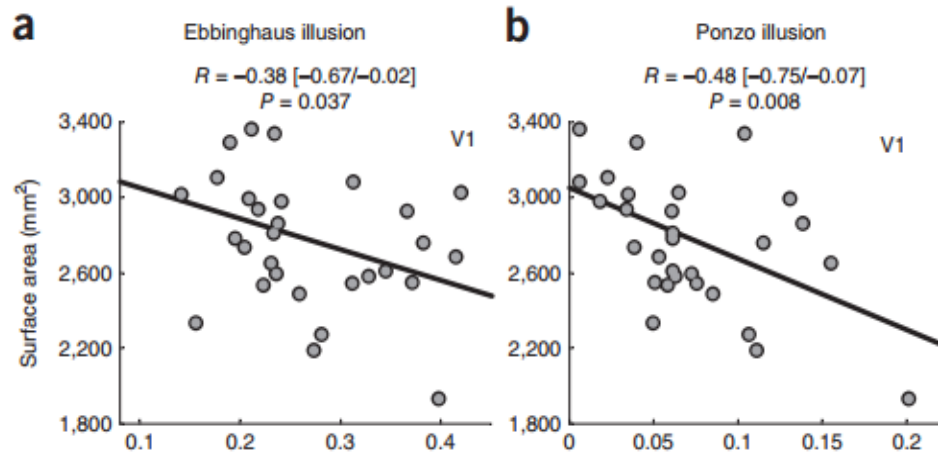
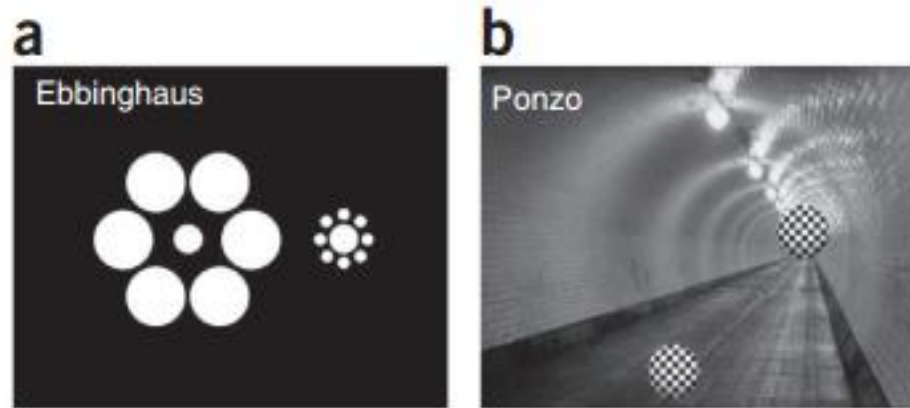


# Test Validity

# Illusions & V1 size



Schwartzkopf et al., 2011 (n = 30)

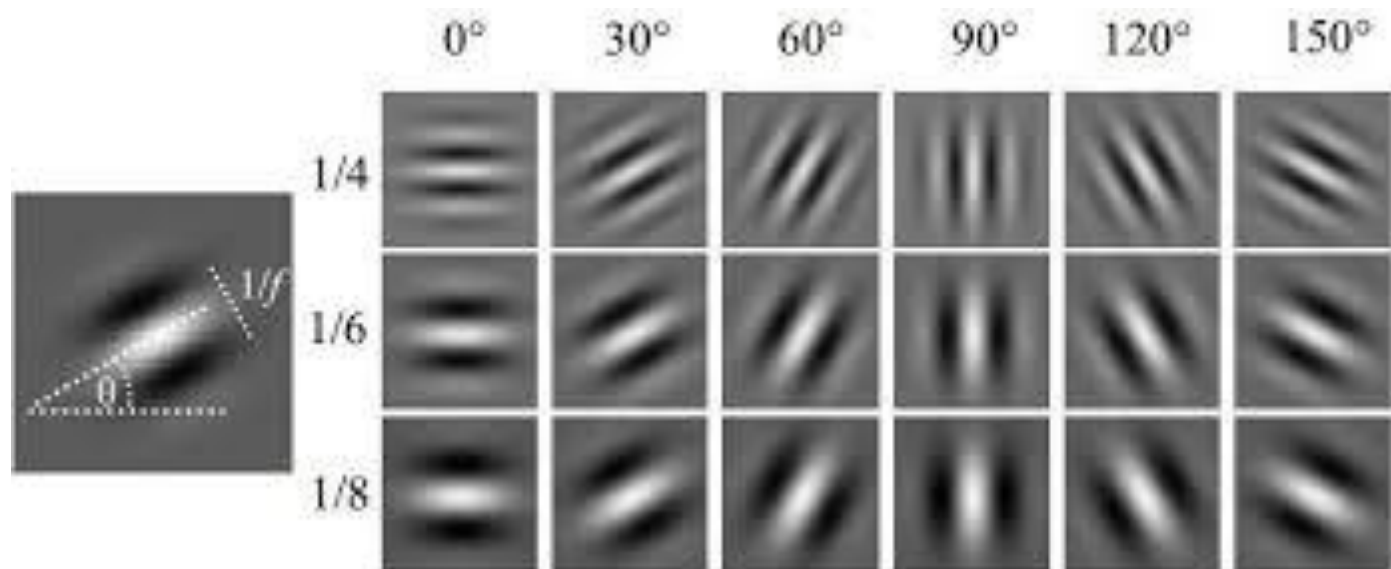
# 1. Tests



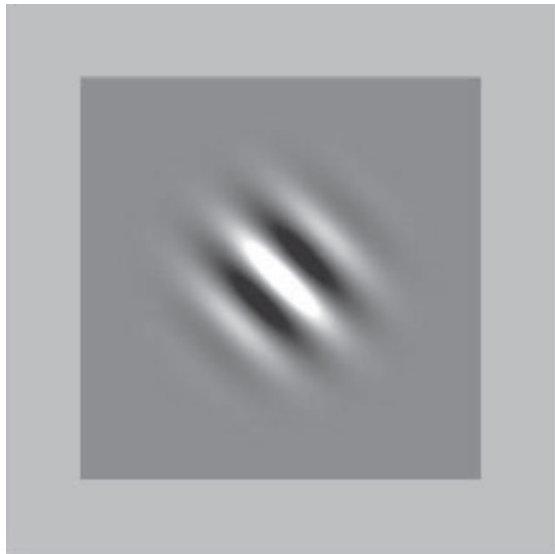
Contrast sensitivity



Contrast sensitivity



# Why this test? Significant result



Contrast sensitivity

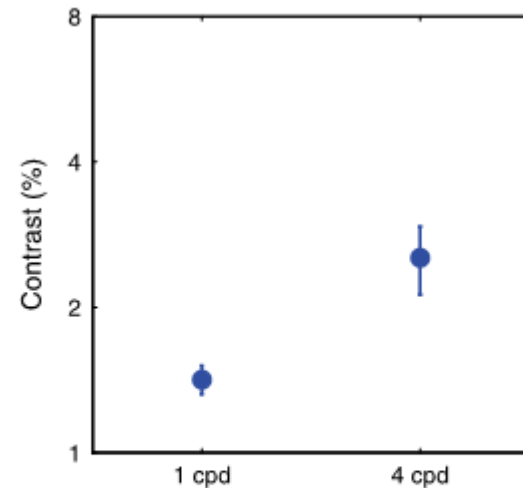
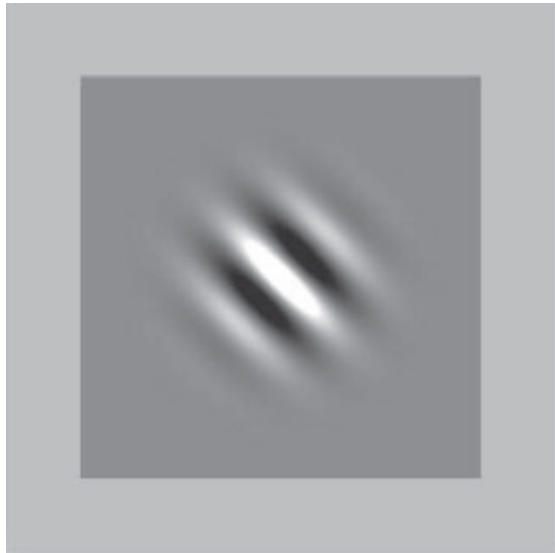


Figure 1. Contrast thresholds for older participants (red squares) and younger participants (blue circles) for 1 and 4 cpd Gabor patterns. Error bars represent  $\pm 1$  SEM across 10 participants.

# Why this test? Significant result



Contrast sensitivity

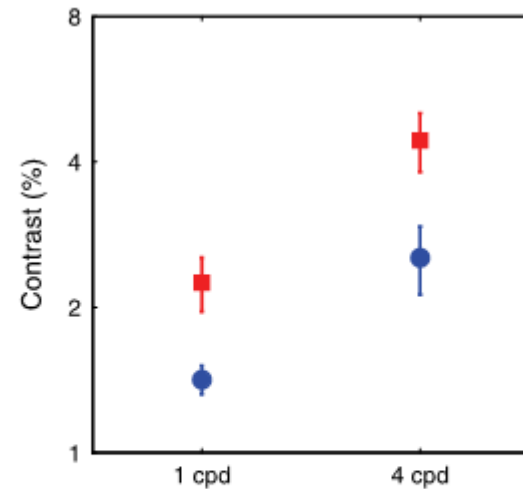
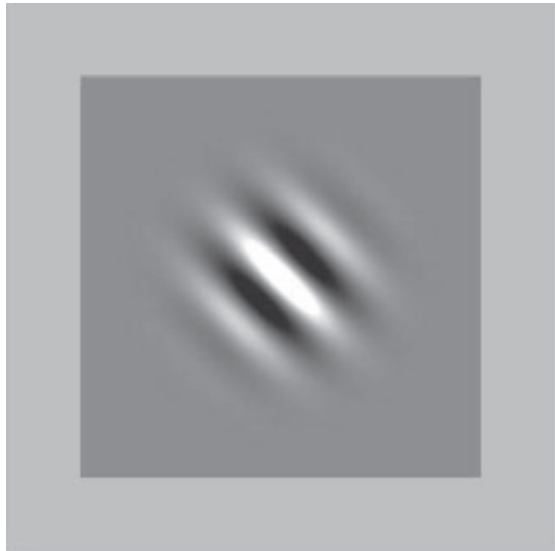
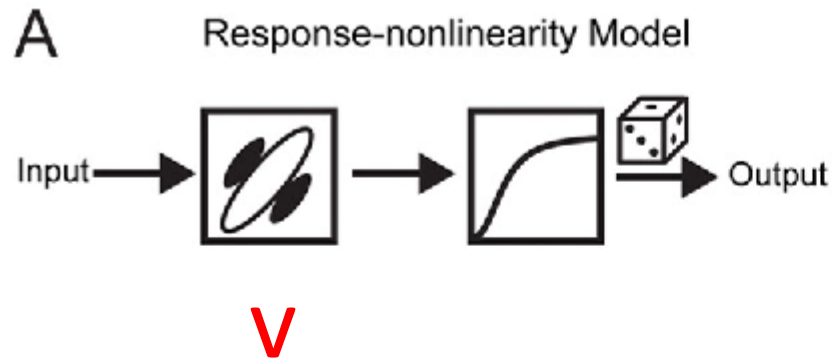


Figure 1. Contrast thresholds for older participants (red squares) and younger participants (blue circles) for 1 and 4 cpd Gabor patterns. Error bars represent  $\pm 1$  SEM across 10 participants.

