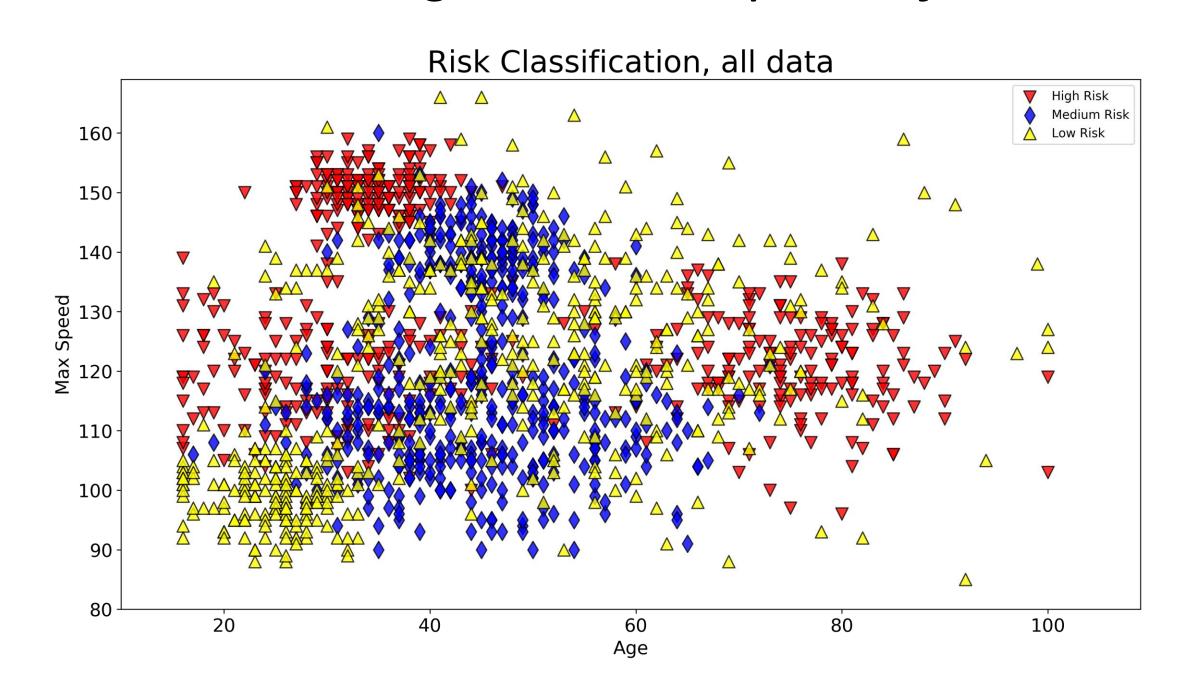
# Scaling Down – Squaring Deep Neural Networks for interpretability and lightweight deployment

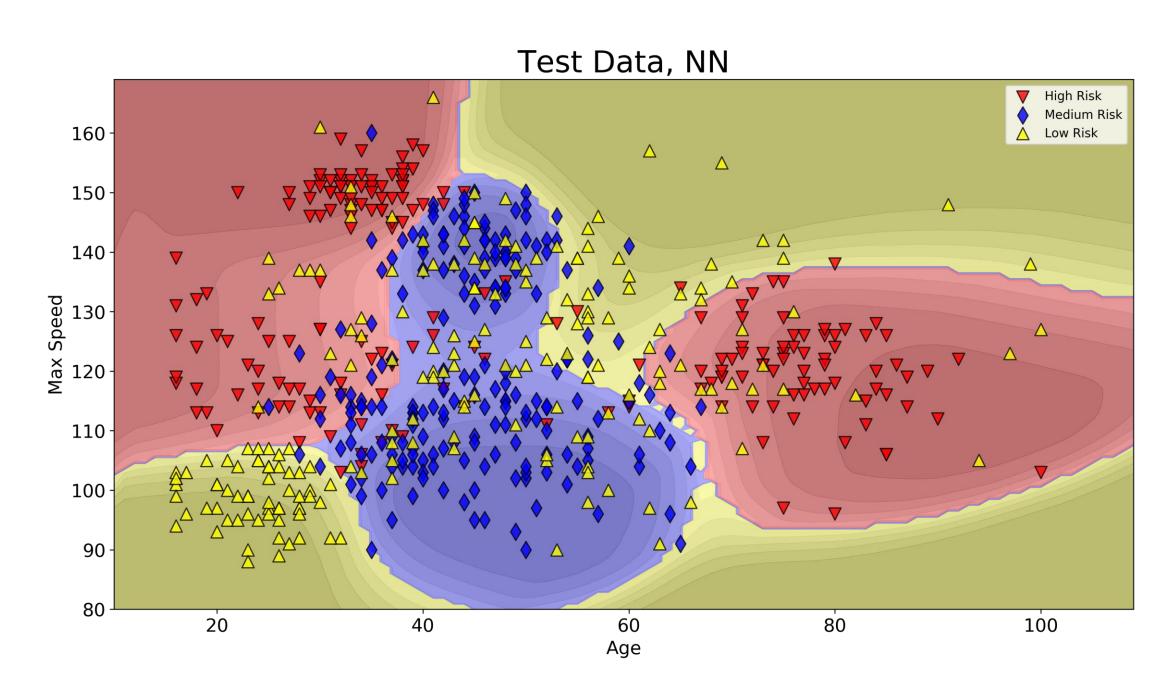
Oliver Zeigermann, <a href="http://scaledml.org/2020/">http://scaledml.org/2020/</a>

