

Clinical evaluation of machine learning approaches for the classification of gait using static & dynamic algorithms in comparison to human performance

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Musculoskeletal (MSK) disorders such as osteoarthritis account for a loss of 36 million workdays & 1% reduction in Gross national Product (GNP) in the UK. Stroke, the biggest cause of mobility impairment costs £9 billion to the UK, annually ^[1]

The style of one's walking (gait) carries information about identity, gender, emotional affect and health. Gait is a highly reliable biomarker for diagnosis & rehabilitation progress monitoring of MSK and neurological conditions ^[2]

Learning Models to Evaluate

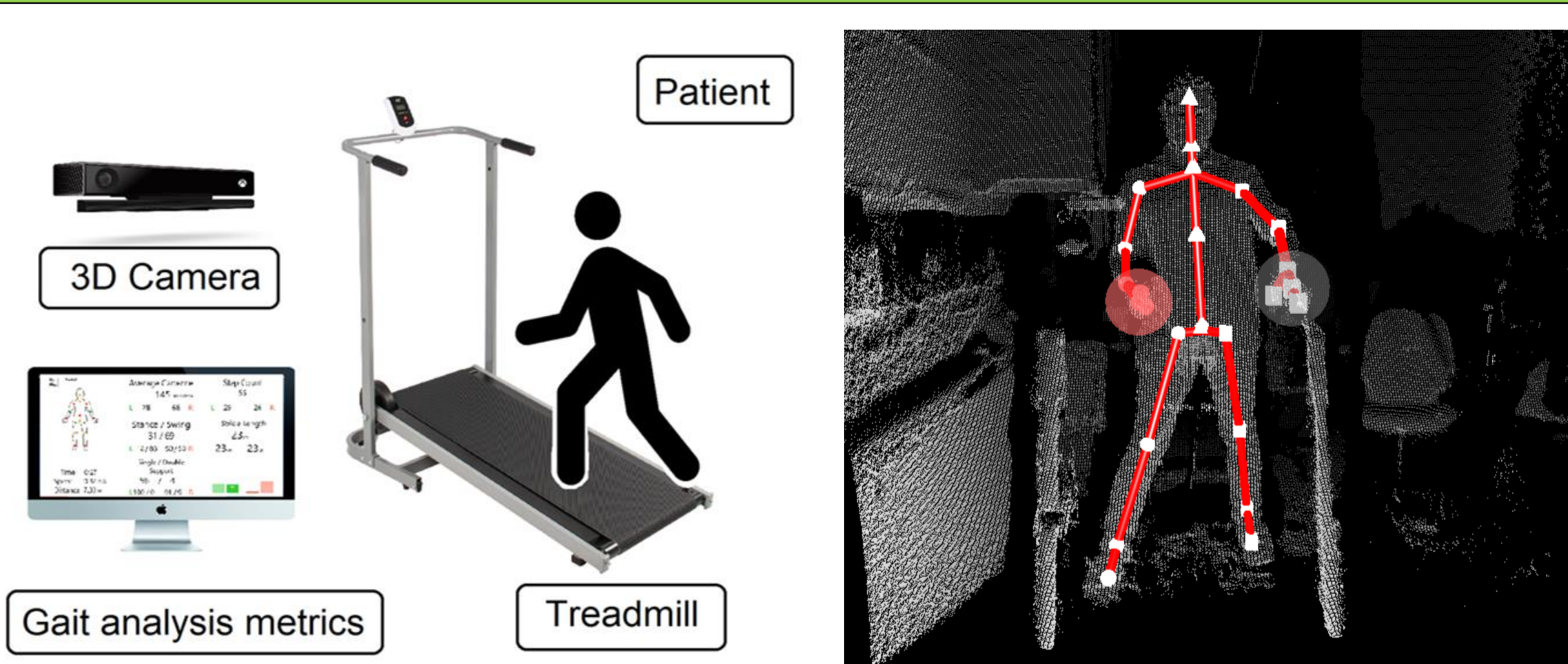
Motion Perception based Dynamic Models

- Human Perception
- Long Short Term Memory (LSTM) ^[3]

Gait Metrics based Static Models

- Support Vector Machine (SVM) ^[4]
- Decision Trees (DT) ^[5]
- Feed Forward Neural Network (FFNN)

Setup



Evaluated On

41 Healthy Subjects (26 M, 15 F); 18-50 years old
12 mins of walking per subject

Models evaluated for gender discrimination from gait. Gender as a proxy for any underlying condition that manifests itself in gait.

Input data types:

- Humans - Point Light Animation (PLA) ^[6]
- LSTM - Raw Motion & derivatives
- Static Models - Clinical Gait Metrics (CGM)
- Static Models - Path Signatures ^[7]

