

# **A MOBILE APP**

# FOR COMPARATIVE FITNESS EVALUATION

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### ABSTRACT

The aim of this project was to create an Android app that asks the user several health questions and then evaluates his/her age-based relative fitness. Therefore it uses a cluster model that was learned from a health survey data set. The app also visually juxtaposes the cluster properties and the user's answers to help him/her understand the results.

#### THE HEALTH DATA SET

- Questionnaire data set from 'The 10,000 Immunomes Project' <sup>[1]</sup>
  - 1,422 participants
  - Questions: age, gender, race, 54 health

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Nein	

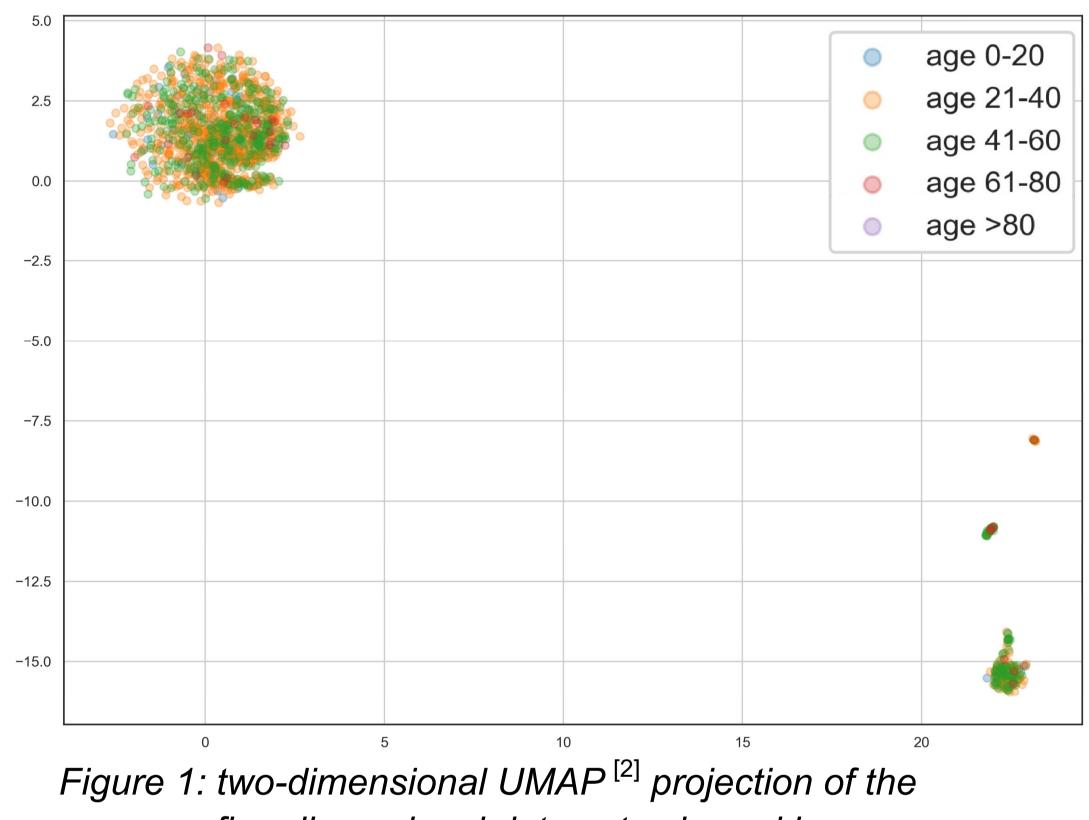
#### THE APPLICATION

- For Android smartphones or tablets
- User can input age and if he/she has some of the 5 medical conditions
- Our of the Uses Weka to assign the user to one of the

attributes  $\rightarrow$  too many for K means clustering

- Extracted 5 binary attributes
  - ◊ Abnormal blood cell count (CSQ1-4)\*
  - Severe pulmonary disease (CSQ11-13)\*
  - Treatment for cancer in the past 5 years
  - Arthritis or rheumatism for more than 3 months
  - Weakness for more than 3 months (CSQ41-43)\*

\* combines multiple attributes using inclusive disjunction ... plus the age attribute (discretised)



Haben Sie ein chronisches Lungenleiden, das Krankenhausbehandlung erfordert, bspw. Lungenfibrose oder schweres Asthma?

🔿 Ja

🔘 Nein

Wurden Sie in den letzten 5 Jahren wegen einer Krebserkrankung behandelt (ausgenommen lokaler Hautkrebs)?

