

Taylor Faucett

Senior Machine Learning Engineer / Physicist

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Professional Summary

Senior Machine Learning Engineer and physicist with 5+ years leading ML development for 3D perception, scene understanding, and geometric reasoning in robotics and advanced manufacturing. Build production ML systems that create learned embeddings from CAD/mesh/point-cloud data, design dataset curation and labeling pipelines, and deploy real-time inference on edge hardware. Deep expertise in geometric deep learning (PointNet-style models, 3D CNNs, graph networks), physics-informed models for material behavior, and multimodal pipelines integrating geometry with LLM-driven agents.

Education

2015 - 2021	University of California, Irvine , Irvine, CA <i>Ph.D. in Physics</i>
2011 - 2015	University of Hawaii, Manoa , Honolulu, HI <i>M.S. in Physics</i>
2005 - 2009	Westminster College , Salt Lake City, UT <i>B.S. in Physics, Minor in Mathematics and Music</i>

Professional Experience

06/2022 - Present	Machina Labs <i>Senior Machine Learning Engineer</i> <ul style="list-style-type: none">- Lead end-to-end ML for industrial robotic sheet-metal forming, from dataset curation and labeling pipelines through model design, evaluation, and deployment on edge hardware for real-time inference.- Train real-time anomaly detection models deployed on edge devices, enabling in-process quality monitoring and early fault detection during robotic forming operations.- Develop physics-informed ML models trained on material stamping literature to predict failure likelihood, incorporating material properties and forming parameters to prevent defects.- Build mesh adjustment models for Double-Sided Incremental Forming (DSIF) that learn springback compensation—predicting geometry corrections so parts resolve to their final desired shape after elastic recovery.- Create learned embeddings and representations from 3D geometry (meshes, point clouds, CAD) that encode spatial structure and physical constraints for downstream reasoning and path planning.- Build multimodal pipelines combining geometric context with language-based interfaces for LLM-driven robot operation, focusing on constrained action spaces and operational safety.- Own data infrastructure (ETL, dataset versioning, metrics/dashboards, monitoring) and mentor junior engineers on ML best practices and geometric data pipelines.- Partner with robotics, controls, and software teams to integrate ML into safety-critical systems, including CI/CD and infrastructure-as-code for deployed robotic cells.
06/2015 - 06/2022	University of California, Irvine <i>Graduate Research Assistant & Postdoctoral Researcher</i> <ul style="list-style-type: none">- Researched deep learning and computer vision methods for particle-physics detectors, focusing on robust classification, reconstruction, and anomaly detection under real-world constraints.- Developed techniques for interpreting ML models and relating learned features to underlying physical mechanisms, improving trust and scientific insight.- Built end-to-end ML pipelines for large simulated and experimental datasets, covering data generation, preprocessing, feature engineering, model design, hyperparameter optimization, and uncertainty-aware evaluation.- Collaborated across international experimental collaborations and communicated ML results to both domain experts and non-specialists.

08/2011 - 05/2015	University of Hawaii, Manoa <i>Graduate Research Assistant</i> - Designed and implemented numerical simulations of Bose–Einstein condensates for a DoD-funded lattice-gas quantum computing experiment, bridging physics models and high-performance code. - Integrated analysis and trigger software with FPGA-based readout for a neutrino telescope, working across hardware, firmware, and scientific analysis teams.
09/2009 - 05/2011	Northrop Grumman Aerospace <i>Systems Engineer</i> - Designed and deployed secure precision-time (PTP) radio networks for U.S. Air Force installations, focusing on reliability, timing accuracy, and security. - Held an active Secret security clearance.

Selected Open-Source & Personal Projects

2024 - Present	Rust 3D Geometry & Robotics Data Stack <i>Private (active development; details available upon request)</i> - Developing a Rust-native alternative to PCL/Open3D for performant, safe mesh/point-cloud processing and geometry utilities. - Designed for interoperability with robotics workflows and telemetry/visualization platforms (ROS2, Foxglove, Rerun), with a focus on clean APIs and production deployment.
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Publications

1. Faucett, T. Decoding Black Box Models to Find New Physics at the LHC. Ph.D. Dissertation, University of California, Irvine (2021). <https://escholarship.org/uc/item/63x9r13b>
2. Faucett, T., Hsu, SC. & Whiteson, D. Learning to identify semi-visible jets. J. High Energ. Phys. 2022, 132 (2022). [https://doi.org/10.1007/JHEP12\(2022\)132](https://doi.org/10.1007/JHEP12(2022)132)
3. Faucett, T., Thaler, J., Whiteson, D. Mapping machine-learned physics into a human-readable space. Phys. Rev. D 103, 036020 (2021). <https://doi.org/10.1103/PhysRevD.103.036020>
4. Collado, J., Faucett, T., Witkowski, E. et al. Learning to isolate muons. J. High Energ. Phys. 2021, 200 (2021). [https://doi.org/10.1007/JHEP10\(2021\)200](https://doi.org/10.1007/JHEP10(2021)200)
5. Collado, J., Faucett, T., Howard, J. et al. Learning to identify electrons. Phys. Rev. D 103, 116028 (2021). <https://doi.org/10.1103/PhysRevD.103.116028>
6. Baldi, P., Cranmer, K., Faucett, T. et al. Parameterized neural networks for high-energy physics. Eur. Phys. J. C 76, 235 (2016). <https://doi.org/10.1140/epjc/s10052-016-4099-4>

Technical Skills

	LANGUAGES	●●● Python	●●○ Rust	●●○ JS/TS
		●●○ SQL	●●○ Bash/Shell	●●○ C++
	ML ENGINEERING	●●● PyTorch	●●● NumPy	●●● Pandas/Polars
		●●● scikit-learn	●●○ TensorFlow/Keras	●●○ ONNX
		●●○ MLflow	●●○ TensorBoard	●●○ Weights & Biases
		●●○ Optuna	●●○ Torch Lightning	●●○ Ray
3D GEOMETRY & CAD		●●● Mesh/Point Cloud	●●● Coordinate Frames & Kinematics	●●○ CAD / 3D Formats (STEP, STL, OBJ, PLY)
		●●○ SDFs / Occupancy Grids		

GEOMETRIC	DEEP LEARNING	••○ PointNet / PointNet++ •○ 3D Transformers	••○ 3D CNNs •○ Diffusion Models (3D)	••○ Graph Networks
	VISION & PERCEPTION	••• 3D Pipelines ••• Real-time Detection	Perception Anomaly	••• Scene Understanding ••○ OpenCV ••○ Sensor Fusion (3D + time-series)
	ROBOTICS	••• ROS2	••○ LLM-Assisted Robot Operation	••○ Edge/Embedded ML
	PHYSICS & SIM	••○ Material behavior forming	••○ Physics-informed ML	••○ Constraints-based optimization
	DATA & CLOUD	••• ETL Pipelines ••○ Azure/AWS	••• Kafka/RabbitMQ ••○ PySpark	••• Databricks ••○ InfluxDB
	DEVOPS	••• Docker ••• CI/CD	••○ Kubernetes ••• Linux/Unix	••○ Terraform ••• Git

Honors & Awards

2020	Chateaubriand Fellowship , Chateaubriand Fellowship Prestigious fellowship awarded by the French Embassy to support research in machine learning and AI at a leading French institution.
2016-2018	NRT-DESE: Team Science for Integrative Graduate Training in Data Science and Physical Science , National Science Foundation 2 year NSF research grant for work in the interdisciplinary field of machine learning and the physical sciences.