



New Mexico Research Spotlight Forum

1/29/2019 Artificial Intelligence & Machine Learning

SAND2019-0851PE

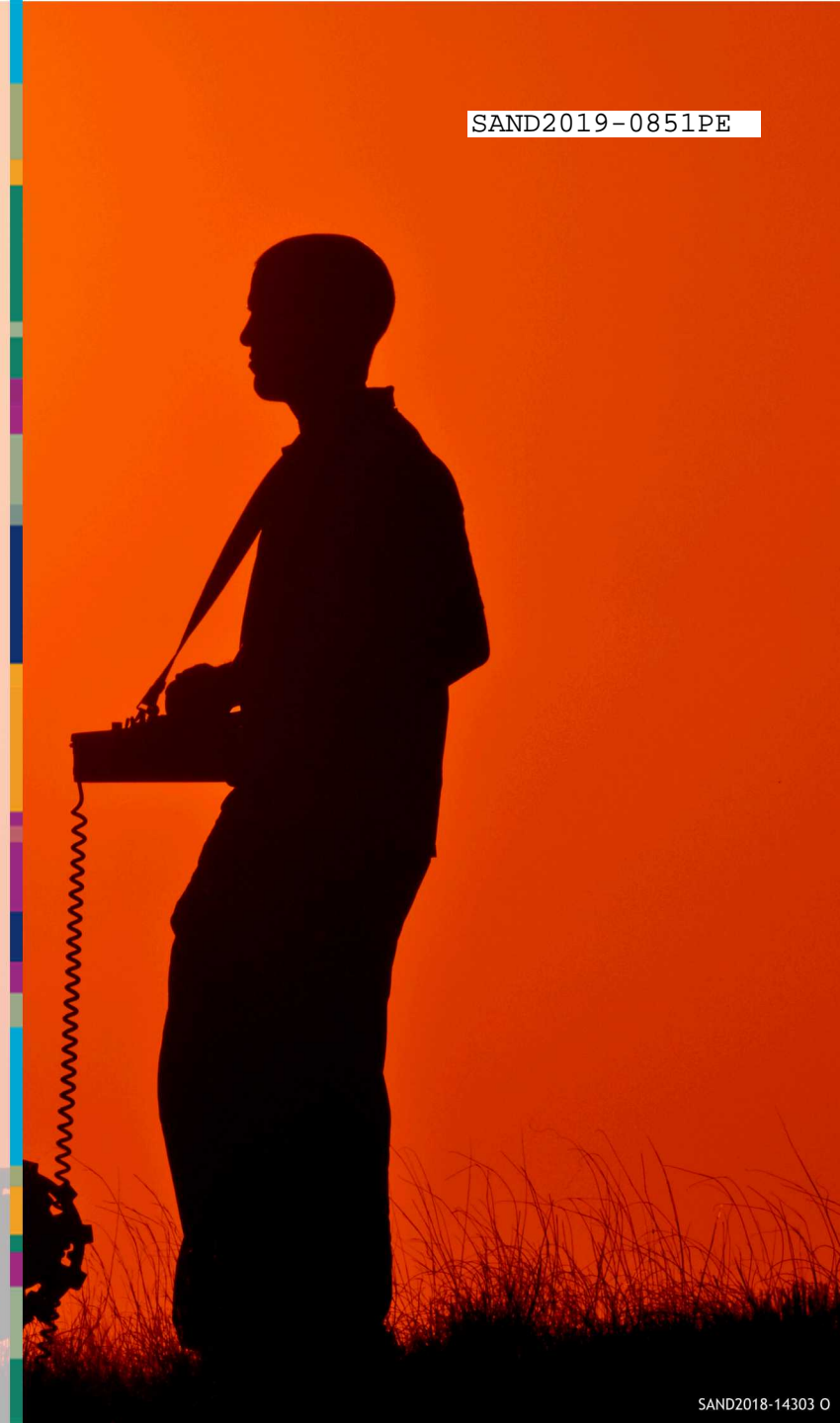
Falls Risk Classification Using Smartphone Based Inertial Sensors and Deep Learning