#### Use Case - Risk Prediction

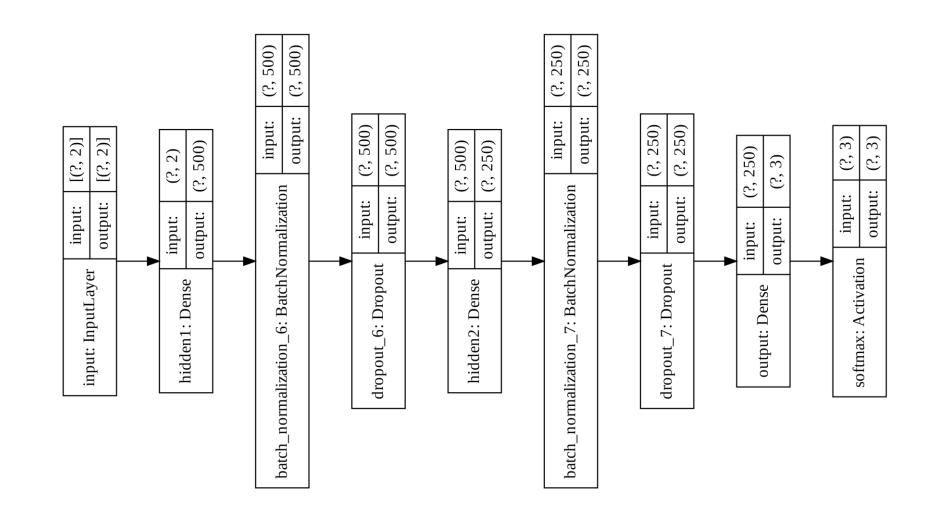
We want to score the risk of potential customers based on their age the max speed of their cars.



