

# Leveraging the Testing Effect using Parametrized Tasks in Learning Management Systems

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contact information

# Introduction



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Material

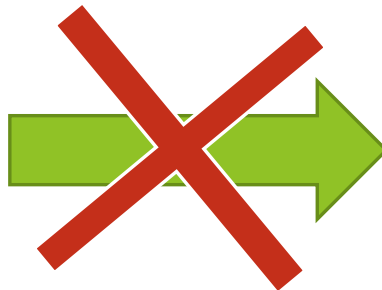
Hypothesis

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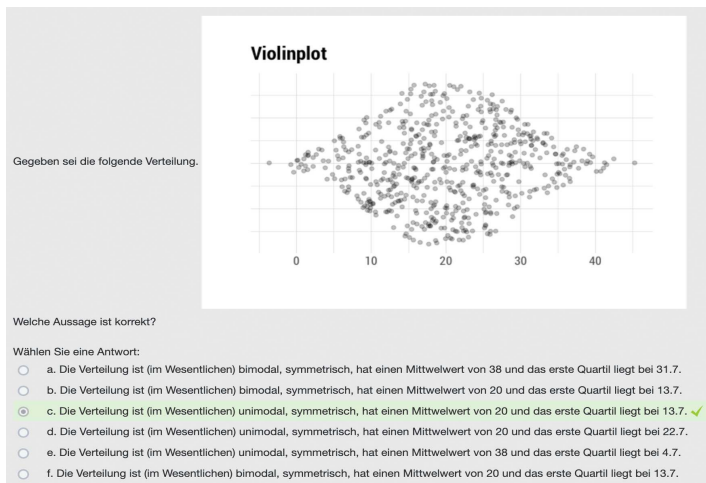
strong evidence for  
effectiveness of the testing  
effect



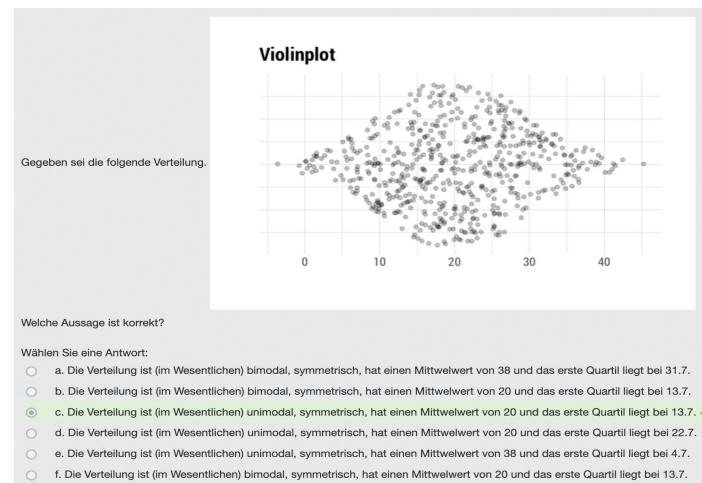
intensive use of tests  
by students and  
teachers