# One test is usually used



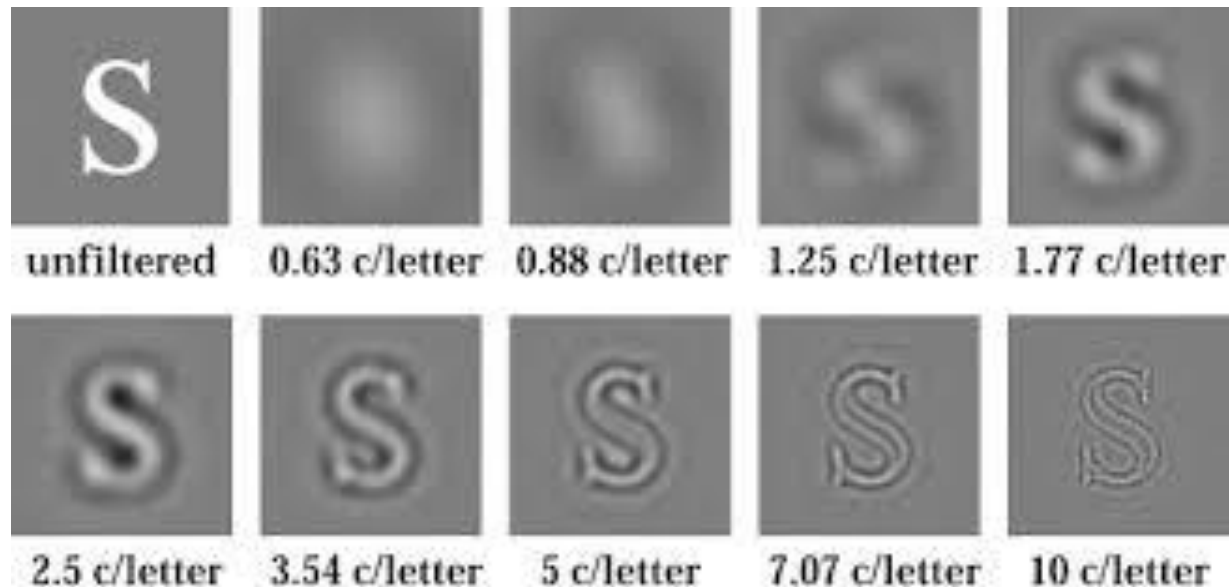
Contrast sensitivity



One test is sufficient because  
one mechanism/factor is at work

Why **one** test?

**Similar** tests do the same job and, hence, **correlate** with each other



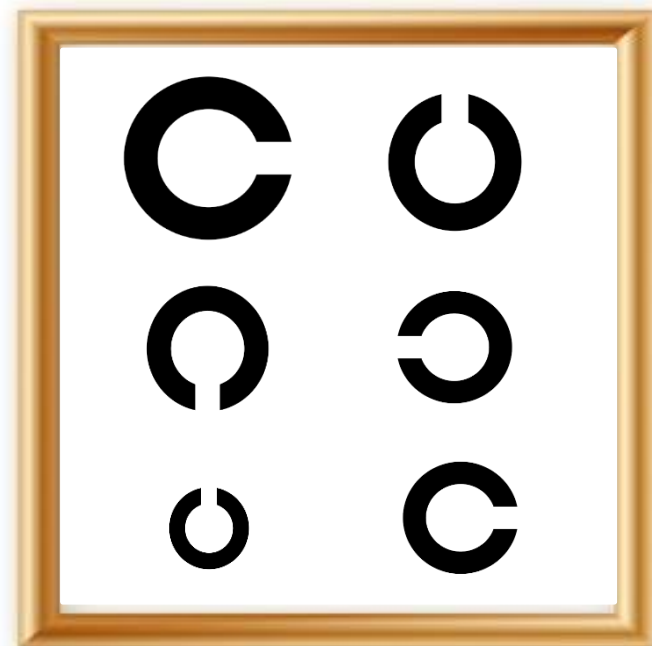
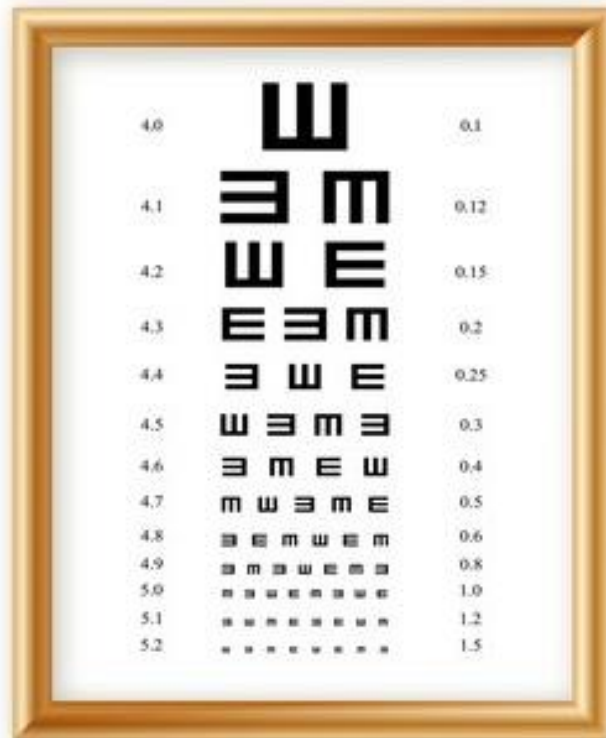
# Assumptions

Why **one** test?

- **One** mechanism
- **Similar** tests do the same job and, hence, **correlate** with each other

Significance justifies use of test

Tests are **useless** when they do **not**  
**correlate** with each other



2. Vision tests do **not** correlate with  
each other

## 2.1. Spatial Tests

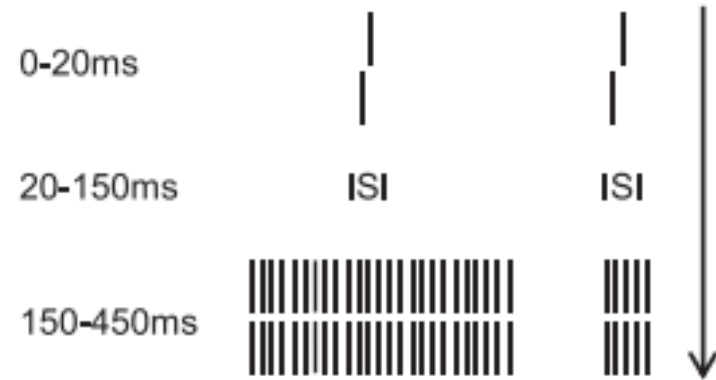
**A. Freiburg visual acuity test**



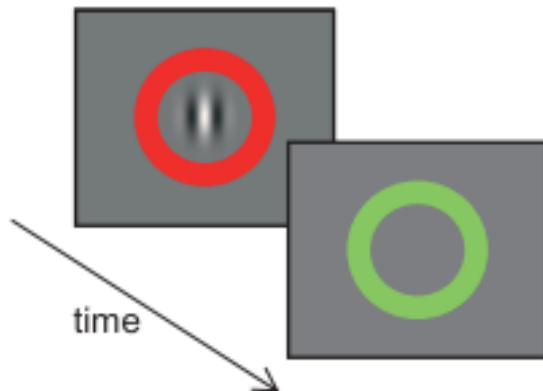
**B. Vernier offset discrimination**



**C. Visual backward masking with 5 and 25 elements**



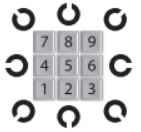
**D. Gabor detection**



**E. Bisection discrimination**



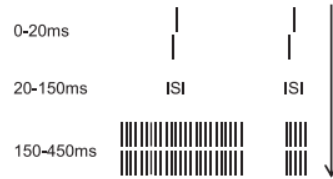
A. Freiburg visual acuity test



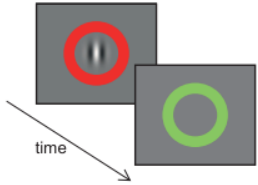
B. Vernier offset discrimination



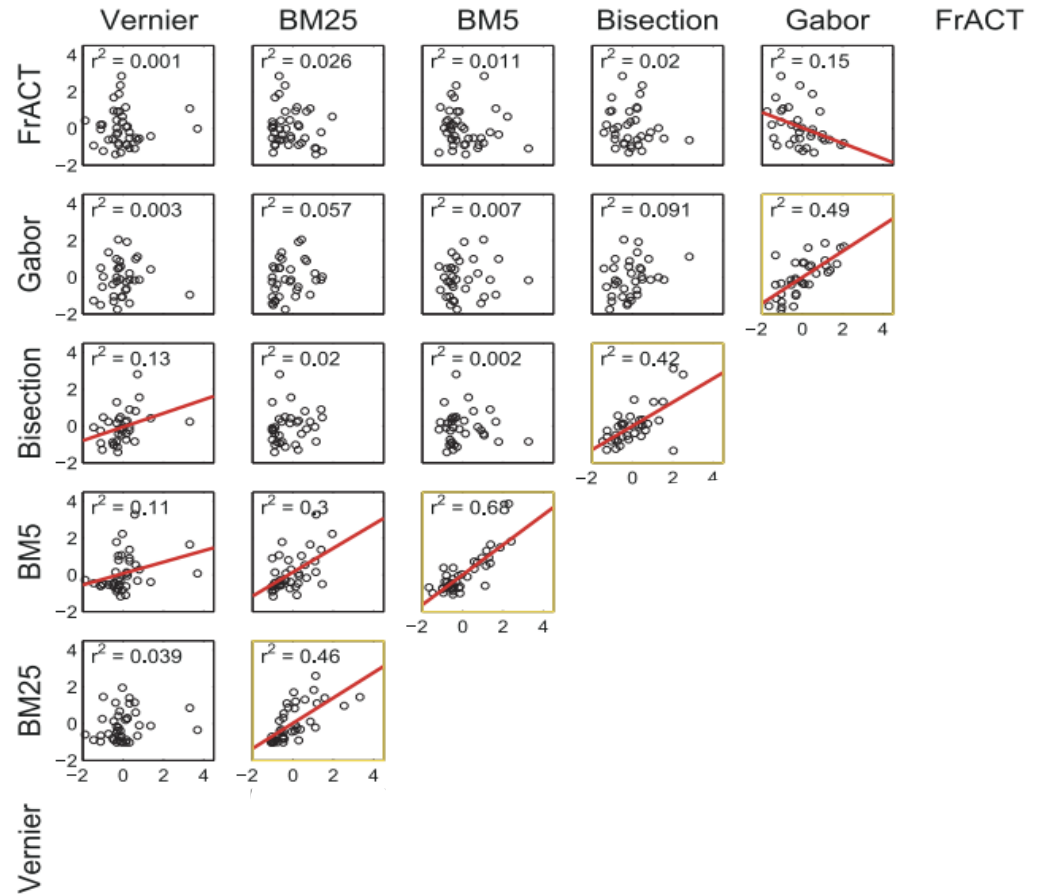
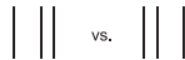
C. Visual backward masking with 5 and 25 elements



D. Gabor detection

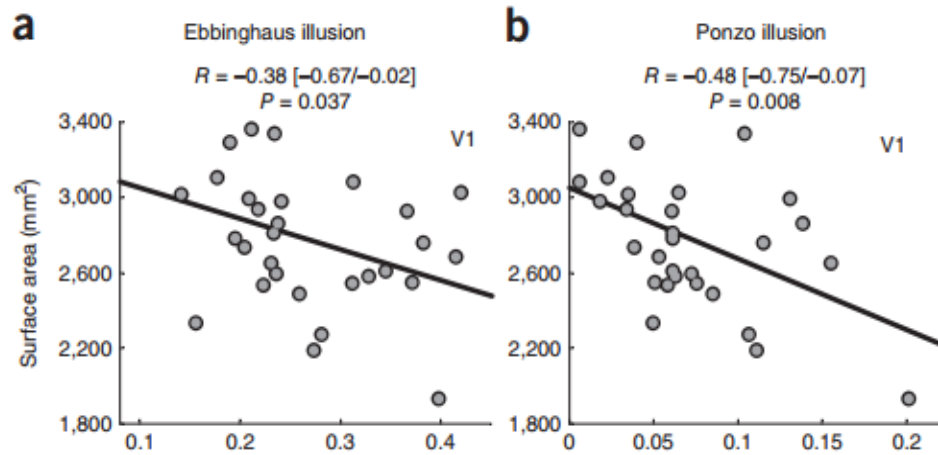
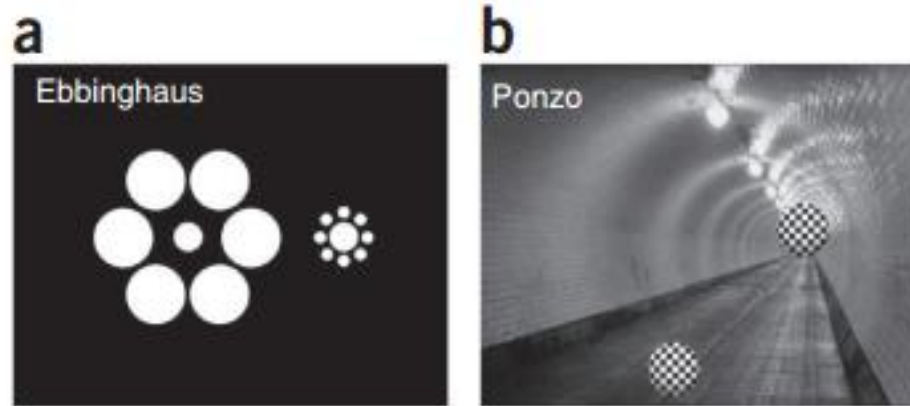


