

Paul J. Wright, PhD

e: paul@pauljwright.co.uk w: pauljwright.co.uk in: linkedin.com/in/pauljwright91 t: +1 (xxx)-xxx-xxxx

EXPERIENCE	Postdoctoral Scholar (Hansen Experimental Physics Lab), Stanford University Department of Physics, Stanford University, Stanford, CA, USA 2019 – present <ul style="list-style-type: none"> Supporting operations of a flagship NASA mission: The Solar Dynamics Observatory (~\$1B).f Modifying state-of-the-art convolutional neural networks (CNNs) for scientific applications. Collaborating with 9 researchers (5 time zones) to complete an applied deep learning project. Courses Audited: CS229 (Machine Learning [ML]; Autumn 2019)	
	Team Lead & Core Domain Mentor , NASA Frontier Development Lab (FDL) SETI Institute/NASA Ames Research Center, Mountain View, CA, USA NASA FDL is an 8-week applied artificial intelligence (AI) research accelerator that applies ML techniques to challenges in space science and exploration. 2019 – present <ul style="list-style-type: none"> Designed and developed a project to up-scale and convert data between space-based instruments using state-of-the-art deep learning architectures for super-resolution. Facilitated a 3-day Design Sprint at Google Cloud HQ to define the project deliverables. Recruited, led, and managed a multi-national team of 12 (four PhD/Postdoctoral-level researchers and eight mentors, including two super-resolution experts from Element AI). Communicated and managed expectations of stakeholders (Google Cloud, Intel AI, NASA). Presented an <i>invited</i> talk at the American Geophysical Union Fall Meeting (~ 30, 000 attendees), and guided two NeurIPS/(NIPS) (peer-reviewed) workshop papers (in <i>Machine Learning and the Physical Sciences</i>, and <i>Bayesian Deep Learning</i>) to submission. 	
	Post-Graduate Research Assistant (PhD Student), University of Glasgow SUPA School of Physics and Astronomy, University of Glasgow, Glasgow, UK 2014 – 2019 <ul style="list-style-type: none"> Developed Python code to analyse observations of the Sun with a telescope that was not designed for heliophysics; this enabled numerous highly-collaborative peer-reviewed publications. Generated a press-release image that was published by numerous news outlets, included in books, and is one of the five iconic images from <i>NuSTAR's first five years in space</i>. Analysed time-series with Fourier analysis, wavelet analysis, and local intermittency measure. Studied the temperature distribution of the solar atmosphere through the recovery of an ill-posed inverse problem using ridge regression, Markov chain Monte Carlo (MCMC), and sparse inversion (basis pursuit). 	
	Researcher , NASA Frontier Development Lab (FDL) SETI Institute/NASA Ames Research Center, Mountain View, CA, USA 2018 <ul style="list-style-type: none"> Cleaned and curated 12 PB of raw (science) data to produce a 6.5 TB ML-ready data set. Implemented and modified CNNs such as U-Net, AlexNet, and ResNet to predict a 14-element vector (spectral line intensities) from narrowband images ($4096 \times 4096 \times 9$). Nowcast spectral line intensity with median absolute relative uncertainties of less than 1.6% per emission line using a CNN augmented with a Multi-Layer Perceptron (MLP), saving \$280M on a new instrument; the results were published in <i>Science Advances</i> (a high-impact journal). Developed and wrote an ebook chapter (Jupyter Notebook) on how to implement a 1×1 CNN in PyTorch to solve an ill-posed inverse problem (supervised learning; $10\times$ speed increase). 	
EDUCATION	PhD Physics , University of Glasgow, UK 2014 – 2019	
	MPhys Physics & Astrophysics (First-Class Honours), University of Southampton, UK Visiting Student: Harvard University; Smithsonian Institution; NASA Goddard Space Flight Center 2010 – 2014	
ADDITIONAL SKILLS	Adobe InDesign, Algorithms, Bash, Computer Vision, Data Analysis, Data Science, Data Visualization, Experimental Design, Git, Google Cloud Platform, Jupyter, MCMC, Python (matplotlib, pandas, scikit-learn), PyTorch, R, Shell scripting.	