# Material: Task sets for self-testing

## ► Example question within parametrized task set



## ► Example question within fixed task set



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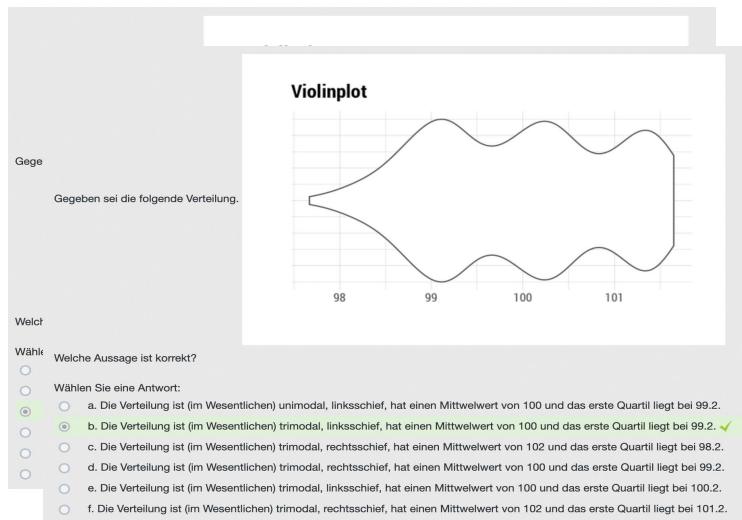
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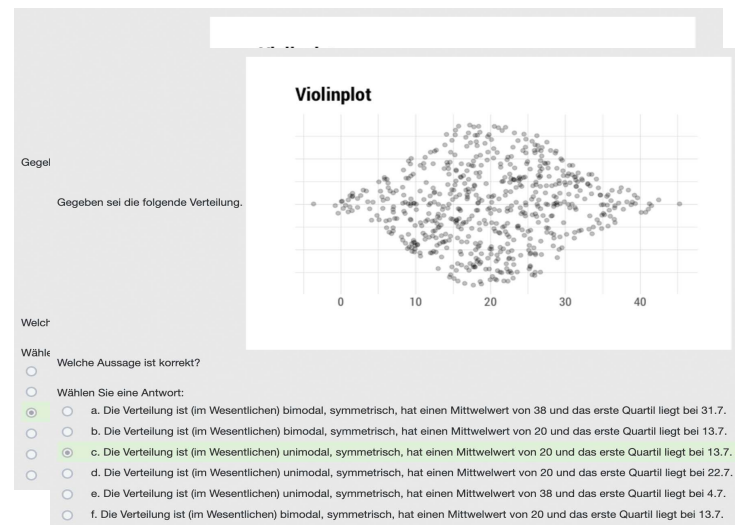
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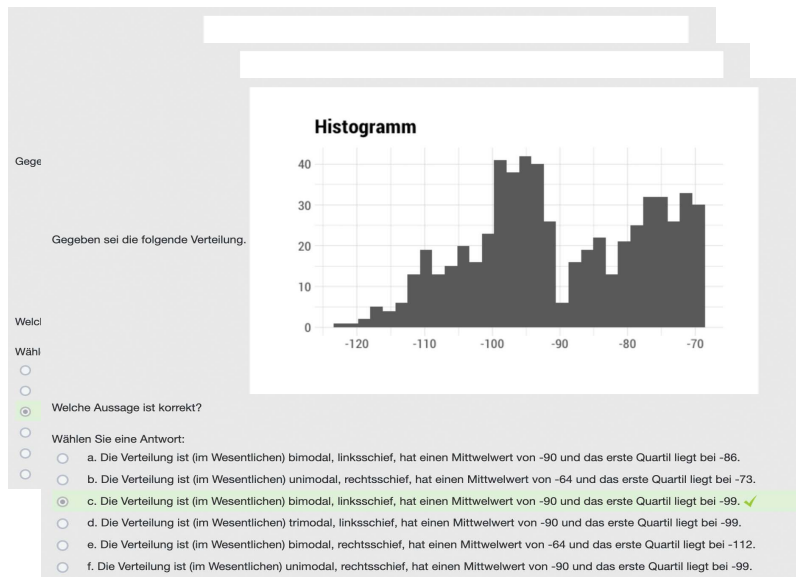
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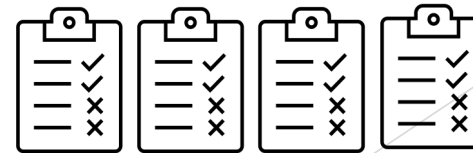
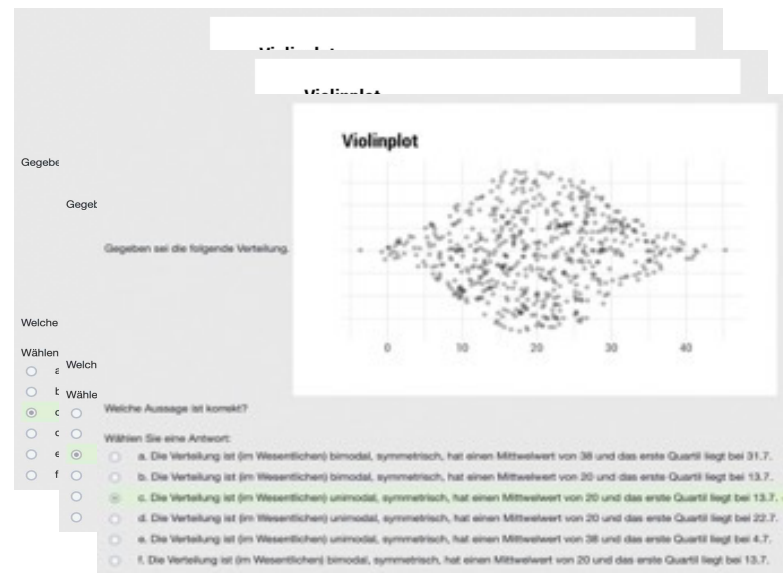
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# Material: Task sets for self-testing