Ja
 Nein
 FITNESS VERGLEICHEN

clusters based on his/her answers

- Evaluates fitness:
  - Here fitness is defined as the difference between the user's age and the average age group of his/her cluster. For example if the user is 30 years old but belongs to a cluster with an average age of 41-60, he/she is in relatively poor health.
  - Depending on the result of this comparison the app displays a message with congratulations, a neutral or a concerned comment.
- Visualises the user's answers next to the cluster properties to make the cluster assignment comprehensible (see below)

## THE CLUSTER-USER DIAGRAM

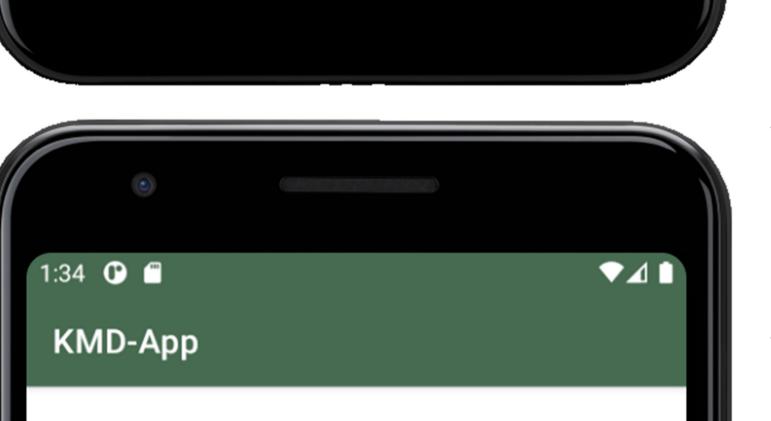
 One example for this diagram is shown in the second screenshot of figure 3. The diagram will

five-dimensional data set coloured by age groups

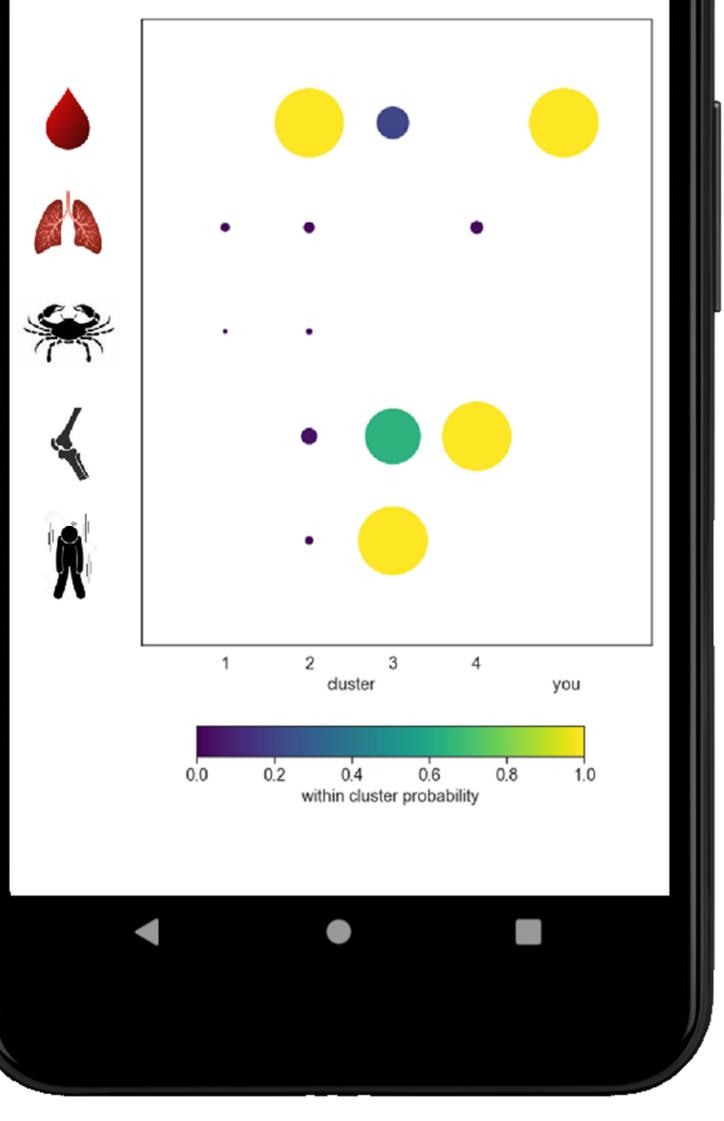
- Structure analysis: Two-dimensional projection of the five-dimensional data set shows 4 clear groups with different age composition
- Only the health-related attributes are used for the following clustering. No clustering by age.