E. Bisection discrimination



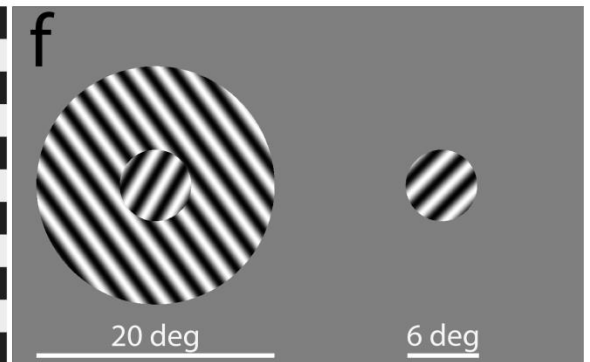
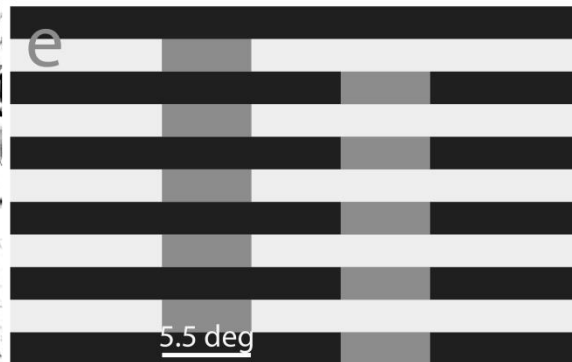
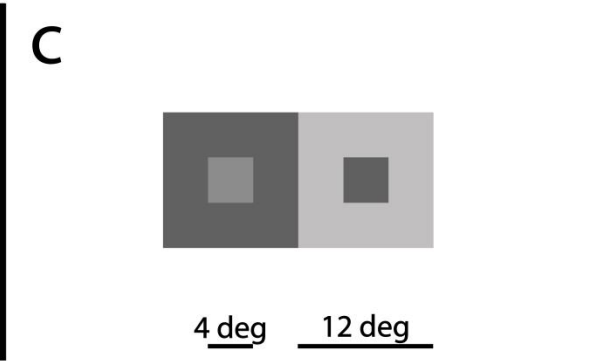
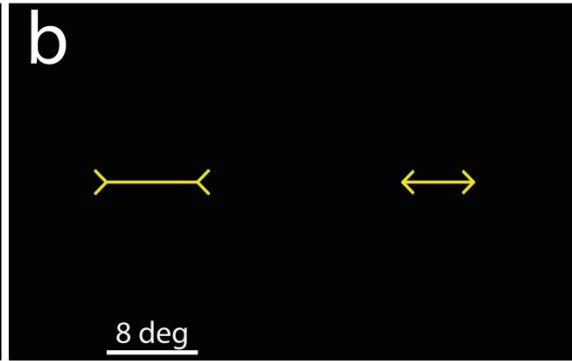
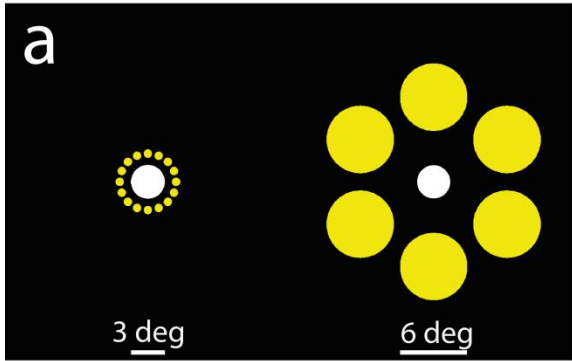
## 2.1. Spatial Illusions

# Illusions & V1 size

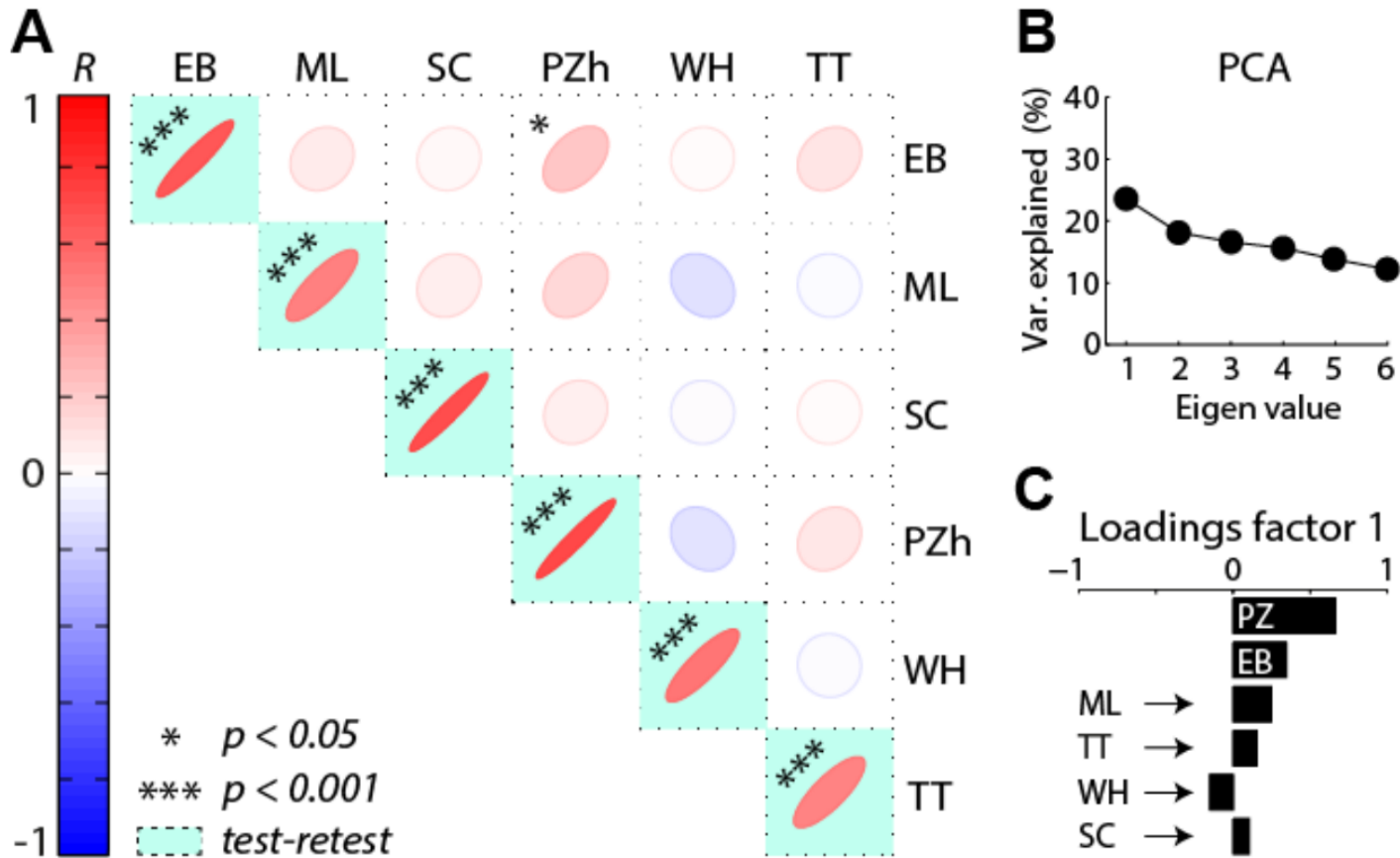


Schwartzkopf et al., 2011 (n = 30)

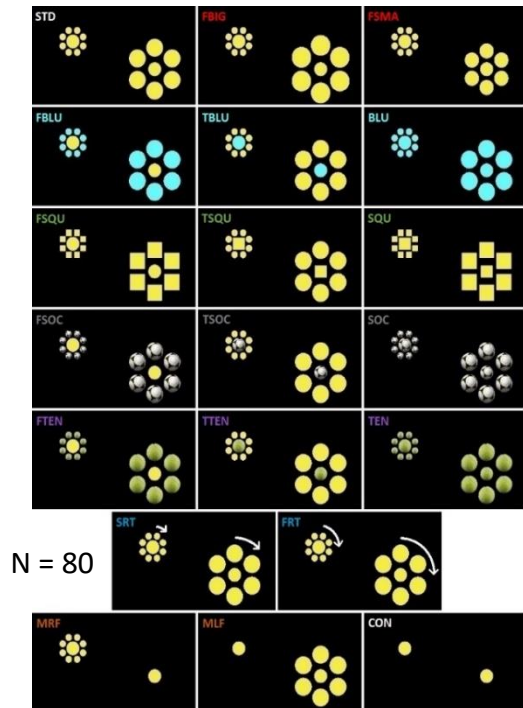
# Illusions (n=144)



# Illusions - correlations



# Ebbinghaus factor



	STD	FBIG	A	FBLU	TBLU	FSQ	HSQ	U	SQU	FSQC	C	SOC	FTEN	TTEN	TEN	SRT	FRT	MRF	MLF
STD	.59	.52	.39	.56	.40	.41	.38	.39	.29	.34	.50	.43	.48	.29	.48	.44	.30	.23	
FBIG	.93	.72	.50	.60	.63	.49	.50	.33	.59	.45	.37	.57	.55	.35	.43	.63	.56	.33	.36
FSMA	.96	.84	.50	.46	.54	.30	.34	.38	.57	.36	.42	.62	.52	.30	.49	.52	.49	.27	.38
FBLU	.65	.86	.81	.67	.63	.60	.36	.28	.45	.52	.39	.45	.58	.35	.30	.45	.48	.30	.34
TBLU	.98	.97	1	1	.60	.52	.53	.44	.53	.50	.52	.61	.59	.47	.43	.55	.55	.29	.35
BLU	.67	.73	.53	.95	.86	.63	.42	.30	.40	.49	.38	.41	.46	.43	.36	.47	.47	.24	.30
FSQU	1	1	.84	.71	1	.91	.44	.46	.39	.43	.42	.43	.51	.61	.41	.49	.47	.21	.14
TSQU	.72	.48	.74	.39	.76	.48	1	.59	.46	.36	.32	.46	.39	.45	.31	.33	.27	.25	.29
SQU	.74	1	1	.79	1	.72	.95	.89	.50	.45	.46	.65	.55	.41	.58	.51	.39	.40	.41
FSOC	.46	.63	.61	.77	.79	.75	.90	.56	.77	.68	.47	.40	.54	.29	.40	.58	.49	.18	.30
TSOC	.57	.51	.74	.56	.84	.57	.87	.48	.81	.71	.65	.58	.58	.47	.45	.48	.46	.37	.22
SOC	.89	.89	1	.70	1	.66	.95	.79	1	.63	.95	.60	.57	.50	.54	.54	.50	.33	.25
FTEN	.79	.86	1	.94	1	.76	1	.63	1	.87	.97	.99	.60	.51	.53	.55	.53	.27	.31
TTEN	.86	.52	.55	.54	.81	.71	1	.78	.76	.43	.77	.87	.90	.58	.49	.40	.49	.31	.16
TEN	.61	.69	1	.44	.78	.63	.85	.44	1	.68	.76	1	.97	.93	.60	.53	.42	.23	.40
SRT	.82	.96	.97	.69	.93	.76	1	.53	.94	.91	.76	.92	.96	.66	1	.61	.76	.10	.37
QRT	.76	.83	.88	.72	.90	.74	1	.37	.68	.75	.69	.81	.87	.82	.70	1	.65	.15	.24
MRF	.56	.49	.51	.44	.46	.37	.32	.35	.76	.24	.59	.54	.40	.51	.26	.10	.15	.57	.16
MLF	.38	.52	.69	.51	.55	.47	.23	.46	.73	.44	.31	.39	.49	.22	.71	.59	.35	.22	.63

→ EFA suggested 1 factor explaining ~44%

Cretenoud, A. F., Karimpur, H., Grzeczkowski, L., Francis, G., Hamburger, K., & Herzog, M. H. (2019). Factors underlying visual illusions are illusion-specific but not feature-specific. *Journal of Vision*, 19(14):12, 1–21.

