Greenwood, & Parker, 2013). Peer tutoring is a commonly provided support service for trainees with learning disabilities (LD) in institutions of higher education. The positive effects of peer tutoring have been demonstrated across subjects such as reading, math, and science (Guild & Garger, 1985). In addition, peer tutoring is effective for trainees with and without disabilities, native English-speaking students, and English language learners (Okilwa & Shelby, 2010). The positive effects of peer tutoring have been demonstrated across subjects such as reading (Oddo, Barnett, Hawkins, & Musti-Rao, 2010), math (Hawkins, Musti-Rao, Hughes, Berry, & McGuire, 2009), social studies (Lo & Cartledge, 2004), and science (Bowman-Perrott, Greenwood, & Tapia, 2007). Findings suggest that peer tutoring is an effective intervention, regardless of dosage, grade level, or disability status. Among students with disabilities, those with emotional and behavioral disorders benefitted most (Bowman-Perrott, Davis, Vannest, Williams, Greenwood, & Parker, 2013). However, there are no references relating trainees' learning styles to their tutors' instructional strategies. Therefore, in this research, we investigated whether a good match between trainees learning styles to their tutors' instructional strategies (LS-IS match) influences trainees' achievements.

2.2 Definition of Learning Style and Models of Learning Style

Learning styles can be defined, classified, and identified in many different ways. They can also be described as a set of factors, behaviors, and attitudes that enhance learning in any situation. How the students learn and how the teachers teach, and how the two interact with each other, are influenced by different learning styles (Chermahini, Ghanbari & Talab, 2013). Within the last three decades, the proposition that students learn and study in different ways has emerged as a prominent pedagogical issue. Learning styles (Coffield, Moseley, Hall, & Ecclestone, 2004) and learning style models (Gregore, 1979, 1985; Kolb, 1984; Felder & Silverman, 1988; Dunn, Dunn, & Price, 1989; Fleming, 2001; Duff, 2004) have offered descriptive typologies. Researchers have developed a vast array of models and instruments in an attempt to understand and develop a framework that explains how students learn. Coffield et al. (2004) have extensively reviewed the learning styles literature, evaluated the major learning styles models, and discussed the implications for practice. They identified 71 learning models and instruments and categorized 13 of them as major models. Hawk and Shah (2007) reviewed and compared five of the more commonly and recently used learning style models and instruments: the Kolb Learning Styles Indicator, Gregorc Style Delineator, Felder-Silverman Index of Learning Styles, VARK Questionnaire, and Dunn & Dunn Productivity Environment Preference Survey. Their conclusion was that no one instrument can capture all the richness of the phenomenon of learning style. Kolb (1984) and Mumford and Honey (1992) describe learning style as an individual preferred or habitual way of processing and transforming knowledge. According to Kolb (1984), psychological attributes, resulting from individual differences, determine the particular strategies a person chooses while learning. Kolb and Boyatzis (1993) present four distinct learning styles (or preferences), which are based on a four-stage learning cycle: concrete experience, reflective observation, abstract conceptualization, and active experimentation.

2.3 Index of Learning Styles (ILS)

In 1988, Richard Felder and Linda Silverman developed a learning model that focuses specifically on aspects of learning styles of engineering students (Felder & Silverman, 1988). The learning style model classifies students as having preferences for one category or the other in each of the following five dimensions: concrete/abstract, visual/auditory, inductive/deductive, active/reflective, and sequential/global. Later on the inductive/deductive dimension was omitted and the visual/auditory was replaced by the visual/verbal (Felder, 2002). The model posits that active students learn by trying things out and working with others; reflective students prefer to think things through and work alone; sensory students are practical and oriented toward facts and procedures; and intuitive students are conceptual, innovative, and oriented toward theories and meanings. Visual students prefer visual representations of material such as pictures, diagrams, or flow charts; verbal students prefer written and spoken explanations; sequential students tend to follow linear and orderly reasoning processes; and global students prefer to learn in intuitive leaps (Felder, 1993; Felder & Silverman, 1988; Litzinger, Sang-Ha, Wise & Felder, 2005). Lumsdaine and Lumsdaine (1995) concluded that the lack of congruence between preferred learning style and the

nature of the subject matter and the method of teaching related to comparatively lower motivation and poorer performance, and hence possible failure to complete a course.