## ► Example question within parametrized task set



## ► Example question within fixed task set



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# Hypotheses

1. The more intensively practice tests are used by students in an online course, the better learning results they achieve.
2. The use of parametrized test tasks leads to higher learning performance than the use of fixed task sets.



3. Parametrized tasks increase the students' situational interest compared to fixed task sets.



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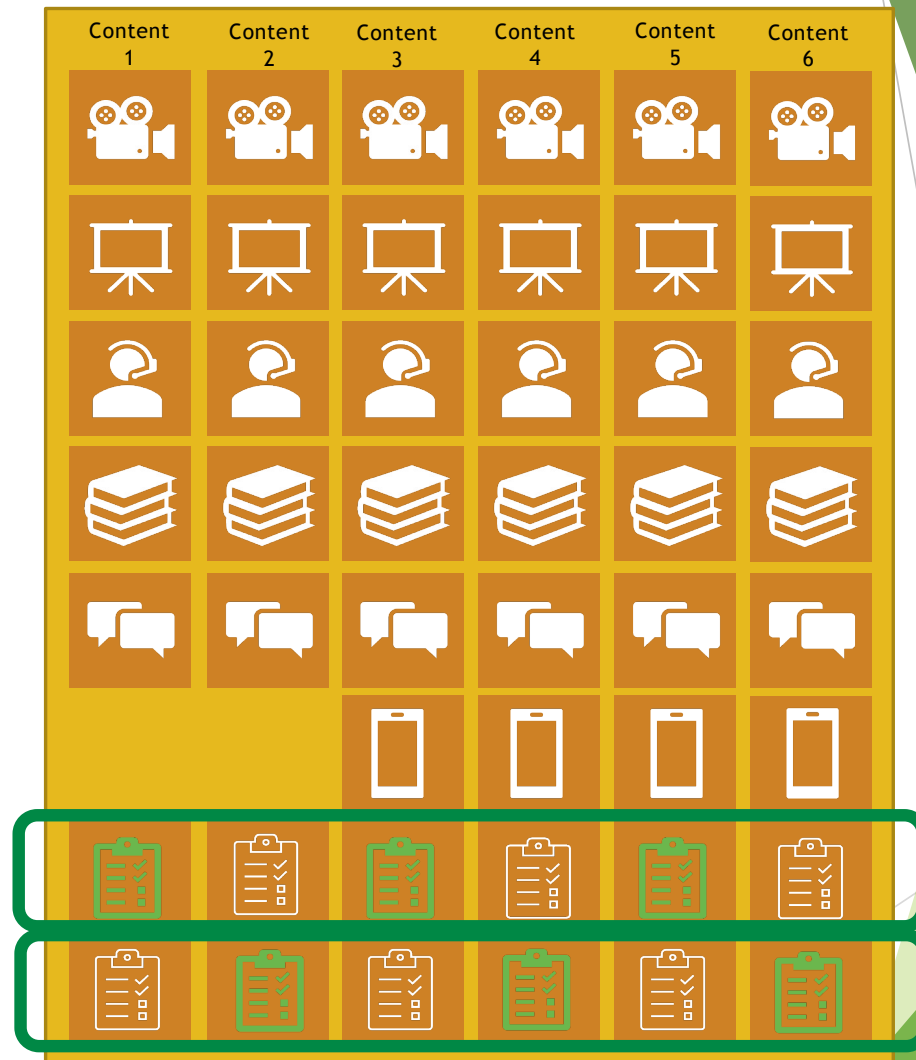
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# Methods

