

# Neural Network-based exploration of construct validity for Russian version of the 10-item Big Five Inventory

Anastasia Sergeeva   Bogdan Kirillov   Alyona  
Dzhumagulova

ITMO University

Skolkovo Institute of Science and Technology

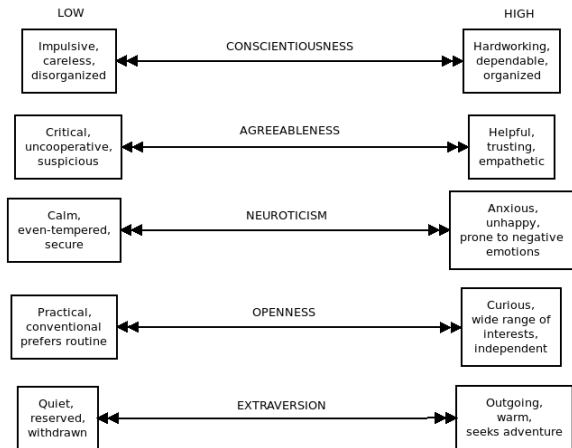


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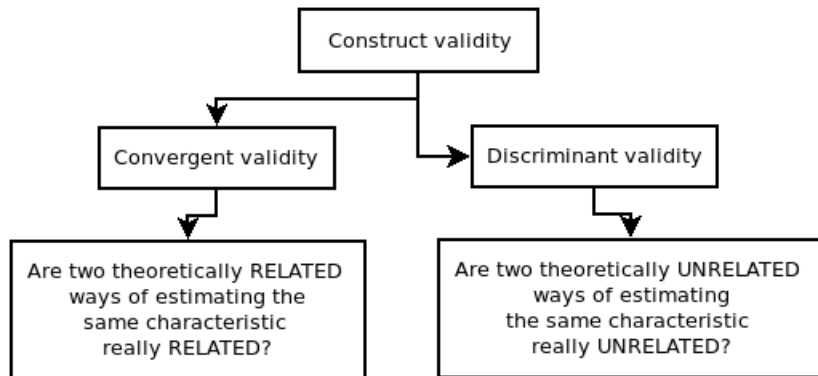
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# Big Five model



# Construct validity

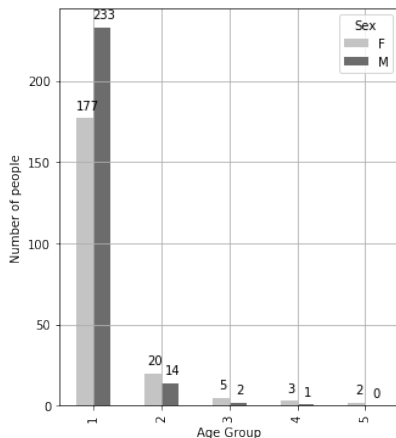


# Main contributions

- ▶ First ever attempt to apply neural networks for construct validity evaluation;
- ▶ Simple qualitative approach to investigate convergent validity of questionnaire via permutation testing;
- ▶ Test of convergent and discriminant validity using interpretation of trained convolutional weights.

The data and source code are freely available at <https://github.com/bakirillov/neurovalidation>.

# Sample



- ▶ 457 participants;
- ▶ 218 of them were taken from (Sergeeva et. al, 2016), available at [https://github.com/bakirillov/tipiru.](https://github.com/bakirillov/tipiru;);
- ▶ Age groups:
  1. 10-19 years;
  2. 20-29 years;
  3. 30-39 years;
  4. 40-49 years;
  5. 50-59 years.

# TIPI computation as neural network

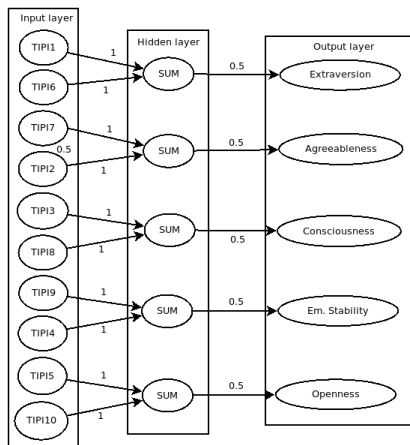
$$E = 0.5(\text{TIPI}_1 + \text{reverse}(\text{TIPI}_6)) \quad (1)$$