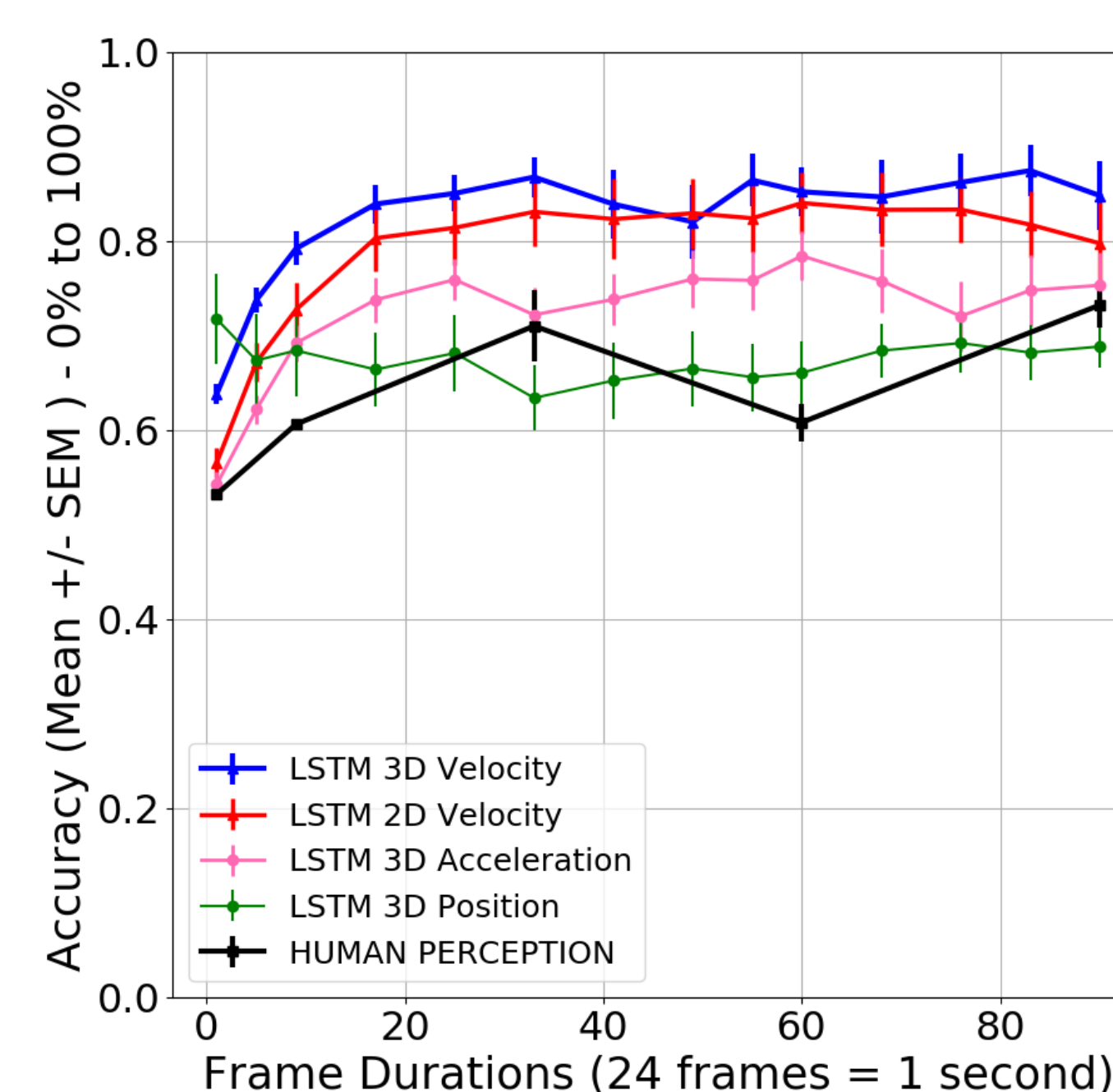
Results

Learning Model	Data Type	Accuracy (%)
LSTM	Velocity	88 +/- 0.99
LSTM	Position	84 +/- 0.98
LSTM	Acceleration	77 +/- 0.67
HUMANS	PLA	74 +/- 0.35
DT	Path Sig.	73 +/- 0.97
FFNN	CGM	61 +/- 1.48
SVM-Poly.	CGM	55 +/- 1.23
SVM-Linear	CGM	54 +/- 0.86
SVM-RBF	CGM	51 +/- 0.43
DT	CGM	45 +/- 0.82

Conclusion

- LSTM based neural networks are the best dynamic models for gender classification from gait with limited data availability
- Path Signatures offer interpretability along with decent classification performance
- LSTM based networks have a higher learning efficiency than humans at an objective humans might be expected to perform well on from an evolutionary perspective ^[8]

Learning Efficiency



[1] Saka, Ömer, Alistair McGuire, and Charles Wolfe. "Cost of stroke in the United Kingdom." *Age and ageing* 38.1 (2009): 27-32.

[2] Jahn, Klaus, Andreas Zwergal, and Roman Schniepp. "Gait disturbances in old age: classification, diagnosis, and treatment from a neurological perspective." *Deutsches Ärzteblatt International* 107.17 (2010): 306.

[3] Sak, Haşim, Andrew Senior, and Françoise Beaufays. "Long short-term memory recurrent neural network architectures for large scale acoustic modeling." *Fifteenth annual conference of the international speech communication association*. 2014.

[4] Scholkopf, Bernhard, and Alexander J. Smola. *Learning with kernels: support vector machines, regularization, optimization, and beyond*. MIT press, 2001.

[5] Quinlan, J. Ross. "Bagging, boosting, and C4. 5." *AAAI/IAAI, Vol. 1*. 1996.

[6] Hill, Harold, Yuri Jinno, and Alan Johnston. "Comparing solid-body with point-light animations." *Perception* 32.5 (2003): 561-566.

[7] Yang, Weixin, et al. "Leveraging the path signature for skeleton-based human action recognition." *arXiv preprint arXiv:1707.03993* (2017).

[8] Pollick, Frank E., et al. "Gender recognition from point-light walkers." *Journal of Experimental Psychology: Human Perception and Performance* 31.6 (2005): 1247.