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Capability Overview



Phillip De Leon

- Professor, Electrical & Computer Eng and Associate Dean of Research
- Time-frequency signal analysis, Speech processing, Machine Learning

Keywords:

Time-frequency signal analysis, empirical mode decomposition(EMD), falls risk prediction, speaker verification (anti-spoofing)

Matthew Martinez

- Statistical Sciences (9136)

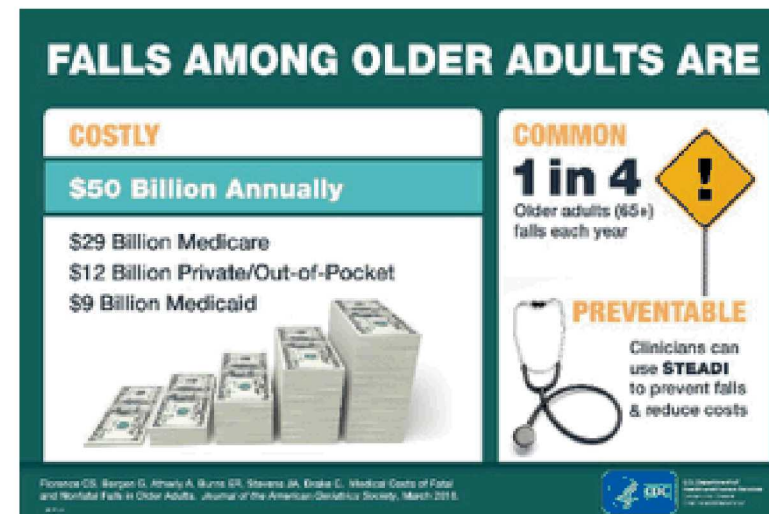
Keywords:

Digital Signal Processing, Deep Learning and Machine Learning for Time Series Analysis, Uncertainty Analysis, Monte Carlo Simulation

FALLS PREVENTION / FALLS RISK RESEARCH BACKGROUND

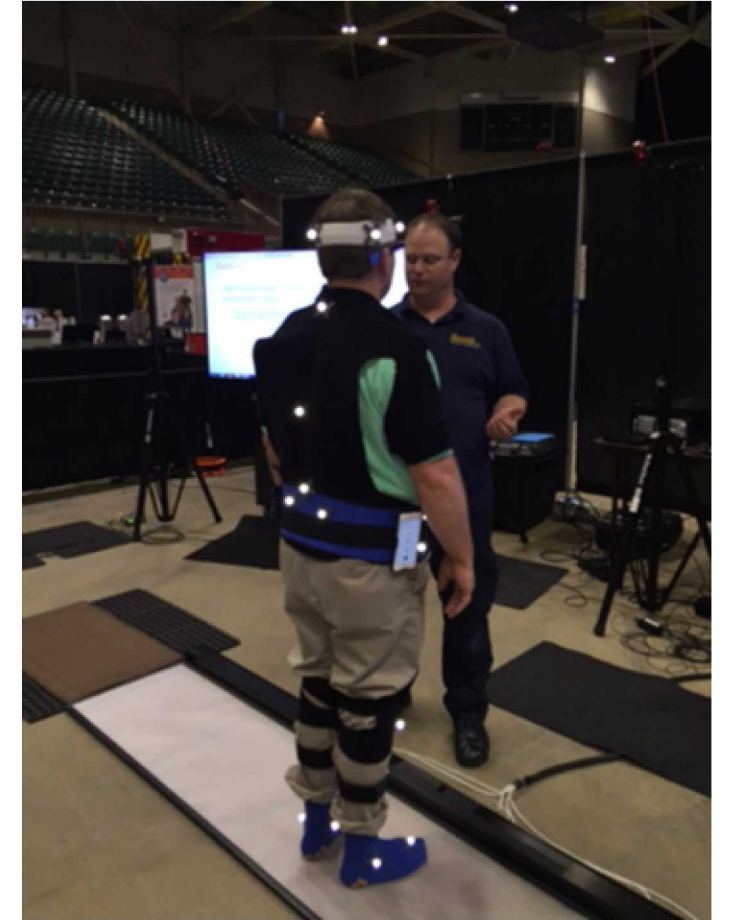
- Each year 2.8 million adults are treated for fall related injuries
 - Broken bones, hip fractures, traumatic brain injury
 - Results in 800,000 hospitalization each year
- Falls Prevention Research
 - Research has focused on assessment, prevention, and rehabilitation
 - Qualitative- and mobility-based assessments
 - Prior research has shown certain gait factors indicate elevated risk of falling
- Sensors for Gait Analysis
 - 3-D motion capture, Pressure sensitive walkways, inertial sensors
 - Inertial Sensors allow for continuous gait monitoring