$$A = 0.5(\text{TIPI}_7 + \text{reverse}(\text{TIPI}_2)) \quad (2)$$

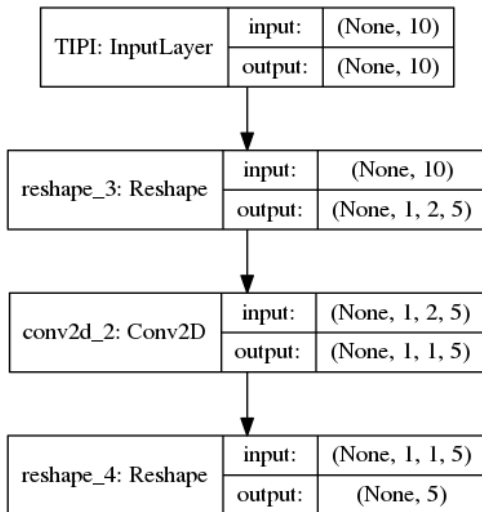
$$C = 0.5(\text{TIPI}_3 + \text{reverse}(\text{TIPI}_8)) \quad (3)$$

$$ES = 0.5(\text{TIPI}_9 + \text{reverse}(\text{TIPI}_4)) \quad (4)$$

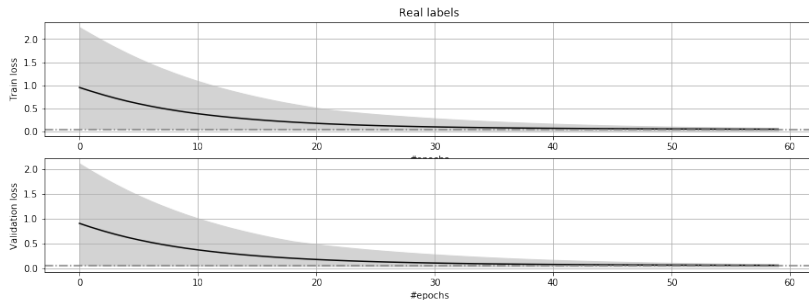
$$O = 0.5(\text{TIPI}_5 + \text{reverse}(\text{TIPI}_{10})) \quad (5)$$



## Actual network that learns TIPI-5PFQ connection



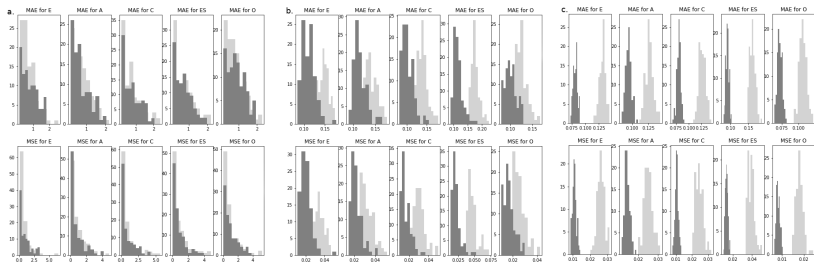
# Training behavior of the network



Key assumption: best possible predictions of 5PFQ that a generalizable (not overfitted) model can make from the TIPI-RU data are actually the TIPI-RU values themselves.

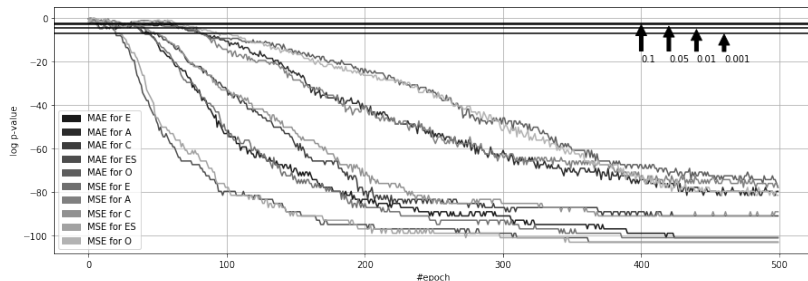


# Permutation testing



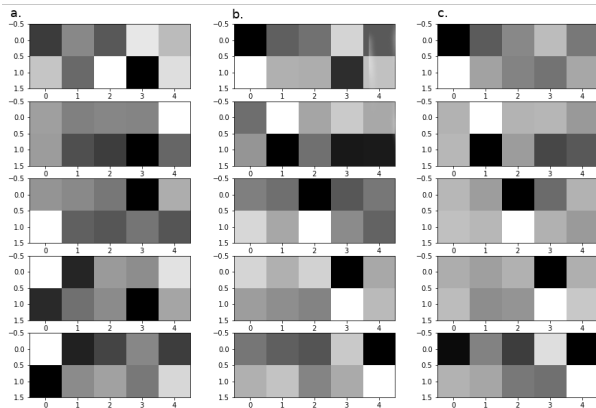
Divergence of two distributions shows presence of learnable connection between TIPI-RU and 5PFQ.

# Kolmogorov-Smirnov test



KS-test converges with the graphical comparison: the p-values go down during the training.

# Interpretation of trained weights



Visualized convolutional weights mimic structure of data. Sign reversal in Agreeableness is captured. Openness is inconsistent.

# Contact information



Anastasia Sergeeva  
an.se.sergeeva@gmail.  
com



Bogdan Kirillov  
Bogdan.Kirillov@  
skoltech.ru



Alyona Dzhumagulova  
aledjuna@gmail.com