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Personalized k-fold Cross-validation Analysis  
with Transfer from phasic to tonic Pain  
Recognition with X-ITE Pain Database

**December 20, 2020**

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# Automated Pain Recognition

- Enables reliable pain assessment for patients with cognitive and verbal impairments
- Ensures equal pain management to all patients
- Examples of previous pain recognition databases:
  - BioVid
  - SenseEmotion
  - UNBC-McMaster

# X-ITE database

- Multiple pain models
- Different pain stimuli
  - Intensity
  - Duration

# X-ITE database

- Heat
  - Phasic electro (5 seconds)
  - Tonic heat (60 seconds)
- Electro
  - Phasic heat (5 seconds)
  - Tonic electro (60 seconds)
- 3 intensity levels

# Machine Learning

- Random Forest
- Dense Neural Network

# Random Forest

Subset	Average Accuracy
Phasic electro	91.8%
Tonic electro	85.3%
Phasic heat	82.1%
Tonic heat	79.2%

Figure: Random Forest Scores

- \*271 trees in the forest and 861 maximum number of nodes for each tree. 10-fold Cross-validation was used to make sure each model was not over-fitting the data of each subset.

# Random Forest Cross-validation

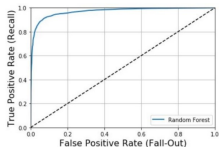


Figure: Phasic Electro Cross-Validation Results

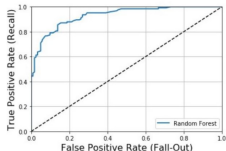


Figure: Tonic Electro Cross-Validation Results

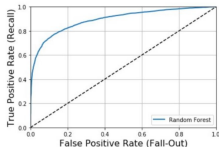


Figure: Phasic Heat Cross-Validation Results

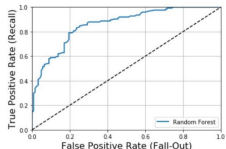


Figure: Tonic Heat Cross-Validation Results



# Personalized Random Forest Cross-validation

Subset	Average Accuracy
Phasic electro	92.0%
Tonic electro	87.3%
Phasic heat	82.0%
Tonic heat	77.1%

Figure: Personalized Random Forest Scores

# Personalized Dense Neural Network

- one flatten input layer
- one hidden dense layer with:
  - 133 neurons
  - Scaled Exponential Linear Unit activation function
  - LeCun normal kernel initializer
- one output layer with sigmoid activation function
- AlphaDropout with a rate of 0.2 between each hidden layer
- first 217 features

# Personalized Dense Neural Network

Subset	Average Accuracy
Phasic electro	90.1%
Tonic electro	81.5%
Phasic heat	82.5
Tonic heat	73.6

Figure: Personalized Dense neural network Scores

# Transfer from Phasic to Tonic

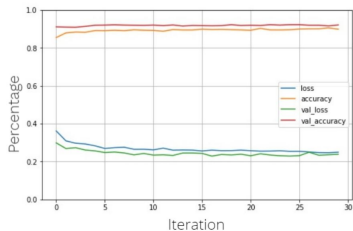


Figure: Transfer Dense neural network Scores

# Transfer from Phasic to Tonic

Model	Average Accuracy	Average Loss
Random Forest	62.3%	N/A
Dense Neural Network	65.2%	2.4328
Personalized Dense Neural Network	63.9%	2.4900

Figure: Phasic electro to tonic electro transfer accuracy

# Discussion and Conclusion

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- Results

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- Results
- Comparison and Added Value



# Discussion and Conclusion

- Results
- Comparison and Added Value
- Limitations

# Discussion and Conclusion

- Results
- Comparison and Added Value
- Limitations
- Future Work

# Questions?

# References I



[1] Gruss, Sascha, et al. "Multi-modal signals for analyzing pain responses to thermal and electrical stimuli." JoVE (Journal of Visualized Experiments) 146 (2019): e59057.



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