## 2.3. Ageing



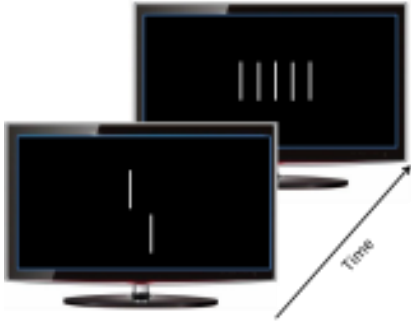
# Methodology

## Participants:

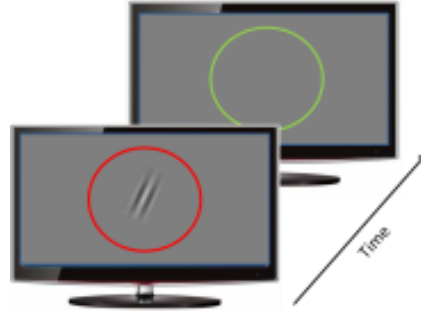
Group	Old	Young
Participants	131 (73 Males) 96 with MoCa>26	108 (48 Males) 106 kept
Age	69.8 Years Old (sd 6.76)	21.8 Years old (sd 2.69)
Recruited	TCS (driving tests/course) + Tbilisi, Georgia	EPFL (students) + Tbilisi, Georgia

# Perceptual tests

Vernier acuity & backward masking



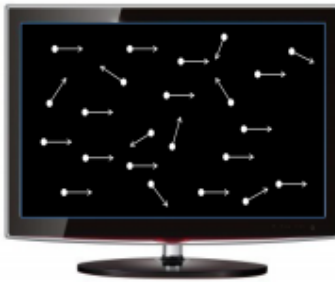
Contrast sensitivity (Gabor)



Biological motion perception



Motion direction discrimination



Simple RT Task



Orientation discrimination



Visual Acuity



Audiothreshold and pitch



Visual Search Task



Visual Simon Task

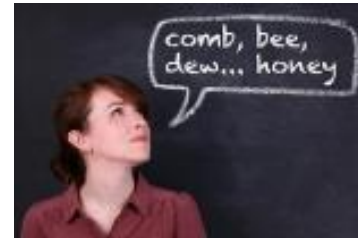


# Cognitive tests and questionnaires

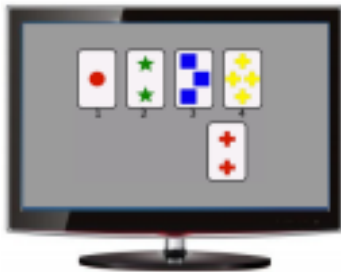
Digit Span forward and backward



Fluency



Wisconsin Card Sorting Test



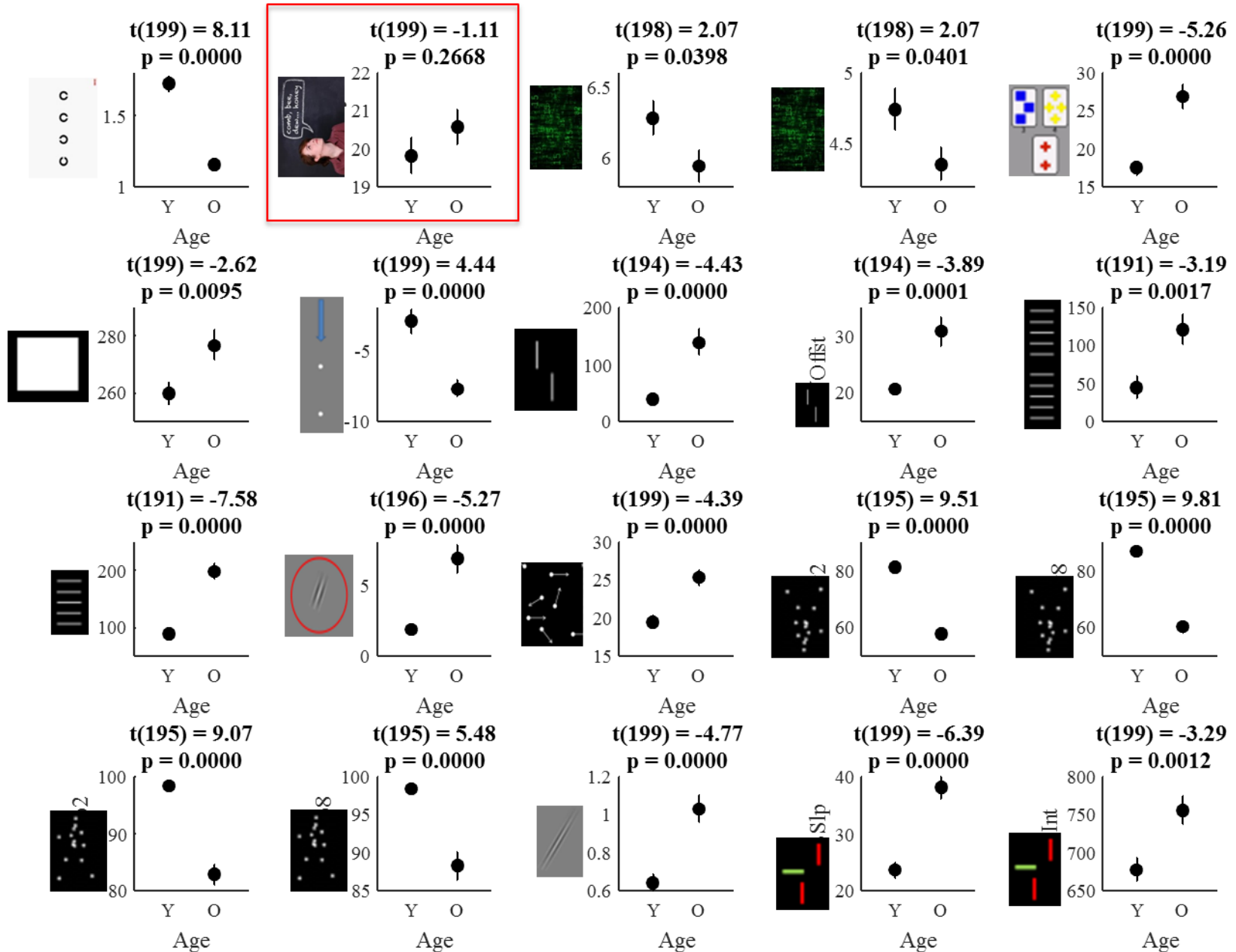
MoCa



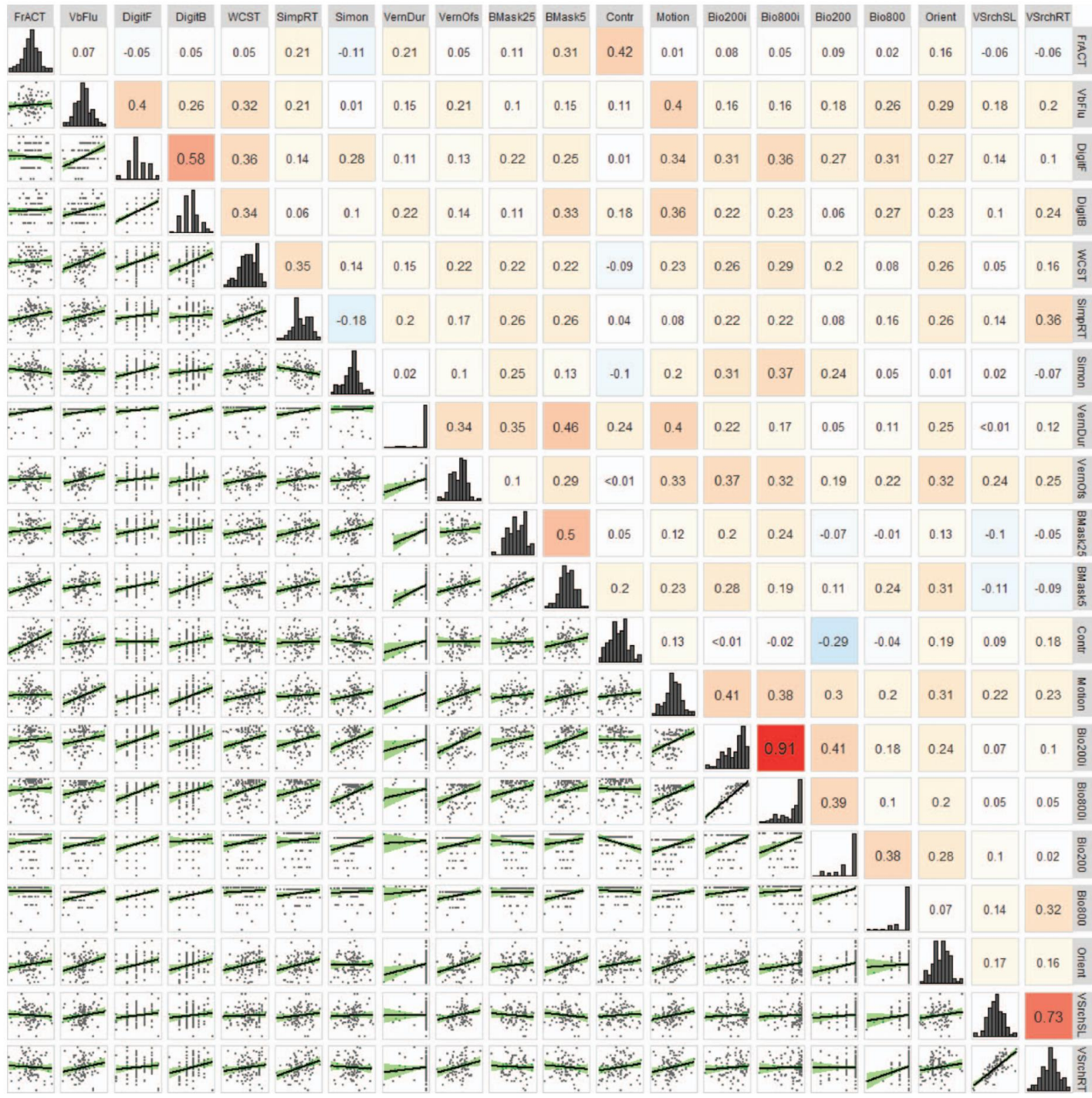
## Questionnaires:

1. General information
2. International physical ability
3. Autism quotient
4. Morning-evening questionnaire
5. Geriatric depression scale