For this study, we elected to use the Felder-Soloman model and Index of Learning Styles (ILS) (Soloman & Felder, 2014), which is an improvement of the Felder-Silverman model (Felder, 2002), because it is an instrument with a significant amount of study and use. Felder and Soloman developed this model in an engineering education environment that was relevant to our college's students. Many researchers relate to the validity and reliability of the Felder-Soloman Index of Learning Styles (Boyle, Duffy, & Dunleavy, 2003; Hlawaty, 2002; Felder & Spurlin, 2005; Litzinger et al., 2005; Zywno, 2003). Zywno uses 557 questionnaires for her Cronbach alpha analysis that resulted in Cronbach alphas between 0.53 and 0.70 for the four LS dimensions, whereas alpha > 0.5 is acceptable for attitude assessment. She also points out that three of the four dimensions are orthogonal and there is a small correlation between the sensing/intuitive and the sequential/global dimensions (Zywno, 2003). The Index of Learning Styles (ILS - Felder and Soloman, 2014) was "designed to capture the most important learning style differences among engineering students and provide a good basis for engineering instructors. The model shares commonalities with other popular learning style approaches, e.g., the Learning Style Inventory (LSI - Kolb, 1984) or the Myers-Briggs Type Indicator (MBTI; Lawrence, 1984). The ILS is explicitly said to not include 'either-or categories' of its bipolar dimensions. All scales are to be understood as continua, which means that a student's cognitive preference to learning on a given *ILS* scale may be either fairly well balanced, moderately, or strongly distinctive for one or the other pole of the scale. The four bipolar ILS dimensions can be described as follows (Felder and Soloman, 2016):

- 1. Active Reflective: Active learners tend to gather and understand information best if they engage with it actively and try things out, e.g., by debating, bringing something to application, or via teaching back. Reflective learners prefer to think about new things for themselves first and learn by thinking things through. The motto of active learners is "Let us try and see how it works", whilst reflective learners pursue the principle "Let me first think carefully about it".
- 2. Sensual Intuitive: Sensing learners tend to do well when learning facts, and follow established approaches and procedures when solving problems. They are more goal-oriented, progress carefully and patiently, but avoid complex issues or surprises. Intuitive learners on the other hand prefer to explore different possibilities, relationships, and innovative approaches. They can better grasp new concepts, work usually faster and more innovatively, and have less difficulty with abstract concepts and mathematical expressions. However, they tend to avoid rote learning, repetition, routines, and fixed schemes.
- 3. Visual Verbal: Visual learners remember more of what they have seen; for example, in pictures, diagrams, flow charts, films, and demonstrations. Instead, verbal learners prefer linguistically based learning that is written, and spoken information or declarations.
- 4. **Sequential Global**: Sequential learners tend to understand better by learning in logical linear steps, where each step is the logical consequence of the previous step. In contrast, global learners rather tend to make big steps and gather different material and information quasi-randomly and without the recognition of contexts and relationships, but suddenly, they understand the whole context.

The *ILS* consists of 44 statements, 11 for each dimension. The respondent can choose 'a' or 'b', depending how each reflects his or her preference for each statement. For example, for the statement "I understand something after I"..., an active learner would mark 'a – try it out' and a reflective learner would mark 'b – think it through' (Soloman & Felder, 2014). Every 'a' response counts as +1 whereas a 'b' response counts as -1; this scoring method dictates the range of -11 to +11 for each dimension (Felder & Spurlin, 2005; Litzinger et al., 2005; Graf, Viola, Leo, & Kinshuk, 2007).

2.4 Learning Styles and Academic Achievements

Various scholars have defined learning style mostly as a signal for individual differences. These differences may manifest themselves in 'life styles' and even in personality types (Zhang & Sternberg 2005). Different individuals use different learning styles and the effectiveneness of the learning style also varies among individuals (Warn, 2009). Several recent researchers claim that students' academic achievement is influenced by their learning styles (Gilakjani, 2012). Gilakjani (2012) argues that teachers should make every effort to match their instructional strategies to the student's learning style. He also claims that matching between a student's learning style and the teacher's instructional strategies (*LS-IS match*) could produce statistically significant improvements in the students' grades (Gilakjani, 2012). On the other hand, other research showed that learning style had no significant impact on

achievement (Marrison & Frick, 1994). Boyle, Duffy, and Dunleavy (2003) examined the relationship between learning style and academic performance and found moderate negative correlations between learning style and grade point average. They suggest that the *ILS* has a rather limited role to play in predicting academic outcome and that *ILS* may have a more useful diagnostic role to play in higher education, for early-stage detection of learners with inappropriate orientations to learning, who fail to adopt systematic processing strategies and consequently are in danger of failing to achieve the maximum benefit from their time in higher education (Boyle, Duffy, and Dunleavy, 2003).

