

Code Structure Difficulty in OOP

an exploration study regarding basic cognitive processes

Background

- On the way to a competency model of OOP the most important code structures have been identified in former studies (see [7])

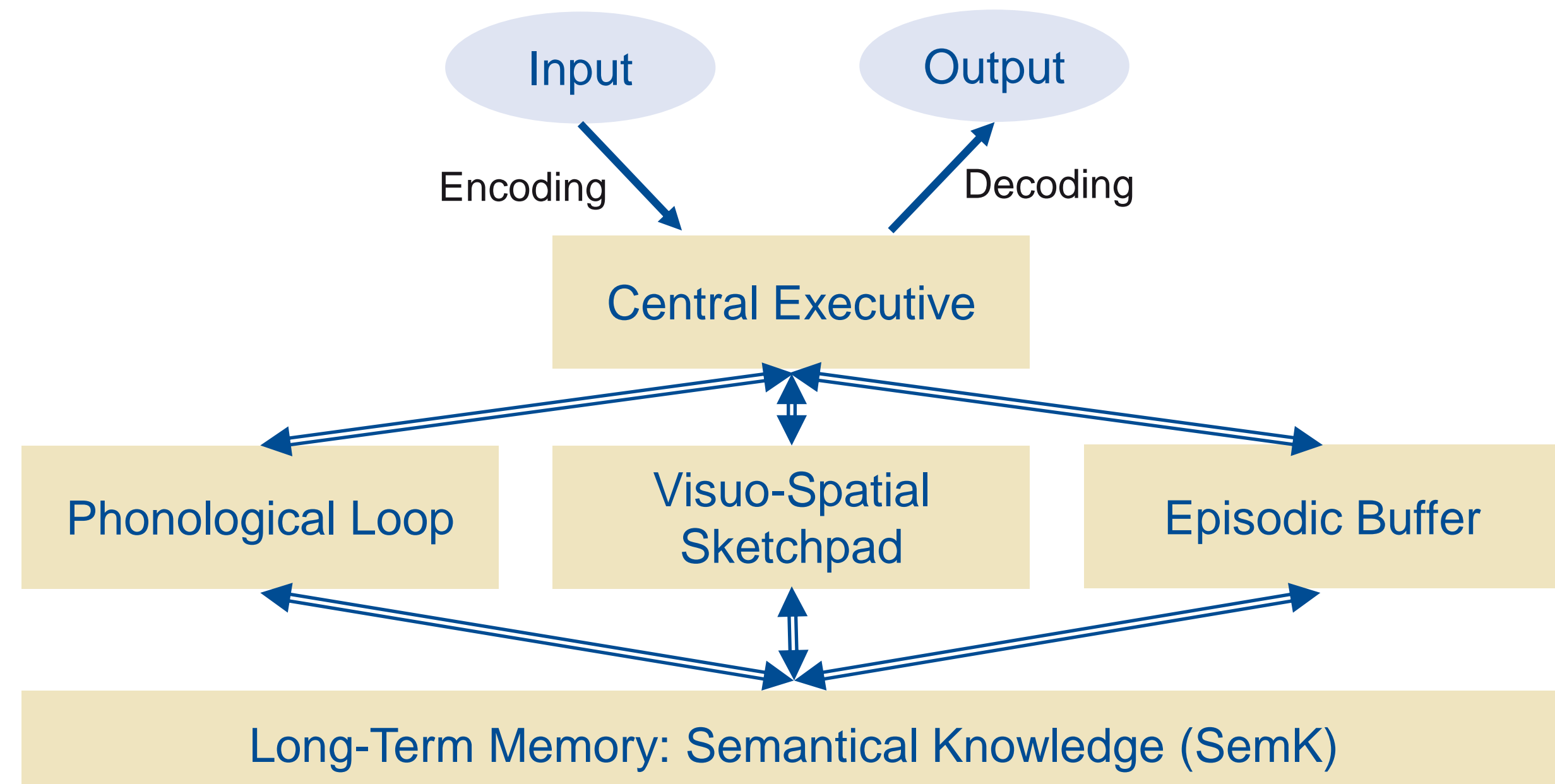


Fig: Working Memory Model with information processing (see [2, 9, 6])

- RQ:** Is there a difference in encoding or decoding of varied code structures?

Method

- Assessing cognitive capabilities to control their influence on the process

- Assessing the ability to decode and encode programming language texts (PLT) and natural language texts (NLT)

- Three levels with three items each were used
- The items of each level were comparable in length and cognitive complexity metrics (see [1, 3, 5, 8])
- Assessing basic demographic information and the self-estimated programming experience
 - How do you estimate your programming experience?
 - How experienced are you with the programming language Java?
 - How experienced are you with object-oriented programming?
 - How experienced are you compared to an expert with 20 years of experience?
 - How experienced are you compared to a fellow student?
- We divided the computer science students into the SemK-levels novices and experts using the self-estimated programming experience