# Differences young-old in all tests



# Correlations: young participants

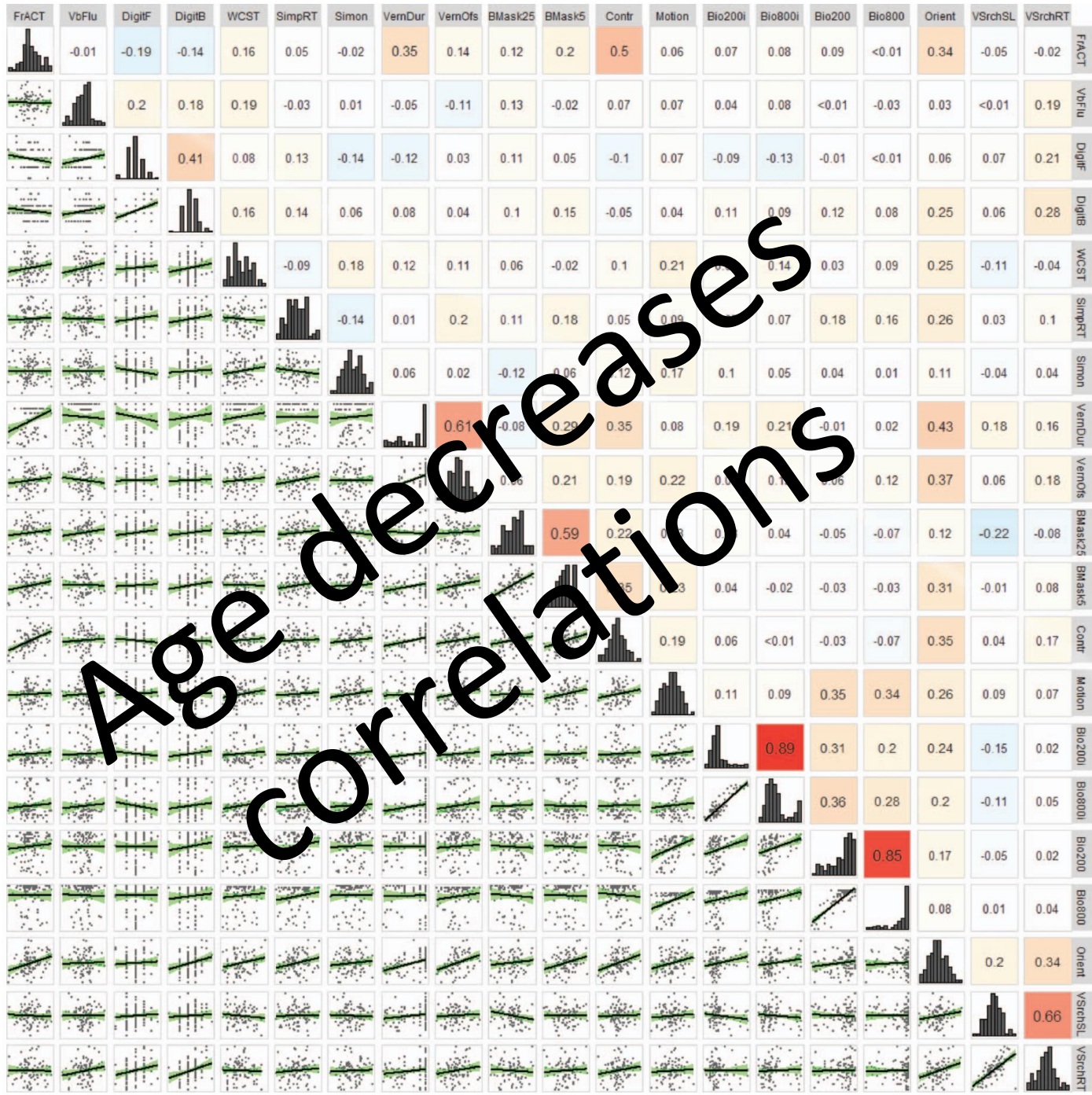




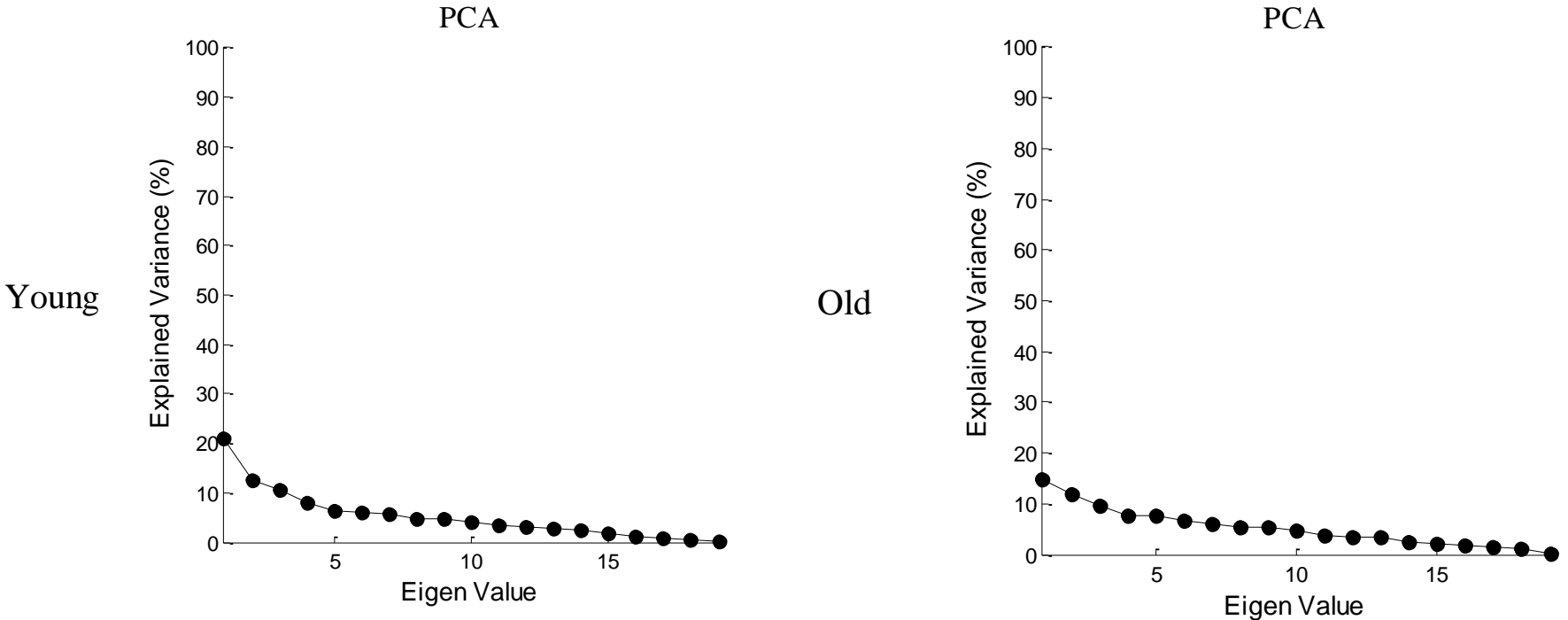




# Correlations: older participants



# PCA/Factor analysis



Hence, each task is a “factor” and each factor ages individually

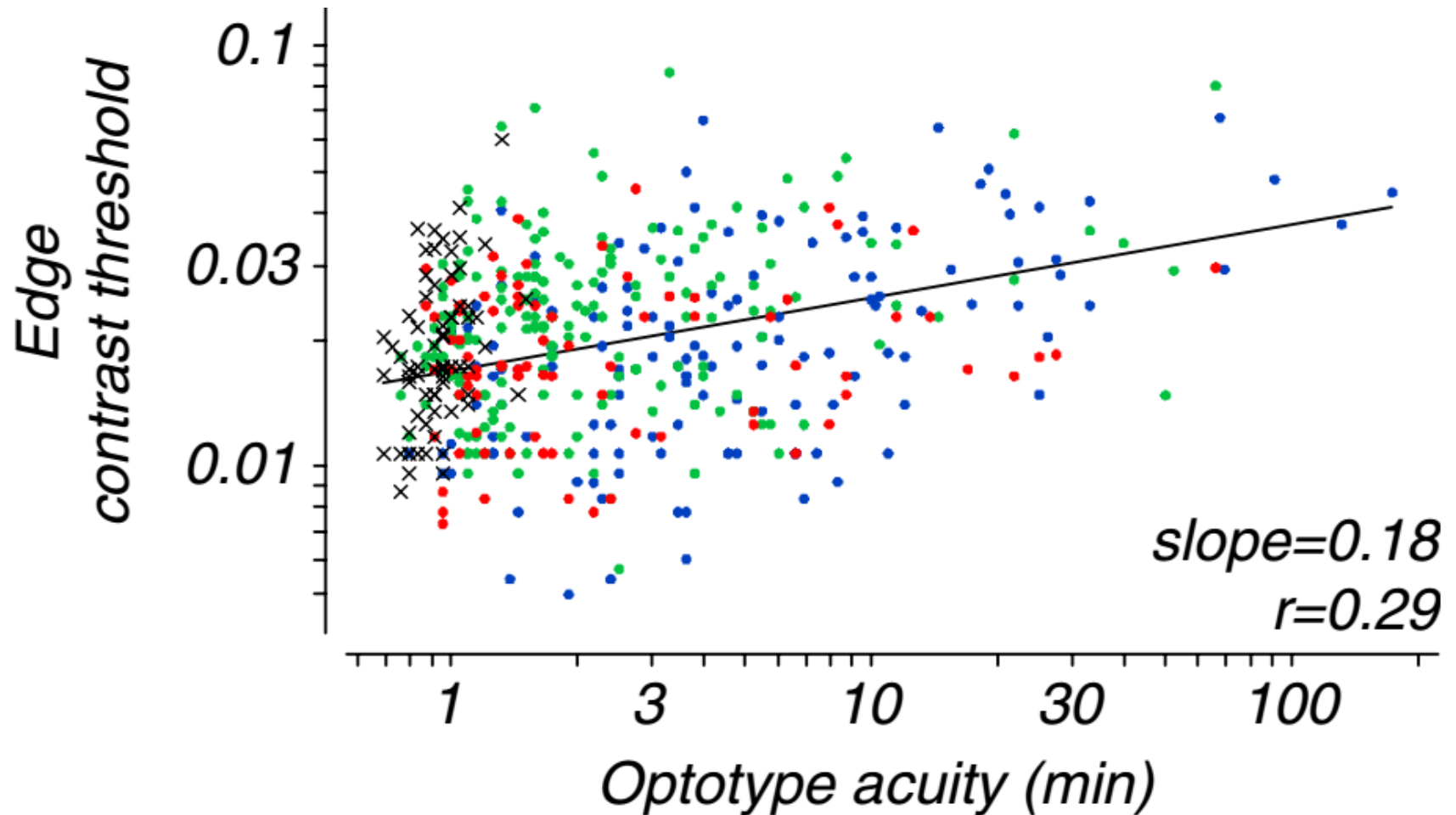


ÉCOLE POLYTECHNIQUE  
FÉDÉRALE DE LAUSANNE

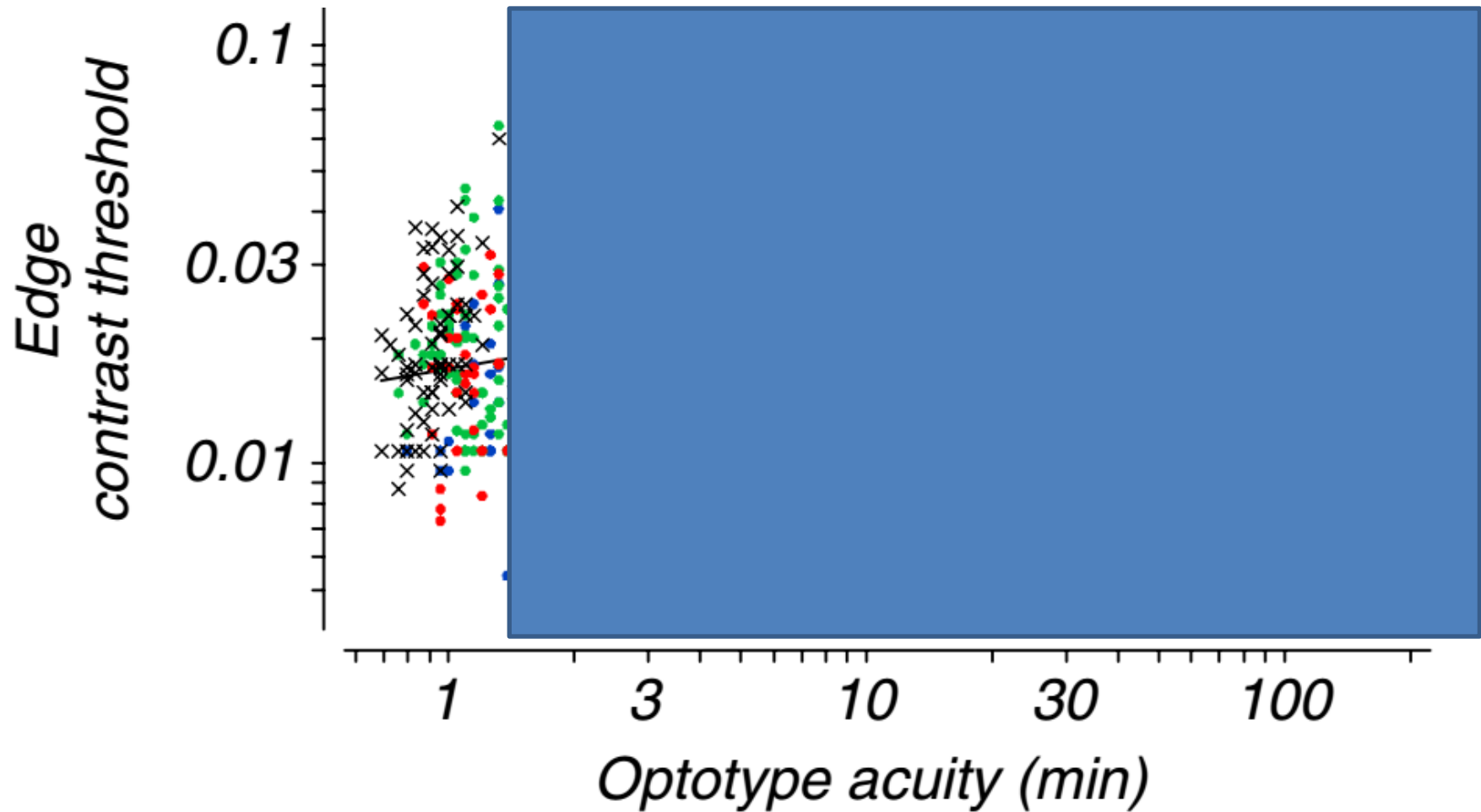


# Important: Healthy Ageing

# Reminder: Healthy aging



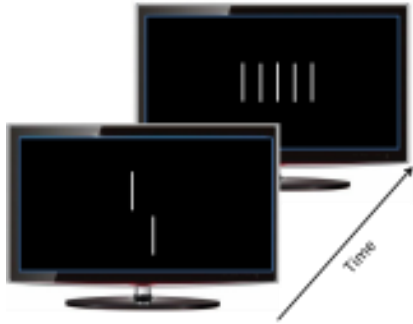
# Reminder: Healthy aging



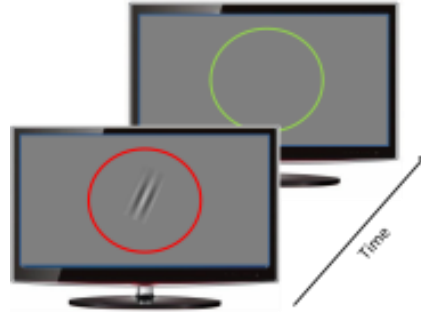
Aren't there older people with better perception than others?

# Aren't there older people with better perception than others?

Vernier acuity & backward masking



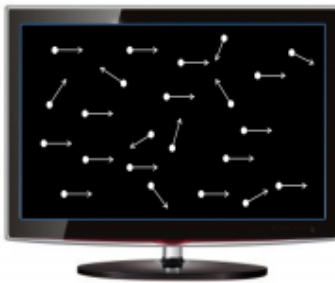
Contrast sensitivity (Gabor)



Biological motion perception