### Deep Neural Networks can solve this

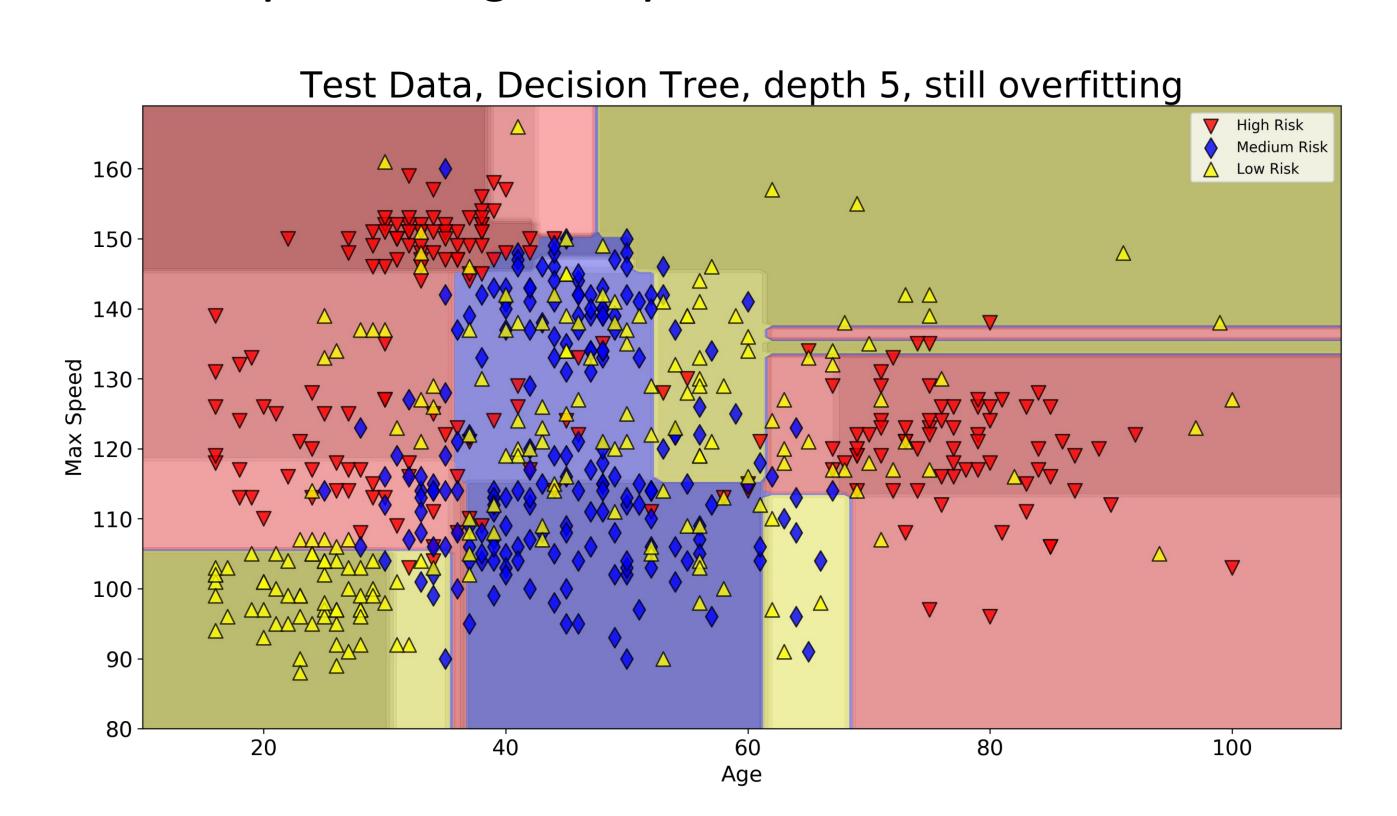


gives high accuracy without any overfitting, but also 130k parameters make this a block box



# Decision Trees offer explainability

- the right to explanation is well established by the GDPR and United States' Credit score
- shallow decision trees allow for at least a basic level understanding,
- but tend to overfit even when regularized
- can be used to generate readable code from
- cheap to bring into production