Our research seeks to use smartphone-based inertial measurements and deep learning to classify older adults as either low or high falls risk



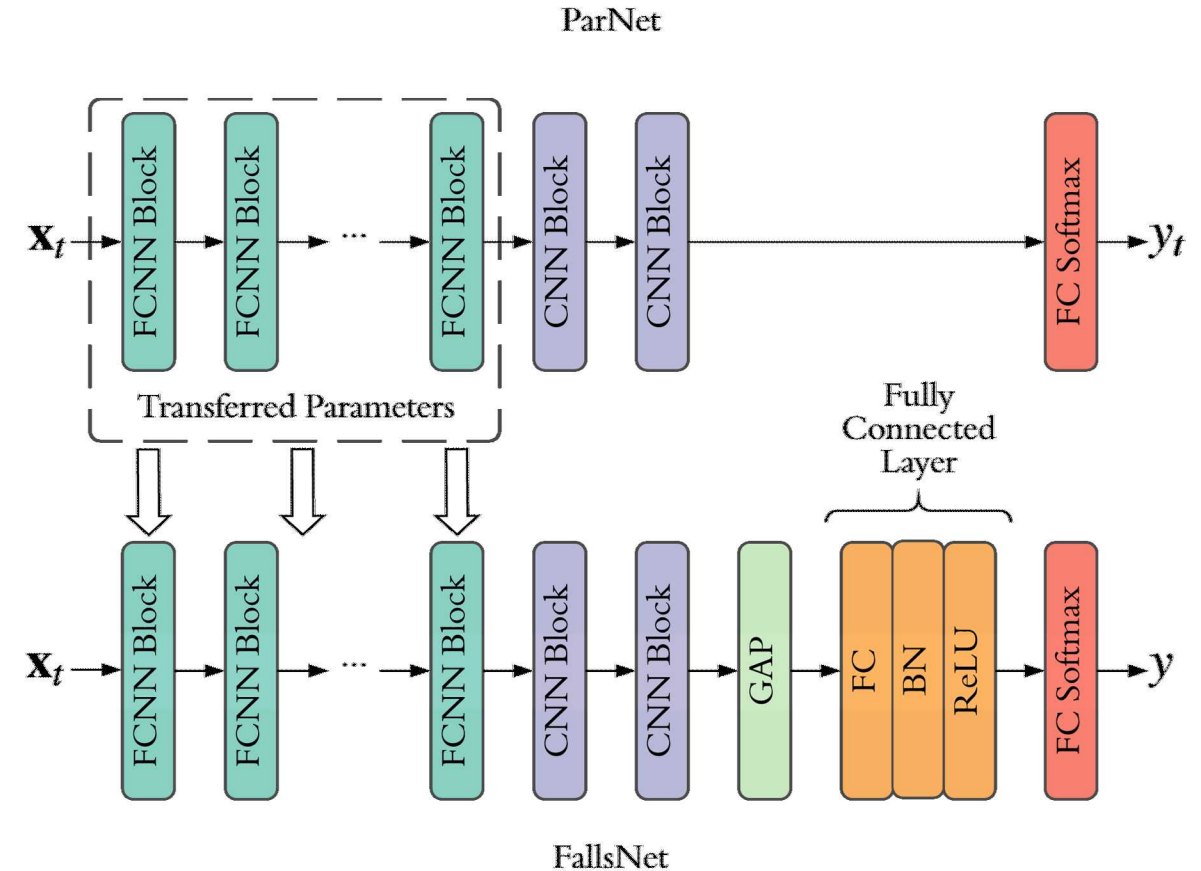
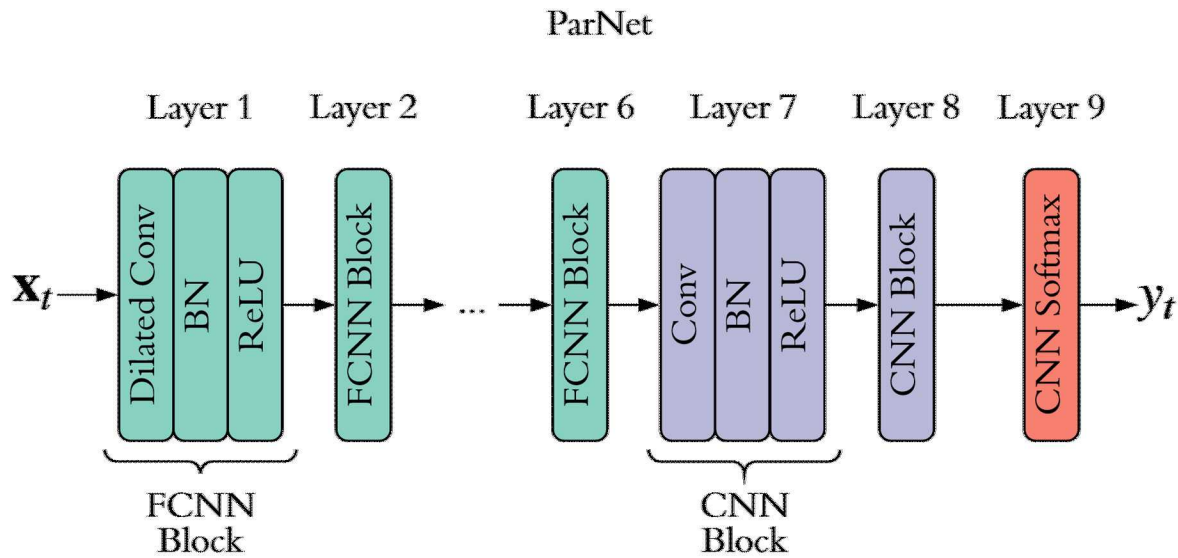
DATA SETS, LABELING, AND TRANSFER LEARNING