Results

- 42 students (5 were excluded due to missing data; 10 female; 27 male; mean of age: 25.03; sd: 3.77)
- The covered Levenshtein-Distance [10] during decoding and the time spent on encoding were used to determine differences between the control group (CG), novices (N) and experts (E)

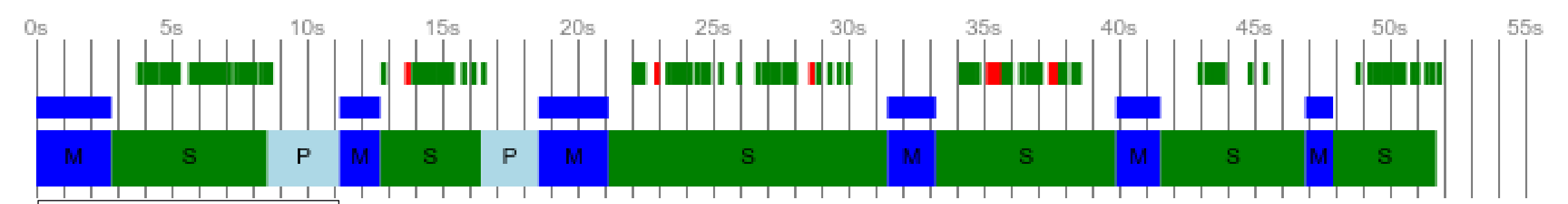


Fig: Timeline of a process
1. Row: Keystrokes (red = Backspace or Delete), 2. Row: Encoding
3. Row: Encoding (M), Decoding (S) und Pause (P)

- The initial phase is a good indicator for the whole process, because it is not affected by a learning effect
- We analysed the differences between the groups by conducting a Whitney-U-Test

	Novices (n = 16)			Experts (n = 15)		
	PLT-C	NLT	PLT-A	PLT-C	NLT	PLT-A
Control Group (n = 6)	IL1		LVD**	LVD*		LVD*
	IL2	E**LVD*	E*	E*D*LVD*	E*	E**D*LVD*
	IL3	LVD*		LVD*		LVD*

Tab: Differences in initial behaviour of the three groups

Note: * p < .05; ** p < .01; *** p < .001
PLT-C: Class Item, NLT: NLT Item, PLT-A: Algorithm Item, IL: Item-Level
LVD: Levenshtein-Distance, E: Encoding Time, D: Decoding Time

- Items of IL2 are best suitable for the measurement of differences between the groups

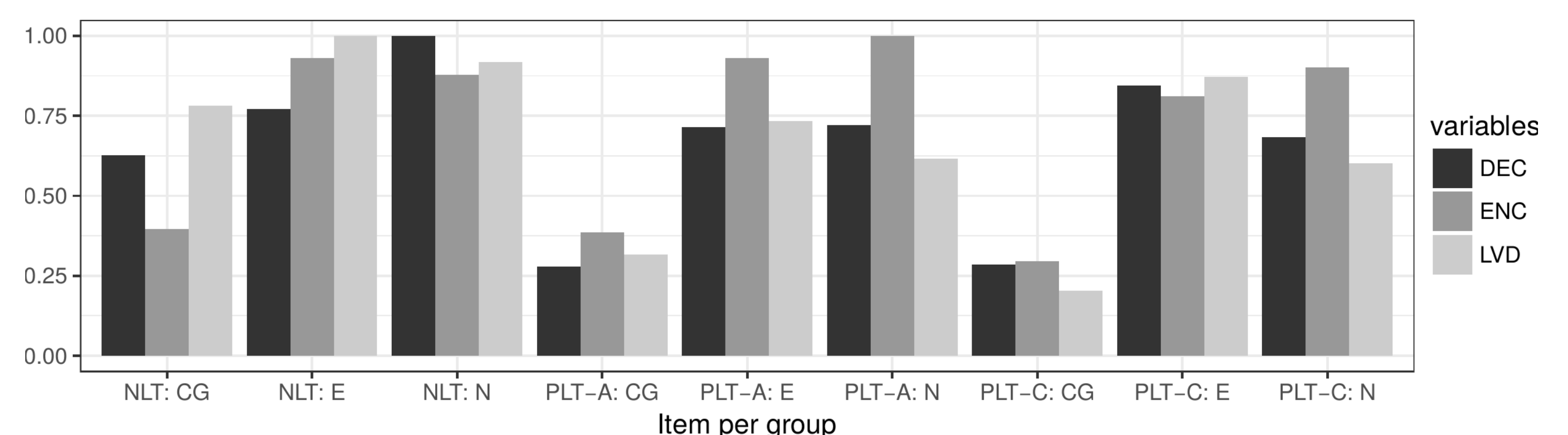


Fig: Normed values of item level 2

Fig: Items of level 2

Discussion and Outlook

- There seems to be an overall positive effect of SemK on encoding and decoding processes of PLT
- The processing time needs a deeper analysis, because there might be an inference between encoding complex code and SemK
- A follow-up study with new items comparable to those of IL2 seems promising to unravel the interpreted underlying interaction of SemK and code structure complexity
- Further details of the procedure and analysis are in [4]

References

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