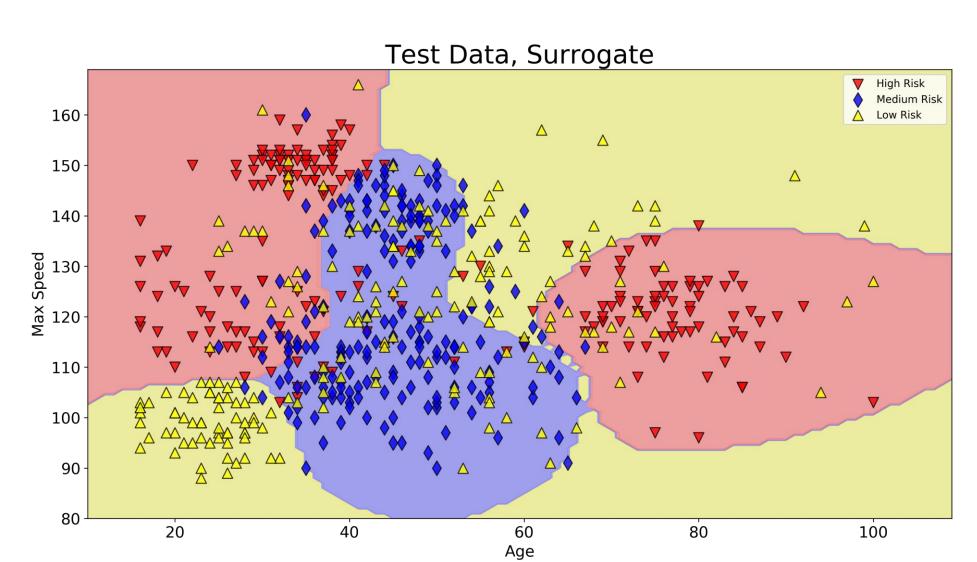


### Best of both worlds – use DNN as a teacher for Decision Trees

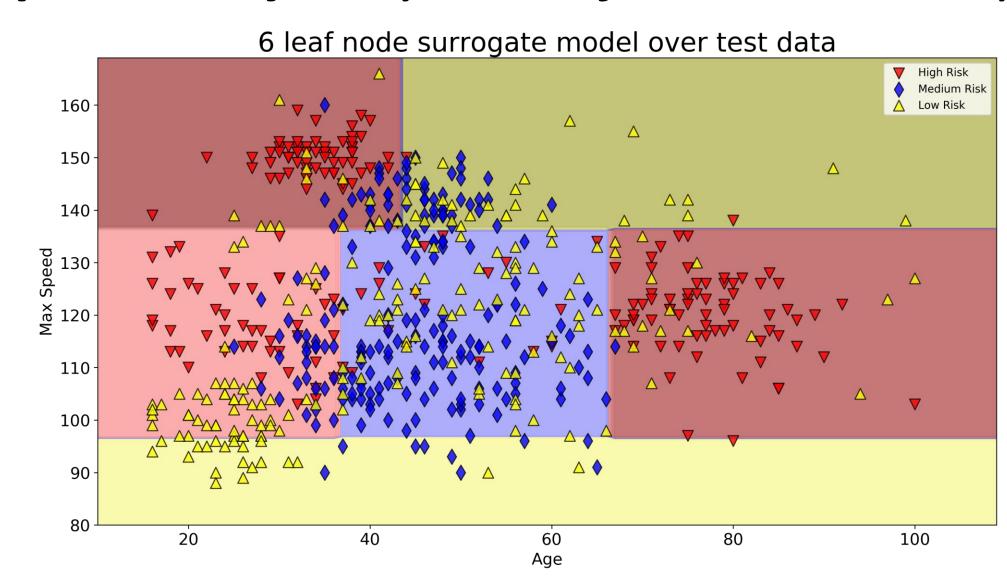
- 1. high capacity deep neural network is trained and regularized like shown in left column
- 2. used to predict a dense set of data over complete range of domain
- 3. this data, not original data, is used to train decision tree
- 4. does not overfit

## Student Decision Tree can be tuned for

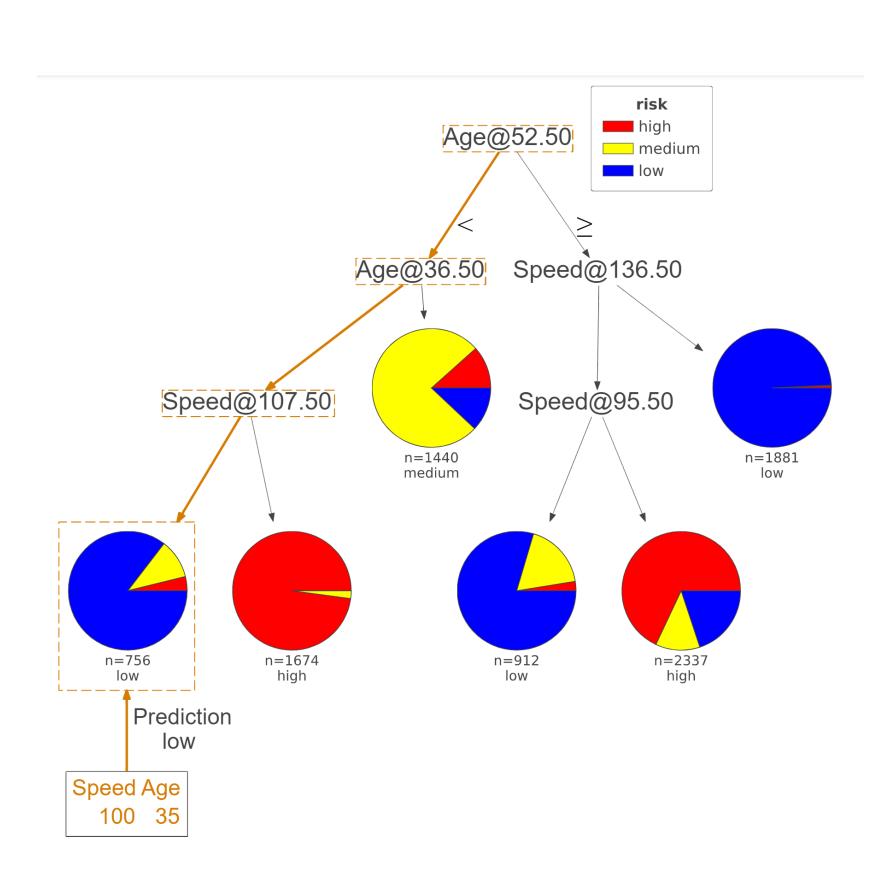
accuracy: same score as teacher, but depth of 12



interpretability: only six leaf nodes, and depth of 3



Sample prediction path



Notebook: <a href="http://bit.ly/scaling-down-2020">http://bit.ly/scaling-down-2020</a>