Motion direction discrimination



Simple RT Task



Orientation discrimination



Visual Acuity



Audiothreshold and pitch



Visual Search Task

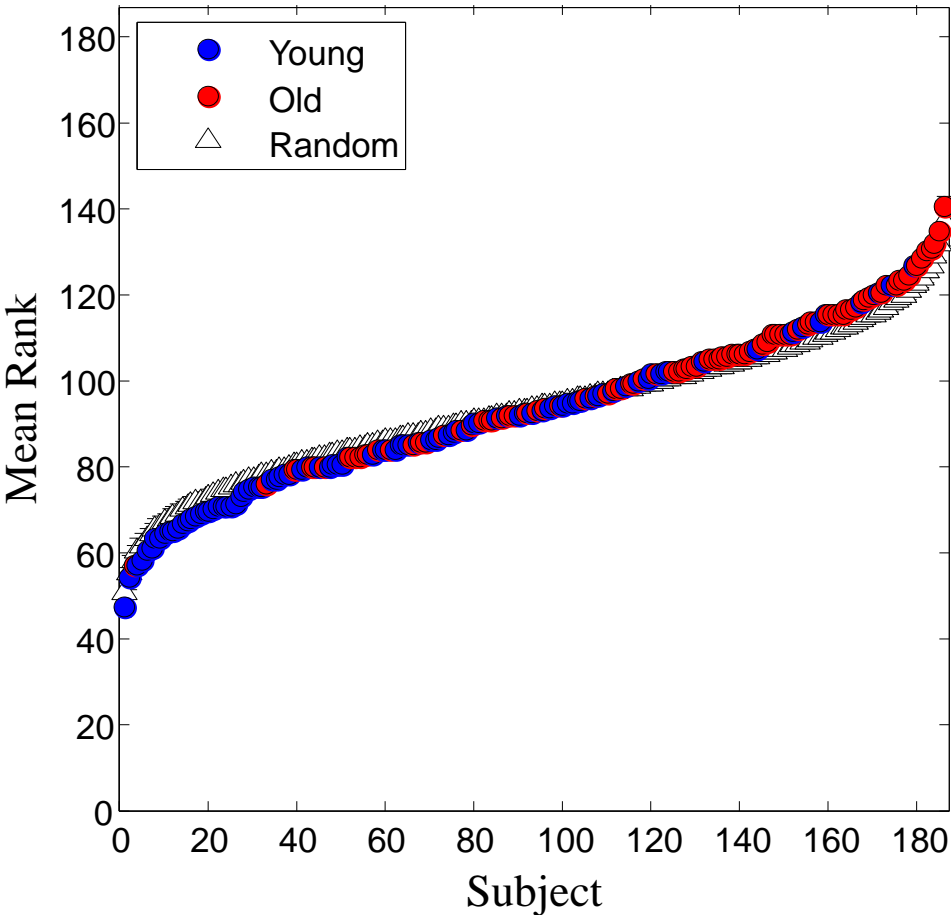


Visual Simon Task

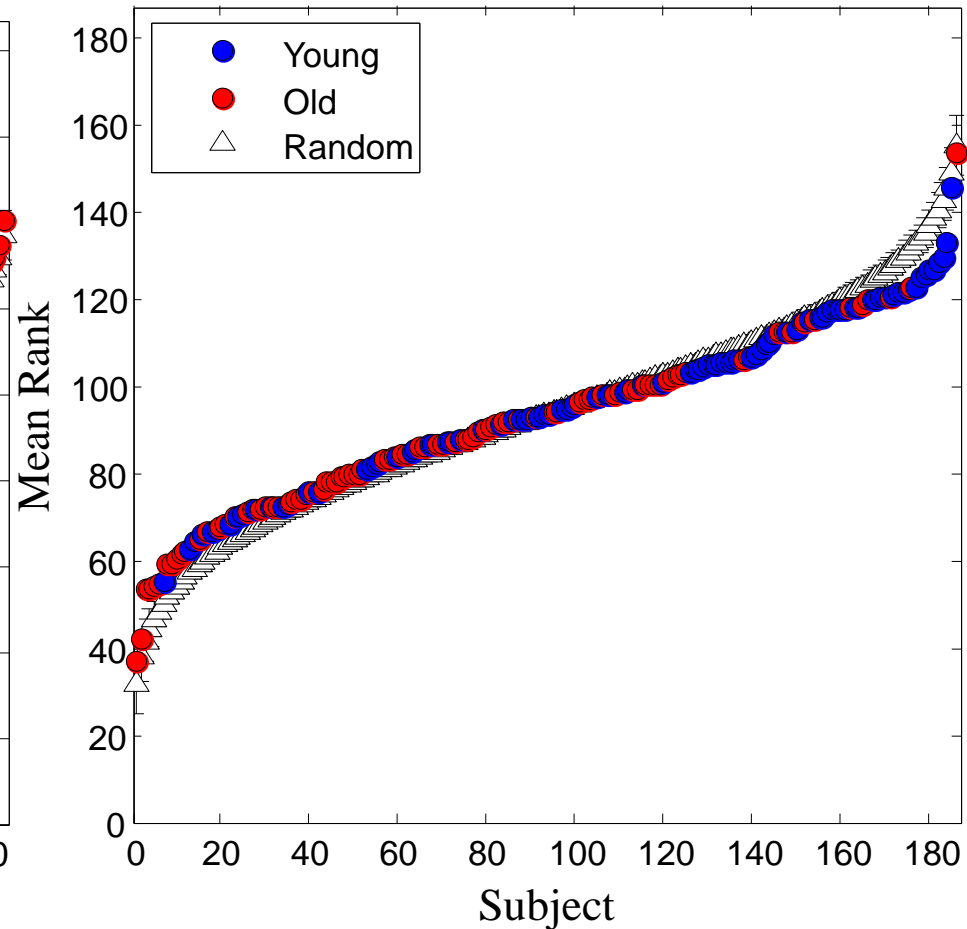


# Ranking analysis

Perceptual Tests  
 $\chi^2(185) = 12.61, p = 1.00$



Cognitive Tests  
 $\chi^2(185) = 34.49, p = 1.00$



Aren't there older people with better perception?

Yes they are better- but not differently than expected by a random selection of "individual" ageing factors for each task



3. Why?

## **There are two aspects**

A methods issue: how we do science- currently

An ontological issue: what do data tell about the  
world

3a. Why? Poor Tests?

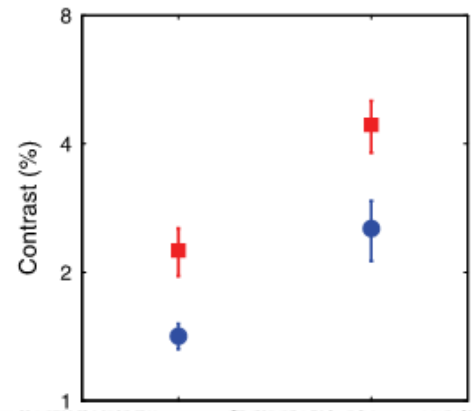
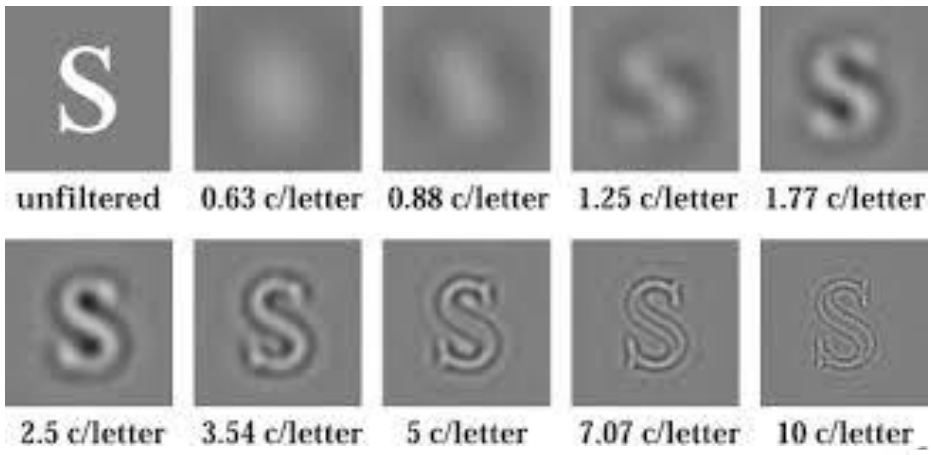
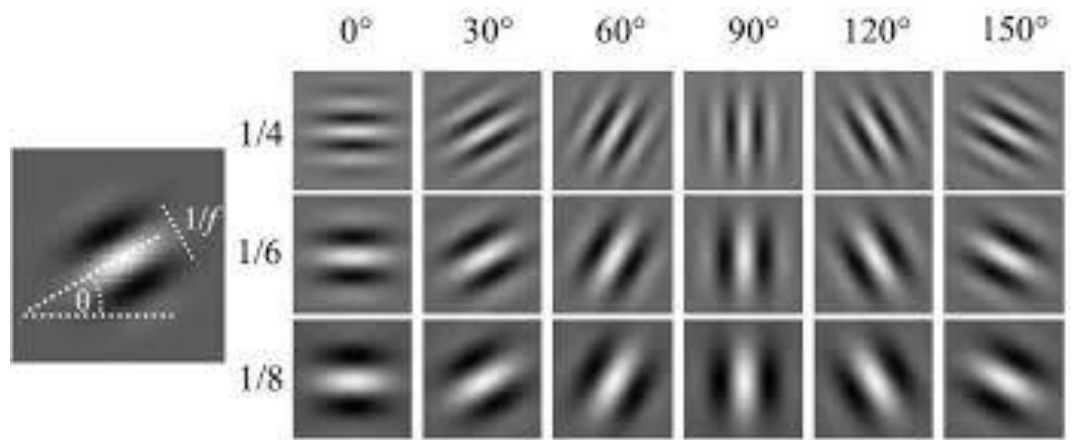
## Why not! Not a matter of....