It should be noted that these researchers examined the impact of specific learning styles on academic achievement and none of them demonstrated that matching a student's learning style to the teacher's instructional strategies is positively associated with academic performance. Furthermore, Brown et al. (2006) found no significant differences in performance between matched and mismatched students. In addition, Graf and Kinshuk (2007) analyzed the students' performance and behavior in the course, and found that students who learned from a course that matches their learning styles spent significantly less time in the course and achieved on average the same marks as students in a course that either mismatched their learning styles or included all available learning styles.

Sabag and Trotskovsky developed a method to measure the distance of student's learning style to the teacher's instructional strategies (*LS-IS distance*) as one of the *LS-IS match* methods, and examined the relationship between the *LS-IS distance* and the students' achievements in three schools with 165 students who studied 17 courses with eight teachers. They conclude that there is no evidence to the claim that matching learning styles to instructional strategies affects the students' achievements.

Moreover, Felder suggests that the professors teach their students so that every student will be able to function with all learning style modes. In his words, "if professors teach exclusively in a manner that favors their students' less preferred learning style modes, the students' discomfort level may be great enough to interfere with their learning. On the other hand, if professors teach exclusively in their students' preferred modes, the students may not develop the mental dexterity they need to reach their potential for achievement in school and as professionals" [1996, p.18].

In general, there is rich data obtained from studies on learning styles; however, there are no references relating trainees' learning styles to their tutors' instructional strategies and examining their *LS-IS match* correlation with the trainees' academic achievements.

3. Methodology

3.1 Research Question and Hypothesis

The purpose of this current study is to resolve the following: Does a close *LS-IS match* between trainees' *LS* to their tutors' *IS* improve trainees' achievements? There is disagreement about the contribution of *LS-IS match* to students' achievements in the literature; therefore, it is interesting to explore whether a good match between trainees' learning styles to their tutors' instructional strategies positively influences trainees' achievements.

3.2 Research Population

The research population comprised 42 trainees and their 39 tutors. They were organized in 42 trainee-tutor pairs, where three tutors had two trainees each. The response rate is 32 out of 42 teams (ten teams missed part of the information and are therefore not included in the calculations). The trainees are students with learning disabilities. The tutors are excellent students in their $2^{nd} - 4^{th}$ years of study, who were selected through comprehensive interviews and had strict preparation before receiving authorization to serve as tutors. The trainee-tutor teams worked together during one academic year (two semesters) to improve the trainees' academic achievements.

3.3 Learning Styles Questionnaire

The Index of Learning Styles Questionnaire *(ILS)* (Soloman, & Felder, 2014) was translated into Hebrew and validated using the following method. Four groups of 20–30 students, who are not part of the research population, answered the questionnaire successively and were asked to write comments about any unclear issue they encountered. All the comments from the first group were discussed and the appropriate adaptations were incorporated into the questionnaire. Thereafter, the second group filled in the questionnaire. This procedure was repeated until the Hebrew version of the questionnaire was fully clear to the fourth group. Internal consistency was checked by calculating Cronbach's alpha for each dimension. The results, shown in Table 1—which presents values of Cronbach's alpha for the research population—are in line with the literature (Felder & Spurlin, 2005; Zywno, 2003). According to Tuckman (1999), an alpha of 0.50 or greater is acceptable for questionnaires that assess attitude and preference. As

shown in Table 1, all the alpha values meet this criterion. Therefore, the internal consistency of the questionnaire is satisfied.

Table 1. Cronba	ch's alpha	of each din	nension in	the LS	questionnaire
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Dimension	Dimension Visual–Verbal S		Sensing–Intuitive	Active–Reflective
	(k=4)		(k=2)	(k=1)
Cronbach's alpha ($n = 165$)	0.6	0.5	0.6	0.5

According to Gilakjani (2012), teachers teach the way they preferred to learn when they were students. Therefore it is reasonable to conclude the same with tutors and to use the same questionnaire for both trainees and for tutors. The tutors were asked to think of the way they guide while answering the questionnaire. For example, exchanging the word *understand* with *explain* makes the tutors' item (4) of the *ILS*, "I tend to" as either:

a. explain details of a subject but may be fuzzy about its overall structure.

or b. explain the overall structure but may be fuzzy about the details.