## CLUSTERING AND FINDING THE OPTIMAL K

- Built several cluster models using Weka <sup>[3]</sup>
  - Clustering method: SimpleKMeans
  - Distance metric: Euclidean distance
  - ◊ Number of clusters K: 2, 3, ..., 10, 11
- Elbow point method: compared the within cluster SSE of the different models to find the optimal  $K \rightarrow$  chose K = 4: is elbow point and



Sie gehören zu Cluster Nr. 2 Glückwunsch, Ihre Fitness entspricht der eines/einer 21 - 40 - Jährigen.



adapt to the user's answers.

◊ X axis:

◊ First 4 columns stand for the clusters

Last column stands for the user's answers

- Y axis: The pictograms represent the medical conditions (same order as in the list on the left).
- Circles: Size and colour both express the probability that we observe the respective medical condition on an individual within the respective cluster. If there is no visible circle this represents a 0% probability. The right column with the user's answers will always show a mix of either big yellow (100%) or non-visible (0%) circles as the answers are definite.

### Acknowledgements

To Prof. Dr. Monika Brunner-Weinzierl from the medical faculty of Otto von Guericke University Magdeburg for the cooperation in the context of

# matches structure from figure 1 best

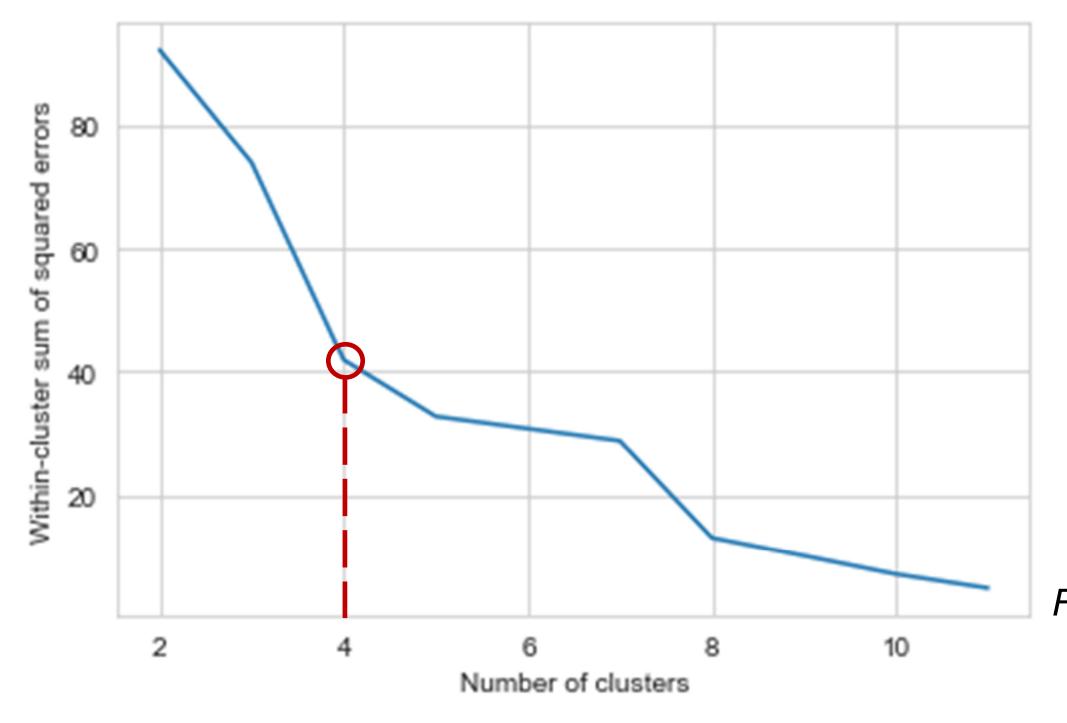


Figure 3: screenshots of the app showing questionnaire (top) and evaluation (bottom)

Figure 2: within cluster SSE in relation to the number of clusters

# the EFRE-funded ImmunLearning project

#### References

[1] Zalocusky KA, Kan MJ, Hu Z, Dunn P, Thomson E, Wiser J, Bhattacharya S, Butte AJ: The 10,000 Immunomes Project: Building a Resource for Human Immunology. Cell reports 25(2), 513-522 (2018)

[2] McInnes L, Healy J: Uniform Manifold Approximation and Projection for Dimension Reduction (UMAP), v0.5. Available: https://umap-learn.readthedocs.io/en/latest/, Accessed on: Apr 28, 2021

 [3] Hall M: Waikato Environment for Knowledge Analysis
 (Weka), v3.8.4. Available: https://www.cs.waikato.ac.nz/~ml/ weka/, Accessed on: Apr 28, 2021