- Parallel data sets (IMU+Walkway)
 - Data collected in partnership with the Electronic Caregiver Company
 - Inertial data (3-axis acceleration and 3-axis gyroscope) collected with (2) Apple iPhone 6 smartphones using custom data logger app
 - Biomechanical data collected from pressure sensitive walkway
 - Data collected 256 participants age 65+
- We used biomechanical measurements from walkway data to label IMU data as low risk/high falls risk based
- Dataset is too small for adequate training of DNN
 - Train DNN for Pedestrian Activity Recognition task using HASC-PAC2016
 - Apply transfer learning to ParNet to adapt for falls risk classification task FallsNet



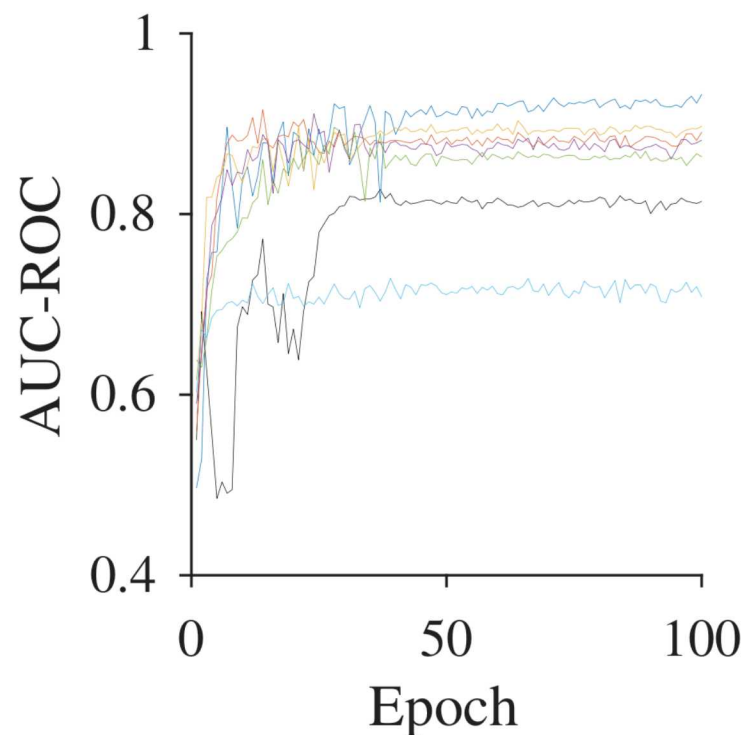
- [1] J. Verghese, R. Holtzer, R. B. Lipton, and C. Wang, “Quantitative gait markers and incident fall risk in older adults,” *J. Gerontol. A Biol. Sci. Med. Sci.*, vol. 64A, no. 8, pp. 896–901, Aug. 2009.
- [2] M. Martinez, P. L. De Leon, and D. Keeley, “Bayesian Classification of Falls Risk”, *Gait & Posture*, vol. 67, pp. 99-103, Jan. 2019.