- Power (with  $n \sim 200$ ; effect size of  $d=0.03$  detectable)
- Test-retest-reliability (which is usually good)
- Reliability paradox (variance is large)
- Linear methods (non-linear methods yield same results)

3b. Why? Idiosyncratic tests?

or

Latent construct does not exist



3c. Why?

Mono-factoriality and significance

# Wait...we got significance



Contrast sensitivity

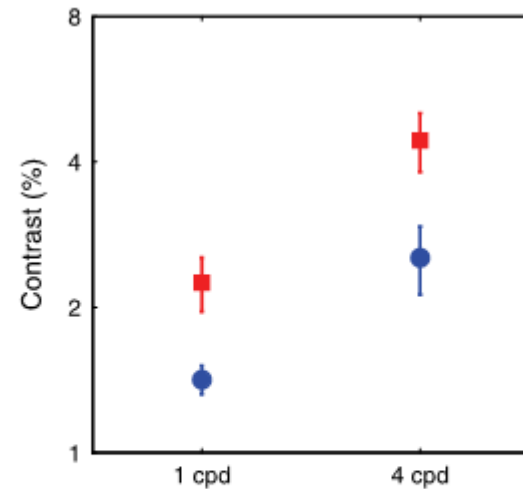
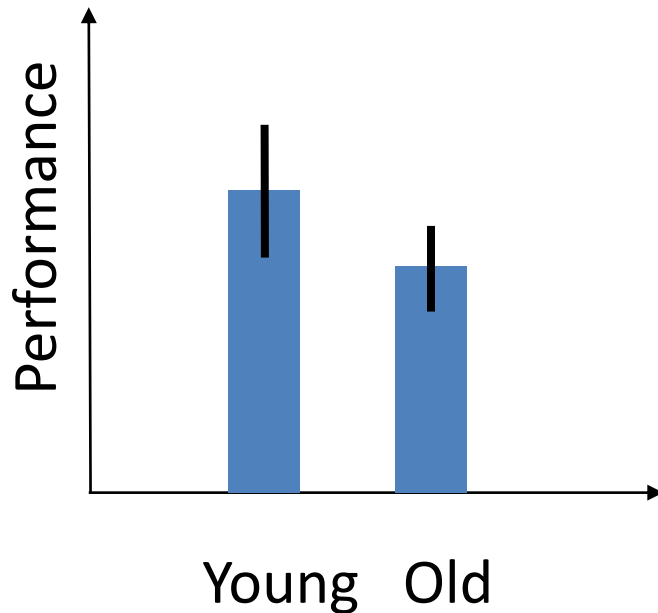


Figure 1. Contrast thresholds for older participants (red squares) and younger participants (blue circles) for 1 and 4 cpd Gabor patterns. Error bars represent  $\pm 1$  SEM across 10 participants.