- between-group and within-person experiment
- log data analysis and additional questionnaires
- blended learning setting
- course on research methodology
- student teacher sample
- weekly formative, digital tests (multiple choice test)
- immediate feedback



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## (Preliminary) Results

- ▶ Sample:  $n=342$  student teachers (age 18-43, mean=21.69)
- ▶ Kendall's rank correlation tau (number of attempts and result in exam)
- ▶  $z = 11.771$ ,  $p\text{-value} < 2.2e-16$ ,  $\tau = 0.4454374$

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## (Preliminary) Results

- ▶ Sample:  $n=342$  student teachers (age 18-43, mean=21.69)
- ▶ Kendall's rank correlation tau (number of attempts and result in exam)
- ▶  $z = 11.771$ ,  $p\text{-value} < 2.2e-16$ ,  $\tau = 0.44$



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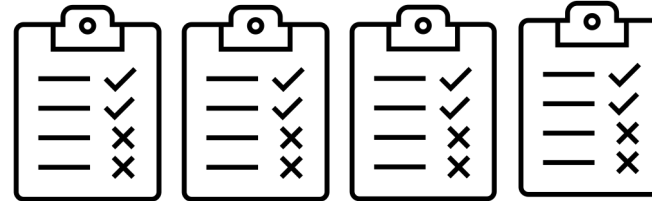
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parametrized  
task sets



>

fixed task sets



week	tau	p-value
1	-0.03	0.51
2	0.04	0.30
3	-0.05	0.28
4	0.08	0.06
5	-0.04	0.32
6	0.02	0.57

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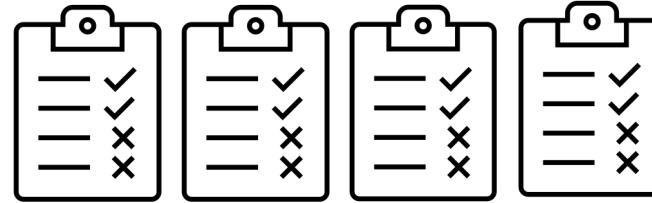
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?

fixed task sets



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# Situational interest



**value:** useful, unimportant (rev.), unnecessary (rev.)

**feeling:** exciting, boring (rev.), entertaining

## Random intercept model

<i>Predictors</i>	<b>value</b>		
	<i>Estimates</i>	<i>CI</i>	<i>p</i>
(Intercept)	21.71	20.99 – 22.44	<b>&lt;0.001</b>
Parametrisierung	-0.34	-0.98 – 0.31	0.304
<b>Random Effects</b>			
$\sigma^2$	13.78		
$\tau_{00}$ Pseudonym	12.93		
ICC	0.48		
N Pseudonym	167		
Observations	563		
Marginal $R^2$ / Conditional $R^2$	0.001 / 0.484		

<i>Predictors</i>	<b>feeling</b>		
	<i>Estimates</i>	<i>CI</i>	<i>p</i>
(Intercept)	14.35	13.61 – 15.09	<b>&lt;0.001</b>
Parametrisierung	0.29	-0.38 – 0.97	0.394
<b>Random Effects</b>			
$\sigma^2$	15.10		
$\tau_{00}$ Pseudonym	13.24		
ICC	0.47		
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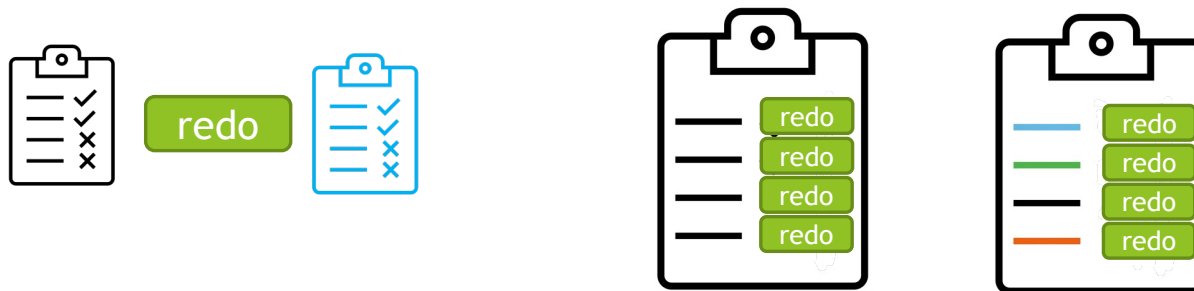
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# Discussion and next steps

