

# Driving fitness in the elderly – validation of visual and driving-aptitude tests

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## Introduction

- By 2050, the number of people in the EU aged 65 and above will have increased by 70%, and over 80 by 170%.<sup>1</sup>
- Mobility is key in facing challenges of demographic change, for independent living, and for promoting health and quality of life.
- Safe driving requires visual and cognitive abilities. The present study aims at a validation of apparatus and methods of testing vision and cognitive aptitude, with driving competence as the criterion of validity.

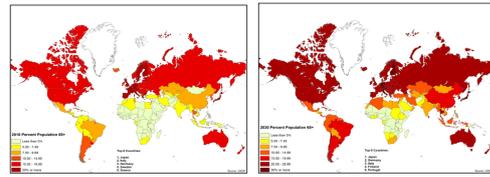


Figure 1. Old-age dependency ratios 2010 (left) und 2030 (right)<sup>2</sup>

## Methods

From May 2004 to February 2005, cognitive, visual and road driving tests were conducted in elderly drivers in Bad Tölz (Germany). Driving-specific abilities were tested by a standardized test battery (“Standard Plus”) in the “Expert System Traffic” (Schuhfried, Austria). Visual diagnostics included visual acuity, visual field, and contrast sensitivity. Results of psychological and visual tests were used as statistical predictors of variance in subjects’ driving performance.

## Psychological Testing

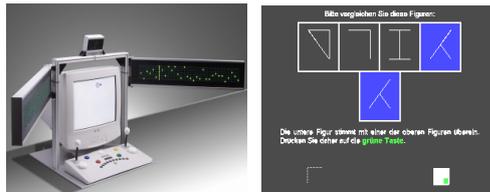


Figure 2. Psychological testing on the Vienna Test System (VTS, Schuhfried, Austria): “Peripheral Perception (PP)”: dynamic visual field and selective divided attention (left), and selective focused attention (right).

## Visual Diagnostics



Figure 3. Manual kinetic perimetry on an Octopus 101 (left), recognition contrast-sensitivity on a standard PC (R\_Contrast, right).



Figure 4. Visual acuity measured on an Oculus Binoptometer.

## On-Road Driving Test

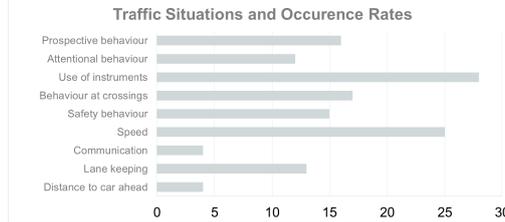


Figure 5. The road driving test lasted ~45 minutes and comprised 134 defined situations that were rated by both an expert and a trained rater. Inter-rater reliability  $\rho=0.59-0.79$  (Spearman). Global driving performance rated on a six-point scale, with “1” denoting the best score.

## Subject Sample

- Sample of 92 drivers (60 m, 32 f). Mean age 68.5 y (range 60–91 y; median 67 y; SD 6.6 y).
- Participants were volunteers with valid driving license; normal visual fields only.

## Results

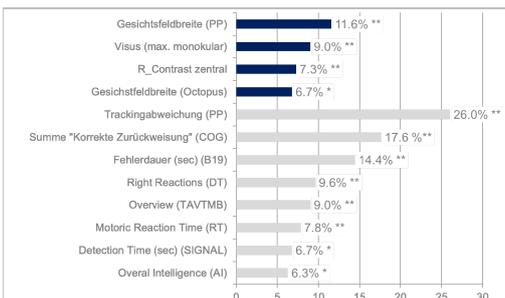


Figure 6. Visual and cognitive performance indicators that correlated significantly with driving competence (Explained Variance, \*\* $p=0.01$ , \* $p=0.05$ ).

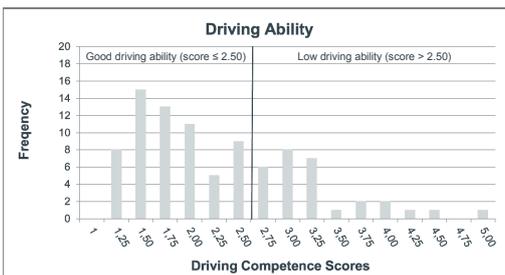


Figure 7. Driving ability was derived from driving competence by a sample-split at score 2.5. Ranges below 2.5 were interpreted as low driving ability. 25 drivers failed the hypothetical license test.

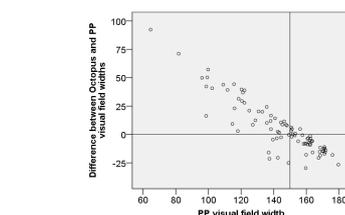


Figure 8. Difference between the two measures of visual-field width (Octopus 101 and PP on the VTS) vs. the PP measure. Horizontal line would indicate equality. Only at 150° are the measures equal. Correlations between measures are low ( $r=0.285$ ,  $p=0.007$ ,  $n=87$ ).

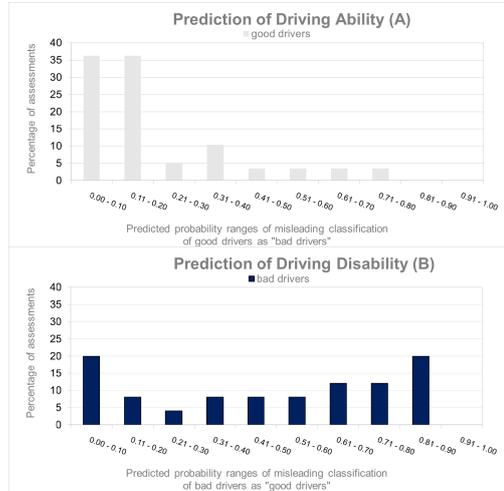


Figure 9. Prediction of driving ability (A) and disability (B).

- The set of significant predictors was reduced by stepwise regression with backward elimination.
- Binary logistic regressions were used for further analysis.
- Logistic regression allows deriving probabilities for driving ability.

	Predicted Driving Ability		
	Good dr. (score $\leq 2.5$ )	Bad dr. (score $> 2.5$ )	percentage of correct classifications
Driving ability	53	6	89.8
Bad drivers (score $> 2.5$ )	13	12	48.0
Total percentage			77.4

Figure 10. Classification rates for driving ability.

- While specificity is good (identification of good drivers: 90%), sensitivity is low (identification of bad drivers: 48%).
- 13 out of 25 persons with low driving ability were classified incorrectly.

	regression coefficients	Wald	df	Sig.
Focused attention (COG)	-.177	4.553	1	.033
Divided attention (PP)	.278	7.618	1	.006
Recognition time (SIGNAL)	.755	.269	1	.604
Dynamic visual field (PP)	.007	.254	1	.614
Visual acuity (Visus)	-.317	1.591	1	.207
Central contrast sensitivity (R_Contrast)	.360	.726	1	.394

Figure 11. Prediction of driving performance by the battery. Selective and divided attention were the only significant measures.

## Conclusion

- Visual performance indicators have only limited predictive power for driving aptitude (20% explained variance, EV); 80% are person-specific.
- Psychometric tests are more important (35% EV).
- Best predictors were lane tracking (26% EV), Schuhfried’s dynamical peripheral vision PP (12% EV).
- The label “visual field” for PP on the VTS is misleading.
- Acuity (9%) and perimetry (7%) are of little importance.