Significance is not a good justification  
for the selection of a test

# Cohen's d



$d = 0.2$  small effect

$d = 0.5$  medium effect

$d = 0.8$  large effect

$d = 0.8 = 65\%$  discriminability

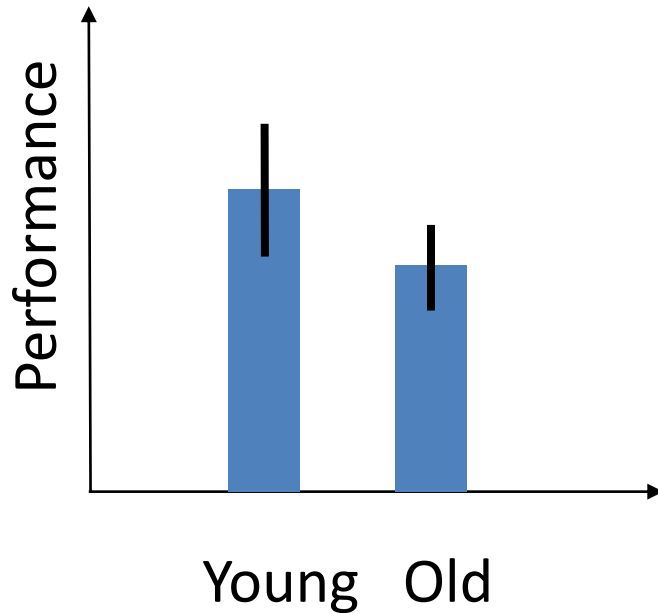
$d \sim 4 = 99\%$  discriminability

$$d = \frac{\bar{x}_Y - \bar{x}_O}{S}$$

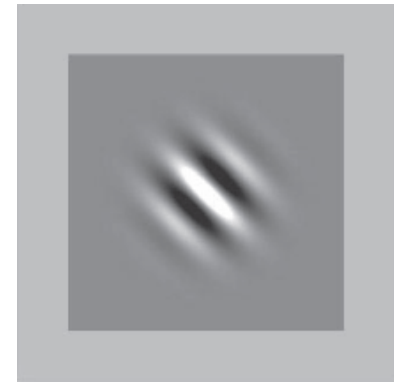
Male vs Female  $d \sim 1.4$

Most  $d$  in science  $< 0.5$

# Cohen's d



$$x_{ij} = v_i + e_{ij}$$



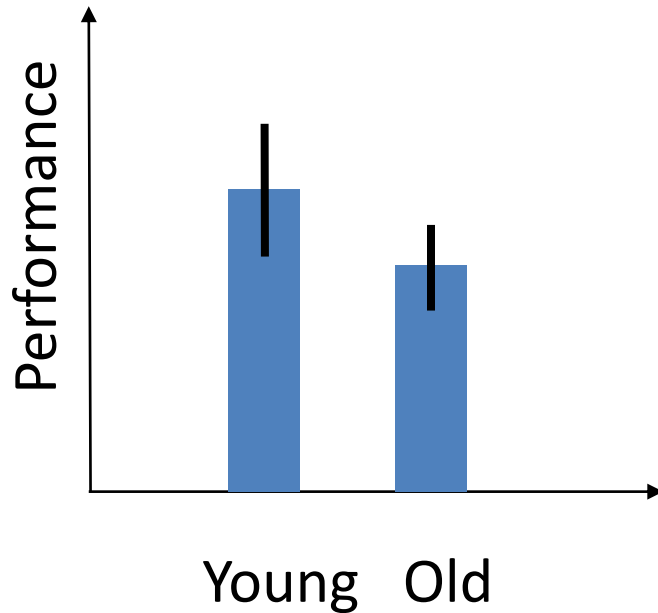
$v_i$ : inter-participant variability

$e_{ij}$ : measurement noise

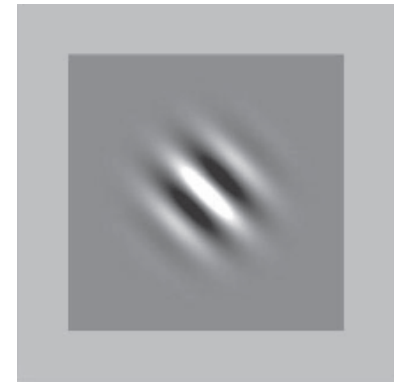
for trial  $j$  for subject  $i$

$$d = \frac{\bar{x}_Y - \bar{x}_O}{s}$$

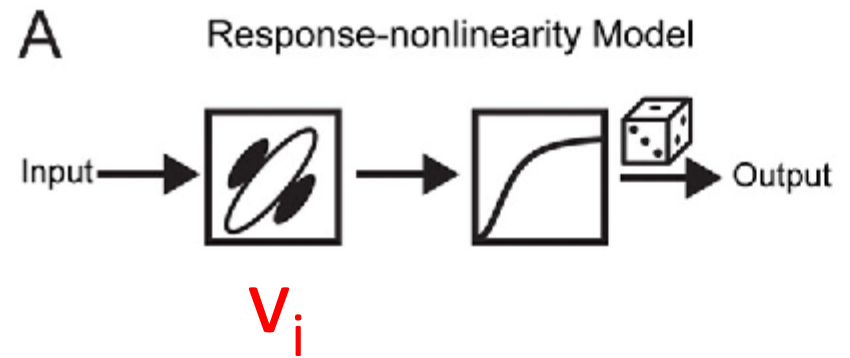
# Cohen's d



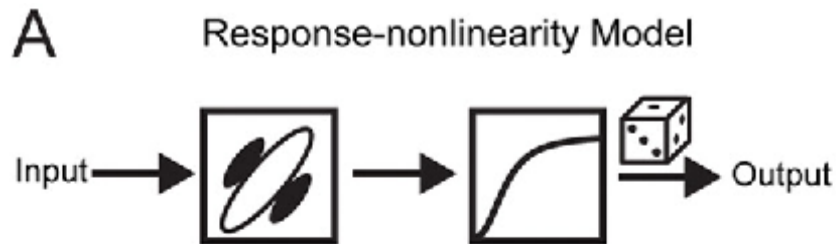
$$x_{ij} = v_i + e_{ij}$$



$$d = \frac{\bar{x}_Y - \bar{x}_O}{s}$$



# Let us look through a microscope:

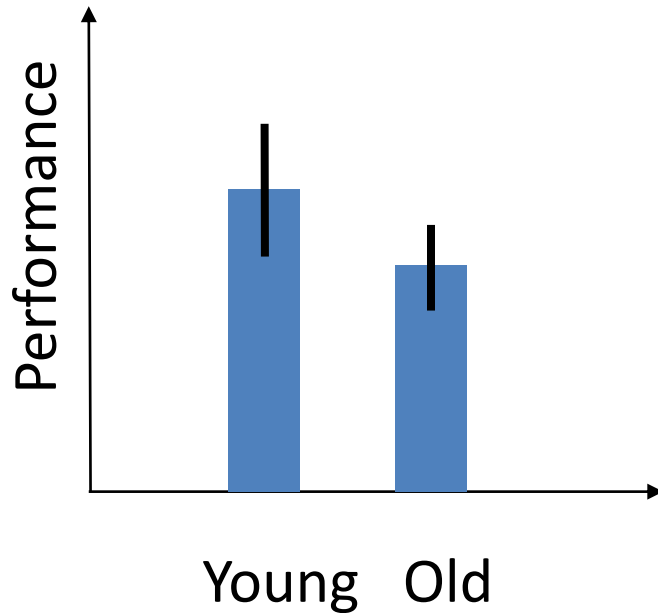


# Lousy microscope paradox



Significant result  
Cohen's  $d = 0.8$   
Same sample size

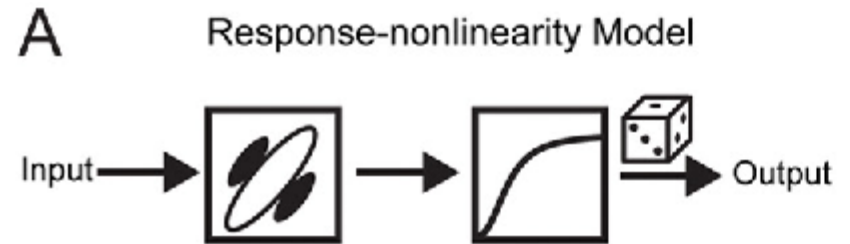
# Cohen's d



$$x_{ij} = v_i + e_{ij}$$

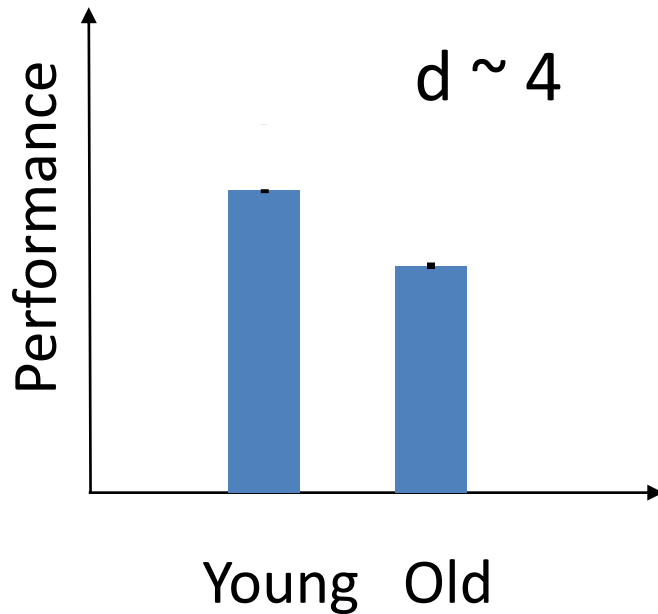


$$d = \frac{\bar{x}_Y - \bar{x}_O}{s}$$



$$d = 0.8 \ll 4$$

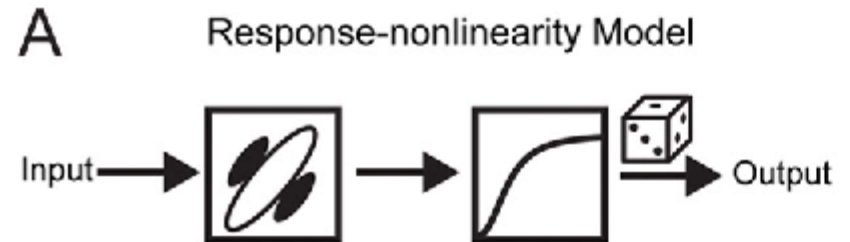
# Cohen's d



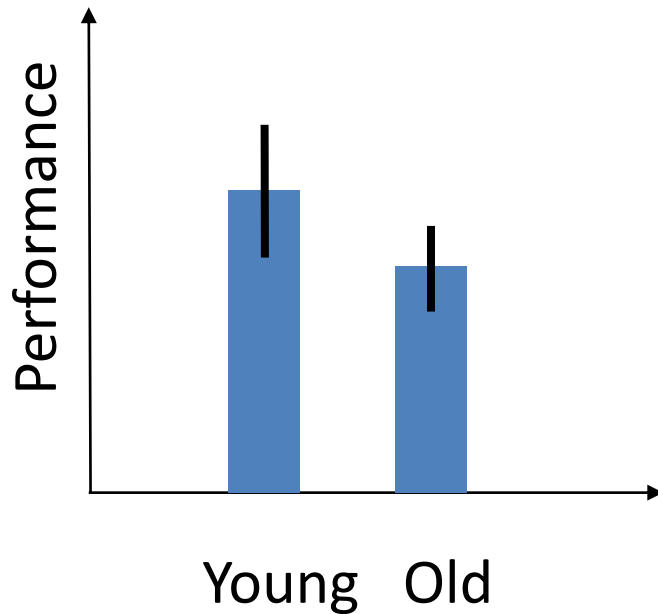
$$X_{ij} = v_i + \cancel{e_{ij}}$$



$$d = \frac{\bar{x}_Y - \bar{x}_O}{s}$$



# Cohen's d



$$X_{ij} = \mu_i + \epsilon_{ij}$$

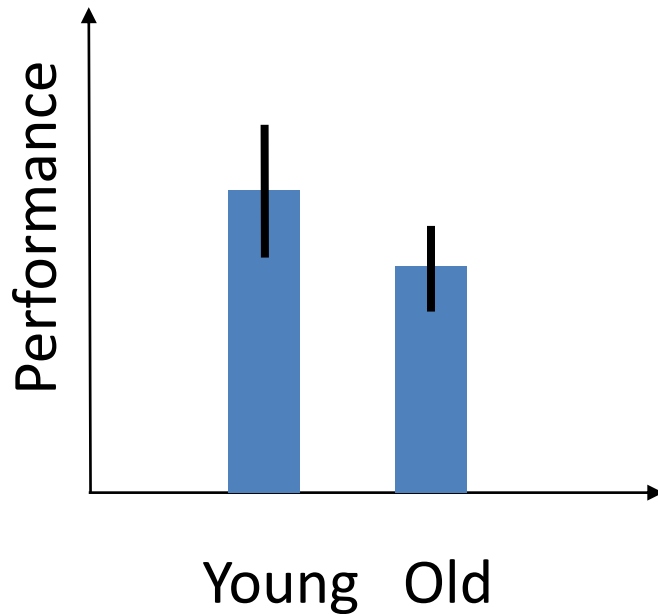


$$d = \frac{\bar{x}_Y - \bar{x}_O}{s}$$



$$d = 0.8$$

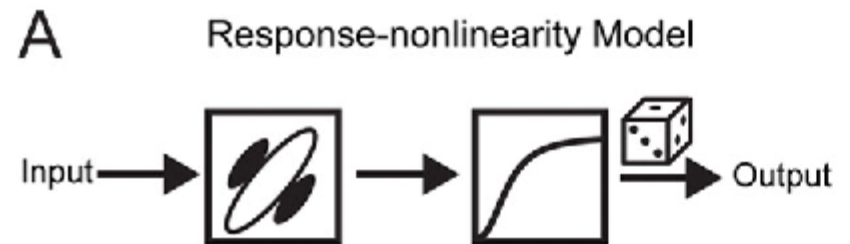
# Cohen's d



$$X_{ij} = V_i$$



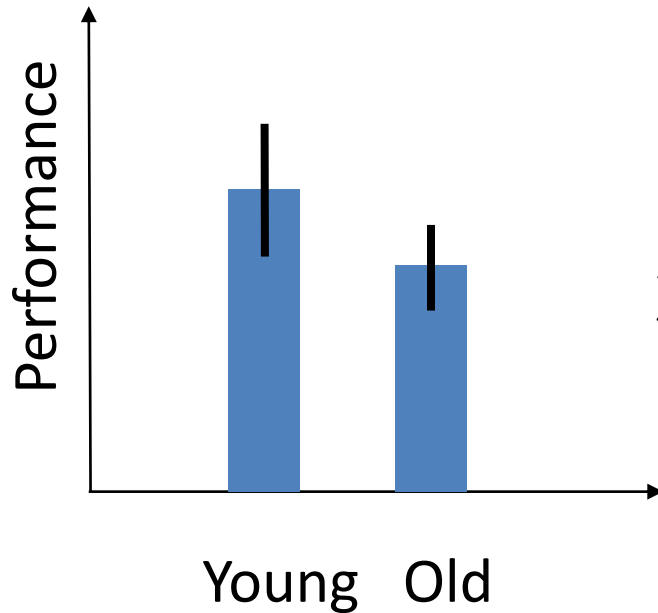
$$d = \frac{\bar{x}_Y - \bar{x}_O}{S}$$



$$d = 0.8 \ll 4$$

65% discriminability

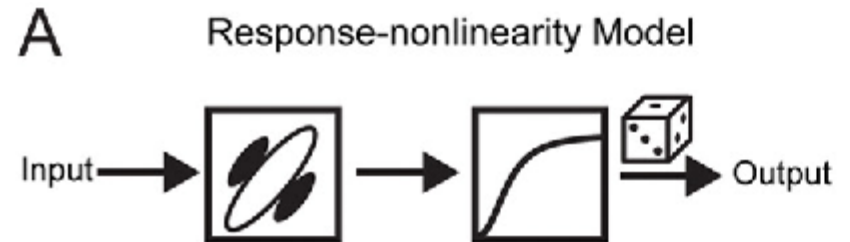
# Cohen's d



$$X_{ij} = V_i + W_i + S_i \dots$$



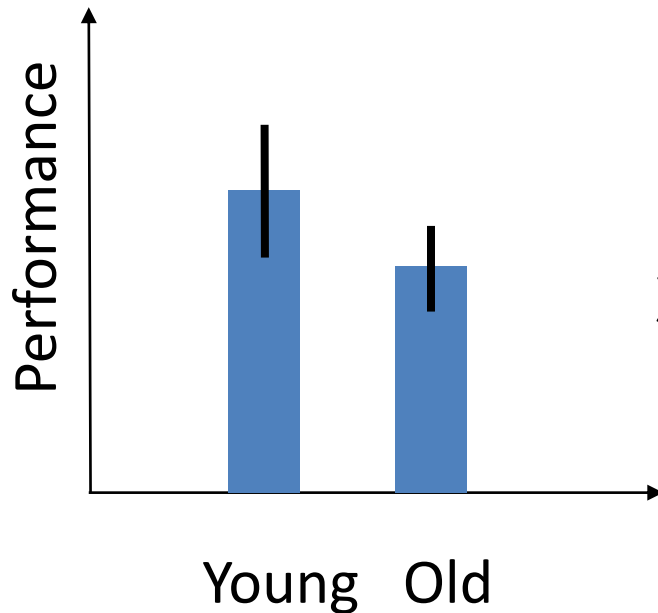
$$d = \frac{\bar{x}_Y - \bar{x}_O}{S}$$



$$d = 0.8$$

65% discriminability

# Cohen's d



Significance

$$X_{ij} = V_i + W_i + S_i \dots$$



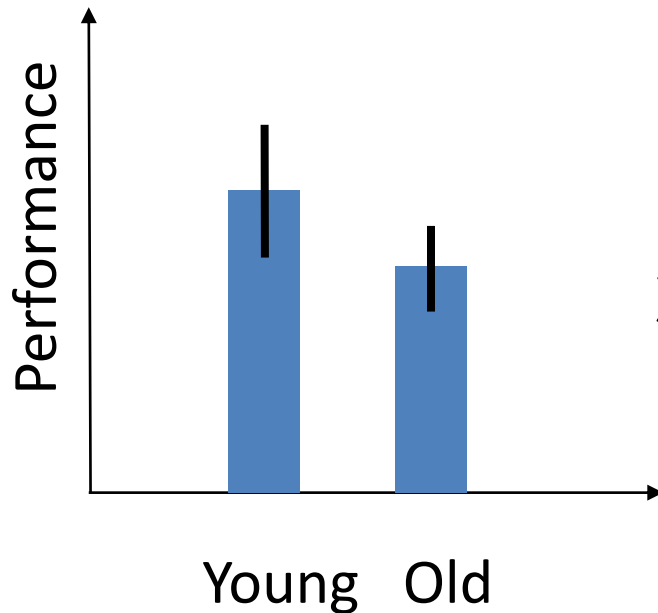
$$d = \frac{\bar{x}_Y - \bar{x}_O}{S}$$



$$d = 0.8$$

65% discriminability

# Cohen's d



Significance  $p \sim d^2 n$

$$X_{ij} = V_i + W_i + S_i \dots$$



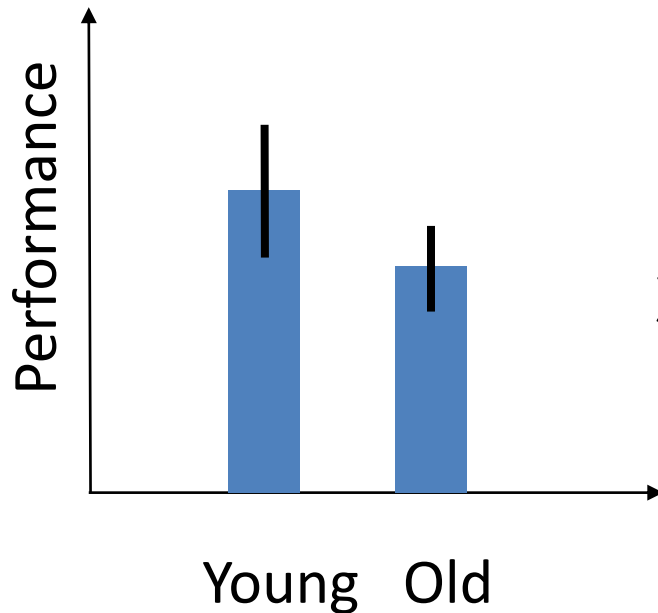
$$d = \frac{\bar{x}_Y - \bar{x}_O}{S}$$



$$d = 0.8$$

65% discriminability

# Cohen's d



Significance

$$X_{ij} = V_i + W_i + S_i \dots$$

Many tests needed



$$d = \frac{\bar{x}_Y - \bar{x}_O}{S}$$



$$d = 0.8$$

65% discriminability

A Cohens d of 0.8 is not large if noise is **low**...

# Strange Paradox I

Technically, we want no noise but we desperately count on it statistically

## Strange Paradox II

$$X_{ij} = v_i + w_i + s_i \dots\dots + e_{ij}$$

We think we are averaging out noise ( $e_{ij}$ ) with large samples but we kill true variability, ie signal ( $v_i, w_i \dots$ )

END Class 8