Based on Graf et al. (2007), Table 2 describes the items in the questionnaire associated with LS dimensions.

Table 2. Questionnaire items associated with LS dimensions

k	Dimension	Items in the questionnaire	°a' preference	'b' preference
1	Active-Reflective	1,5,9,13,17,21,25,29,33,37,41	Active	Reflective
2	Sensing-Intuitive	2,6,10,14,18,22,26,30,34,38,42	Sensing	Intuitive
3	Sequential-Global	4,8,12,16,20,24,28,32,36,40,44	Sequential	Global
4	Visual–Verbal	3,7,11,15,19,23,27,31,35,39,43	Visual	Verbal

3.4 Measuring Methods

To examine the relationship between *LS-IS* closeness and trainees' achievements, two methods were conducted. First, correlation between trainees' *LS* and their tutors' *IS* was calculated, then this *LS-IS correlation* was correlated to trainees' achievements. The second method is explained in detail in paragraph 3.4.1 below.

3.4.1 Measuring LS-IS Correlations

Correlations between the total tutor's IS and the total trainee's LS were calculated for each trainee-tutor pair. It was expected that a higher trainee-tutor correlation would yield higher trainee's achievements. However, there is a problem in calculating correlations for each individual LS dimension; when one of the trainee-tutor pair is consistent in his or her choice for one dimension (e.g., one prefers 'a' for all the items of k4), the correlation calculation results in 0/0, which is not defined, and the measurement of this trainee-tutor pair must be eliminated. A lot of data is lost for this reason.

3.4.2 Measuring LS-IS Distance

The method of measuring *LS-IS distance* was first demonstrated in Sabag and Trotskovsky (2016). Following is a short explanation for the reader's convenience, using numbers updated to the current study.

The score of dimension k for trainee i is written as LS_{ik} where $1 \le k \le 4$ and is calculated by summing all 'a' preferences and 'b' preferences that are associated with dimension k. Note that the total *LS* of trainee i is LS_i and is calculated using Equation (1).

$$LS_i = \sum_{k=1}^4 LS_{ik} \tag{1}$$

In other words, LS_i is the sum of the four dimensions' score for trainee i. LS_i is defined similarly for tutor j.

To define the *LS–IS distance*, let LS_{ik} be the score given by trainee i for dimension k, and IS_{jk} the score given by tutor j for the corresponding dimension. The absolute difference $D_{ijk} = |LS_{ik} - IS_{jk}|$ is the distance between trainee i and tutor j for dimension k. The total distance between trainee i's LS_i and tutor j's LS_j is the sum of all distances of the four dimensions, calculated by Equation (2):

$$D_{ij} = \sum_{k=1}^{4} \left| LS_{ik} - IS_{jk} \right|$$
(2)

For example: the *LS* of trainee 8 for k1 (Active–Reflective) is 11. The *IS* of tutor 8 for k1 is 3. Therefore, $LS_{8,1} = 11$; $IS_{8,1} = 3$ and the distance between trainee 8 and tutor 8 for k1 is $D_{8,8,1} = |11-3| = 8$.

The other measures for trainee 8 are $k^2 = 11$, $k^3 = 7$, and $k^4 = 9$, so the total $LS_8 = 38$. Similarly, for tutor 8, the measures are $k^2 = 3$, $k^3 = -3$, and $k^4 = 7$, so the total is $IS_8 = 10$.

The total distance between trainee 8 and tutor 8 is $D_{8,8} = |38-10| = 28$.

4. Results

The 32 teams of trainees and their tutors who participated in the research filled out the LS and IS questionnaires, respectively. The averages of all final grades in the first semester of 2016 and then again in the second semester of 2016 were calculated for each trainee. At the beginning, correlations between trainee's LS and his or her tutor's IS were calculated (LS-IS correlations) for each dimension and for the total LS-IS, for all trainee-tutor pairs. The blank spaces indicate cases where the correlation resulted in 0/0 and was therefore eliminated. The results are depicted in Table 3. The last two rows present the correlations between LS-IS correlations and the trainees' achievements for each dimension and for the total LS-IS.

