

Appendix A

List of Publications

- Published/ Accepted Papers

1. Shivang Agarwal, C. Ravindranath Chowdary

A-Stacking and A-Bagging: Adaptive versions of ensemble learning algorithms for spoof fingerprint detection, Expert Systems with Applications, Volume 146, 2020, 113160.

2. Shivang Agarwal, Ajita Rattani, C. Ravindranath Chowdary

A comparative study on handcrafted features v/s deep features for open-set fingerprint liveness detection, Pattern Recognition Letters, 2021 (Accepted).

- Papers Under Review/Revision

1. Shivang Agarwal, C. Ravindranath Chowdary

Combating Hate Speech using an Adaptive Ensemble Learning Model with a case study on COVID-19, Expert Systems with Applications. (Minor Revision Submitted)

2. Shivang Agarwal, Ajita Rattani, C. Ravindranath Chowdary

AILearn: An Adaptive Incremental Learning Model for Spoof Fingerprint Detection, Machine Learning with Applications. (Under Review)

- To Be Submitted

1. Shivang Agarwal, C. Ravindranath Chowdary

Multimodal Hate Speech Detection using Adaptive Ensemble Models.

2. Shivang Agarwal, Ankur Sonawane, C. Ravindranath Chowdary
Accelerating Automatic Hate Speech Detection using Parallelized Ensemble Learning Models.

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- Expert Syst. Appl.*, vol. 146, p. 113160, 2020. [Online]. Available: <https://doi.org/10.1016/j.eswa.2019.113160>
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