- repetition on task set level vs. single task level



- proportion of correct responses



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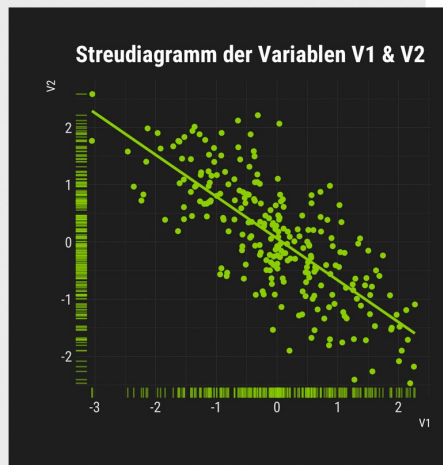
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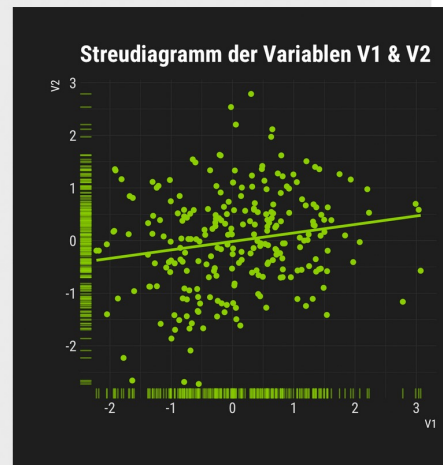
# More examples



Obige Abbildung zeigt die Assoziation der beiden Variablen V1 und V2. Welches Pearson's  $r(V1, V2)$  passt am besten zu dieser Abbildung?

Wählen Sie eine Antwort:

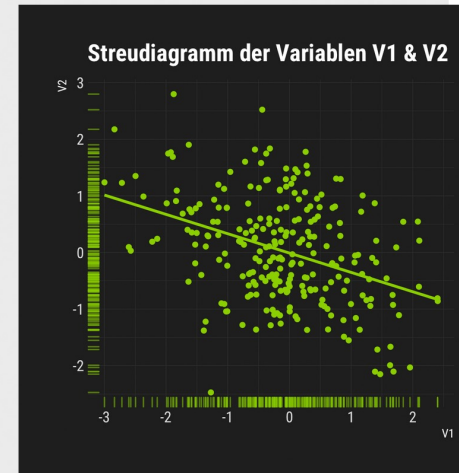
- ☐ a. 0.73
- ☐ b. 0.07
- ☐ c. -0.25
- ☐ d. 0.22
- ☒ e. -0.74 ✓



Obige Abbildung zeigt die Assoziation der beiden Variablen V1 und V2. Welches Pearson's  $r(V1, V2)$  passt am besten zu dieser Abbildung?

Wählen Sie eine Antwort:

- ☐ a. -1.09
- ☐ b. -0.16
- ☐ c. 0.06
- ☐ d. 1.1
- ☒ e. 0.13 ✓



Obige Abbildung zeigt die Assoziation der beiden Variablen V1 und V2. Welches Pearson's  $r(V1, V2)$  passt am besten zu dieser Abbildung?

Wählen Sie eine Antwort:

- ☒ a. -0.28 ✓
- ☐ b. 0.09
- ☐ c. 0.24
- ☐ d. 0.69
- ☐ e. -0.69