DEEP NEURAL NETWORK ARCHITECTURE FOR FALLS RISK PREDICTION



[3] M. Martinez and P. L. De Leon, "Falls Risk Classification of Older Adults Using Deep Neural Networks and Transfer Learning," in review *IEEE J. Biomed. Health Inform.*, Jan. 2019.

RESULTS WITH TRANSFER LEARNING



| Model | Layers Transferred, l , to FallsNet | | | | | |
|-------------------------------|---------------------------------------|-------------|-------------|-------------|-------------|-------------|
| | 1 | 2 | 3 | 4 | 5 | 6 |
| ParNet(All, Accel) | 92.1 | 89.7 | 90.4 | 89.7 | 87.0 | 81.3 |
| ParNet(All, Accel + Gyro) | 92.1 | 91.7 | 91.5 | 90.1 | 87.9 | 71.9 |
| ParNet(Waist, Accel) | 91.3 | 91.2 | 92.1 | 88.8 | 89.7 | 79.2 |
| → ParNet(Waist, Accel + Gyro) | 93.3 | 91.5 | 90.3 | 91.1 | 86.9 | 73.5 |

[3] M. Martinez and P. L. De Leon, “Falls Risk Classification of Older Adults Using Deep Neural Networks and Transfer Learning,” in review *IEEE J. Biomed. Health Inform.*, Jan. 2019.

CONCLUSIONS

- Proposed a method for classifying older adults at either low or high falls risk using inertial gait data acquired from a smartphone
- Show how to pre-train a deep neural network to learn feature representation related to human motion using publicly available pedestrian activity data
- Showed how to use a pre-trained deep neural network as feature extractor for falls risk classification
- End-to-end training of a deep neural network for falls risk classification from inertial measurements of gait

| Model | ACC (%) | SENS (%) | SPEC (%) |
|-------------------------|-------------|-------------|-------------|
| FallsNet (Accel + Gyro) | 86.4 | 85.1 | 87.1 |
| FallsNet (Accel) | 82.6 | 83.0 | 82.4 |
| Logistic Regression | 58.1 | 56.6 | 59.0 |
| Random Forests | 63.8 | 43.9 | 74.8 |
| SVM | 59.6 | 53.9 | 62.8 |

[3] M. Martinez and P. L. De Leon, “Falls Risk Classification of Older Adults Using Deep Neural Networks and Transfer Learning,” in review *IEEE J. Biomed. Health Inform.*, Jan. 2019.



FUNDING SOURCES

Sandia National Laboratories, "Summer Faculty Research in Signal Processing and Machine Learning," 2016-2019

Sandia National Laboratories, "Instantaneous Spectral Analysis for Identification and Classification of Bursty Data in Congested Frequency Bands," 2017-2019

Electronic CareGiver, "NMSU 3D Motion Capture System for Biomechanical Data Collection," 2018

Sandia University Part Time



RESEARCH NEEDS

Vector time series classification (in this work we use 6 channels of data)

ML applications seeking high-resolution signal features (time and frequency)

New approaches for unsupervised and semi-supervised ML

Applications needing signal/source separation