Pair no.	LS-IS	correlat	<i>ion</i> for ea and total	ach dim I	First semester trainees'	Second semester	
	K1	K2	K3	K4	Total	average grades	trainees' average grades
1	0.19	0.52	0.26	0.15	0.09	76.4	84.5
2	0.45	0.39	0.47		0.38	70.9	62.0
3	0.62	-0.26	-0.26	0.10	0.03	82.9	83.4
4	-0.46	0.45	0.26		0.02	82.6	71.6
5	0.21	-0.29	0.24	0.36	0.12	73.5	71.5
6	0.39		-0.15		0.15	78.8	77.1
7	0.04	-0.39		0.07	0.03	84.2	73.9
8			0.36	-0.15	0.16	79.3	89.3
9	-0.42	0.04	0.45	0.45	0.13	73.3	20.8
10	-0.61	-0.29	0.21	-0.15	-0.18	79.0	80.7
11		0.15	-0.07	-0.22	-0.12	85.1	84.9

Table 3. Trainee-tutor LS-IS correlation and trainees' achievements

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12	-0.31	-0.15	0 10	0 24	-0.06	77 2	77 3
13	-0.13	0.29	0.29	0.31	0.17	80.6	66.6
14	0.47	-0.19	0.52	0.38	0.29	72.0	67.9
15	0.15	-0.22	-0.45	-0.10	-0.06	79.9	77.3
16	-0.07	-0.04	-0.10	-0.29	-0.08	80.6	74.1
17	0.24		-0.24		-0.03	88.0	77.3
18	-0.36	0.29	0.24	-0.13	0.00	86.5	83.7
19	0.81	-0.19	-0.04	-0.04	0.32	71.2	40.6
20		-0.19	0.45	-0.24	0.05	86.7	91.2
21	-0.63	0.81	0.26		0.06	75.3	80.6
22	-0.07	0.19	0.61	0.31	0.17	81.8	72.7
23	0.04	-0.24	0.08		-0.09	81.2	76.7
24		0.08	-0.62	0.21	-0.03	75.9	76.9
25	0.21	0.13	0.52		0.24	81.1	73.4
26	0.26	-0.36	0.31		0.11	76.6	59.8
27	0.24	0.21	-0.15	-0.21	0.04	82.9	65.7
28	0.04	-0.35	-0.24	-0.38	-0.18	77.9	85.5
29	0.04	0.13	-0.24	0.42	0.26	63.6	61.0
30	-0.27	-0.08	-0.27	0.42	-0.11	75.9	84.4
31	0.07	0.57	0.46	-0.19	0.26	78.8	69.8
32	0.63	0.45	0.38	-0.04	0.26	64.8	58.1
Correlations with first semester grades	-0.26	-0.14	-0.07	-0.47	-0.53		
Correlations with second semester grades	-0.19	-0.04	-0.21	-0.38	-0.52		

As shown in Table 3 there are significant negative correlations between *LS-IS correlations* and trainees' achievements, whereas positive correlations were expected.

The second method of examining the relationship between *LS-IS* closeness and trainees' achievements is to correlate *LS–IS distances* and the trainees' achievements. For this, the *LS–IS distances* were calculated for each trainee and his or her tutor. Correlations between *LS–IS distances* and the trainees' achievements in the first semester of 2016 and then again in the second semester of 2016 were analyzed.

An example of the measured LS_{1K} for trainee 1 and IS_{1K} for his tutor (1), the total LS_1 and IS_1 , the distances $D_{I,I,K}$ and $D_{I,I,K}$ are presented in Table 4.

Table 4. A detailed example of calculating LS_K , IS_K , and LS-IS distances for trainee 1 and tutor 1 and the trainee's average grades in first semester of 2016

Participant	k1	k2	k3	k4	Total	D _{1,1,1}	D _{1,1,2}	D _{1,1,3}	D _{1,1,4}	D _{1,1}	Average Grades
Tutor 1	-5	7	5	-1	6						
Trainee 1	9	1	1	5	16	14	6	4	6	10	76.42

The calculations of the *LS-IS distances* and their correlations with trainees' average grades for all tutors and their trainees appear in Table 5.

