

# Taylor Faucett

Senior Machine Learning Engineer / Physicist

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## Professional Summary

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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

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

## Professional Experience

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06/2022 - Present	<b>Machina Labs</b> <i>Senior Machine Learning Engineer</i> - 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	<b>University of California, Irvine</b> <i>Graduate Research Assistant &amp; Postdoctoral Researcher</i> - 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	<b>University of Hawaii, Manoa</b> <i>Graduate Research Assistant</i> <ul style="list-style-type: none"> <li>- 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.</li> <li>- Integrated analysis and trigger software with FPGA-based readout for a neutrino telescope, working across hardware, firmware, and scientific analysis teams.</li> </ul>
09/2009 - 05/2011	<b>Northrop Grumman Aerospace</b> <i>Systems Engineer</i> <ul style="list-style-type: none"> <li>- Designed and deployed secure precision-time (PTP) radio networks for U.S. Air Force installations, focusing on reliability, timing accuracy, and security.</li> <li>- Held an active Secret security clearance.</li> </ul>

## Selected Open-Source & Personal Projects

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2024 - Present	<b>Rust 3D Geometry &amp; Robotics Data Stack</b> <i>Private (active development; details available upon request)</i> <ul style="list-style-type: none"> <li>- Developing a Rust-native alternative to PCL/Open3D for performant, safe mesh/point-cloud processing and geometry utilities.</li> <li>- Designed for interoperability with robotics workflows and telemetry/visualization platforms (ROS2, Foxglove, Rerun), with a focus on clean APIs and production deployment.</li> </ul>
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## Publications

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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

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

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

## Honors & Awards

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