	Traine	ee-tutor d					
			total				
Trainee-Tutor	K1	K2	K3	K4	Total	First semester	Second semester
pair no.						average grades	average grades
1	14	6	4	6	10	76.4	84.5
2	2	2	2	10	12	70.9	62.0
3	4	6	6	2	14	82.9	83.4
4	2	2	4	14	22	82.6	71.6
5	0	2	12	10	20	73.5	71.5
6	2	2	4	6	10	78.8	77.1
7	6	8	10	2	26	84.2	73.9
8	8	8	10	2	28	79.3	89.3
9	12	6	2	2	18	73.3	20.8
10	6	8	0	4	18	79.0	80.7
11	16	16	4	0	36	85.1	84.9
12	2	4	2	12	16	77.2	77.3
13	10	12	10	4	36	80.6	66.6
14	2	4	4	10	0	72.0	67.9
15	6	0	4	0	2	79.9	77.3
16	4	8	0	2	10	80.6	74.1
17	12	10	6	18	46	88.0	77.3
18	4	10	2	10	2	86.5	83.7
19	2	4	2	2	2	71.2	40.6
20	16	4	2	6	20	86.7	91.2
21	2	2	4	12	16	75.3	80.6
22	4	14	0	4	14	81.8	72.7
23	8	12	0	8	12	81.2	76.7
24	10	0	10	0	20	75.9	76.9
25	0	4	6	4	6	81.1	73.4
26	4	4	4	6	10	76.6	59.8
27	2	0	4	6	12	82.9	65.7
28	8	10	12	0	30	77.9	85.5
29	8	4	6	6	24	63.6	61.0
30	2	10	2	6	0	75.9	84.4
31	2	6	8	4	4	78.8	69.8
32	0	2	10	8	4	64.8	58.1
Correlations with first semester grades	0.32	0.42	-0.15	0.01	0.32		
Correlations with second semester grades	0.17	0.24	0.09	0.05	0.18		

Table 5	Calculations	of the	LS-IS	distances	and t	he c	correlation	with	trainee'	saverage	orades
rable J.	Calculations	or the	LO^{-1O}	uisiunces	and u		Joniciation	vv I tIII	uanice	s average	grades

As shown in Table 5, there is no single D_{jik} or D_{ji} with consistent significant negative correlations.

5. Discussion

All the research, in the literature, concerning the *LS-IS match* and its correlation to students' achievements was done in the group context, i.e., one teacher and a group of students. Therefore, even if the findings show solid evidence proving that a good *LS-IS match* positively affects students' achievements, this still leaves the teacher with a big dilemma: to which style to adjust his or her teaching, since every group has many different *LS*. The current study suggests a unique opportunity in which the learning–teaching is a one-on-one situation. Therefore, it could be helpful to report the findings of such a case.

The Felder-Silberman *ILS* questionnaire, whose validation and reliability are well proven, measures the trainees' *LS* and their tutors' *IS*. Moreover, the Hebrew version of the questionnaire was validated as explained in paragraph 3.3 and its consistency was calculated as depicted in Table 1. The *LS-IS match* was calculated in two methods. First, the *LS-IS correlation* for every dimension (K1-K4) and for the total *LS-IS were* calculated for each trainee-tutor pair; the results appear in Table 3. If the assumption is that a close *LS-IS match* influences the trainees' achievements, then a significant positive correlation must appear between trainees' achievements and *trainee-tutor LS-IS correlation*. Table 3 reveals no such positive correlation. It means that there is no influence of the *LS-IS match* on the trainees' achievements (at least, no positive influence).

In the second method to examine the relationship between *LS-IS match* and trainees' achievements, the *LS-IS distance* (for every dimension and for the total) between each trainee and his or her tutor was calculated as explained in paragraph 3.4.2 above; the results are detailed in Table 5. Again, if the assumption is that a close *LS-IS match* influences the trainees' achievements, then a significant negative correlation must appear between trainees' achievements and *trainee-tutor LS–IS distance* (i.e., the smaller the distance, the higher the achievement). The results in Table 5 show no significant negative correlation to support the assumption that a good *LS-IS match* has a good influence on trainees' achievements.

6. Conclusion

The current paper presents a study designed to investigate the correlation between trainees' achievements and the *LS-IS match*. There is no report of research where one-on-one education is implemented so it is a good opportunity to report on the described case. Although it should be noted that the tutors are not teachers; they do not teach a certain course, but meet their trainees for a few hours a week, every week, and help their trainees overcome learning obstacles. They have a meaningful influence on their trainees' learning, but their influence is not the only one; teachers of the different courses also have an impact on the trainees' achievements.

The findings of this study should be considered in light of the following limitations. One limitation is that the described study deals with engineering students; it is suggested to run similar research with different disciplines and different populations (e.g., humanities). Another recommendation is that future studies should apply the correlation of trainees' achievements with their attitudes toward peer tutoring and/or toward their own tutors. Future research should also examine the relation between academic and behavioral outcomes for students engaged in peer tutoring.

The results of the future studies suggested above might contribute to an understanding of the learning styles and instructional strategies of students in the context of one-on-one situations.

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