

Vishal Pramod Kasliwal

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Experience

Intel Corporation - Senior Deep Learning Software Engineer

SANTA CLARA, CA

I work in the Machine Learning Performance (MLP) organization in the Machine Learning Distributed Compute (MLDC) group on accelerating Deep Learning workloads on Intel's Xe-HPC discrete accelerator cards for Deep Learning & High-Performance Computing. My role focuses on pre-Si performance optimization via hardware-software co-design. I develop & test computation- & communication-kernels using hardware simulators.

Hardware-Software Co-Design

April '19 – present

My responsibilities include

- defining architectural features for enhancing the architecture for Deep Learning workloads.
- driving adoption of best practices, i.e. architecture-specific know-how, in software products.
- evaluating & projecting pre-Si Deep Learning workload performance.
- creating the overall strategy for distributing Deep Learning workloads across multiple-cards & -nodes i.e. scale-up/scale-out strategy.

Wave Computing - Senior Staff Research & Development Software Engineer

CAMPBELL, CA

Wave Computing is developing the next-generation of solutions for speeding up Deep Learning applications using Dataflow Processing Units (DPUs), which contain thousands of interconnected dataflow Processing Elements (PEs). DPUs power Wave Computing's custom appliance for developing, testing, and deploying Deep Learning models. I use to develop the compute- and data movement- software kernels which are executed by Wave Computing's Dataflow Processing Units (DPUs) for Deep Learning acceleration.

Deep Learning Kernel- & Library-Development

Dec '17 – April '19

I have worked extensively on compute- & communicate-kernels, tools, and the supporting library for enabling Deep Learning workloads on Wave's DPUs.

- Developed compute kernels for various Deep Learning layers such as Average Pooling, Convolutions, Feed-Forward, Activations, Concatenation & fork, etc...
- Developed a tool for visualizing the place & route performed by JitPR.
- Owner of the library of routines for performing IEEE 754 rounding.
- Owner of the library of routines for changing precision.
- Authored various block Matrix Multiplication operations.
- Authored various multi-byte addition operations.
- Authored various multi-byte shift operations.

Tools Development

Nov '18 – Apr '19

I have developed tools for

- Simulating (functional & performance) Deep Learning kernels.
- Visualizing the placement & routing performed by JitPR.

Management Role

March '18 – April '19

I assisted my superior with the management of the kernel team consisting of seven engineers. My duties included

- maintaining the schedule of work being done by the members of the team.
- assisting in the planning of new work items.
- identifying & interviewing new compute-kernel team candidates.
- report on the progress of the compute-kernel team to the VP of product engineering.

Major accomplishments include

- I developed a workflow for creating new compute-kernels.
- I integrated the schedule of work into a JIRA managed project.

The outcome of my efforts is better project management leading to a significant reduction in the time taken to develop new compute-kernels.

Colfax International - High Performance Computing (HPC) Research Engineer

SUNNYVALE, CA

HPC Computational Fluid Dynamics (CFD) Application Development

Mar '17 – Dec '17

HPC consulting project. I parallelized a CFD simulation code written in C for a client in the oil & gas sector resulting in a ~10X speedup. I have also refactored the client's codebase to enable independent time-evolution in different regions of the simulation via domain-decomposition methods.

C/C++ Compiler Analysis

Nov '17

HPC research project. I investigated suitability of C++ compilers for HPC applications. I developed optimized scientific computational kernels and investigated the performance obtained by each compiler from each kernel. I analyzed compiled binary code to determine reason for differences in performance. A technical report of my findings can be obtained at [Colfax Research](#).

Intel Advisor Lecture

May '17 – Jul '17

I presented a lecture on [Intel Advisor](#) for the Stanford University course ME 344: Introduction to High Performance Computing on July 20th, 2017.

Large Synoptic Survey Telescope (LSST) Data Management (Princeton University)

PRINCETON, NJ

Postdoctoral Research Associate

Sept '15 – Feb '17

LSST Data Management is building a C++ & Python software stack to analyze raw imaging data from LSST. I worked on the software stack to add functionality, documentation, & tests. I developed & implemented algorithm to propagate covariance when stacking images and worked on techniques for optimal image stacking & differential chromatic refraction. I worked on a machine-learning based star-galaxy classifier and on converting the LSST stack to use py.test.

Department of Physics & Astronomy (University of Pennsylvania)

PHILADELPHIA, PA

Postdoctoral Researcher

Sep '15 – Feb '17

I developed and implemented a parallelized Bayesian algorithm to estimate orbital parameters from stochastic light curves of binary supermassive black holes. I also developed and implemented Python framework to automatically wrangle astronomical time-series data from a variety of sources including web-servers, SQL servers, data servers and local data files.

Principle Developer

Sept '15 – Feb '17

I architected and implemented [kālī](#), an open-source high performance library to model stochastic time-series data in a Bayesian framework. [kālī](#) is capable of modeling time-series data as variants of C-ARMA processes (a type of Gaussian random process). Written primarily in C++ and exposed to Python using Cython, [kālī](#) uses scikit-learn for machine learning, Intel MKL for fast linear algebra, Intel Bull Mountain technology for hardware random number generation, & OpenMP 4.0 for vectorization & parallelization. [kālī](#) is being used to study astronomical time-series data by multiple research groups at Caltech, UPenn, & Drexel.

Department of Physics (Drexel University)

PHILADELPHIA, PA

AGN Variability Analysis

June '09 – Aug '15

I developed C++ software for Intel Xeon Phi accelerator cards to model AGN variability. I developed vectorized & parallelized the C++ pipeline to forward-model and fit data to model using MLE of 2nd-order statistics.

LSST Photo-z Analysis

Sept '08 – May '09

I used MLE & machine learning (neural networks) to establish optimal y-band filter for LSST galaxy photo-z distance estimation.

Department of Physics (Virginia Commonwealth University)

RICHMOND, VA

Adjunct Instructor

Jun '07 – Aug '08

I taught *Introduction to Astronomy* course.

AFM Image Analysis

Aug '05 – May '07

I implemented an IDL pipeline to analyze AFM images of silicon surfaces etched using oxygen.

Department of Physics (University of Richmond)

RICHMOND, VA

Cosmic Microwave Background Analysis

May '03 – May '05

I used IDL to perform statistical tests of the utility of the bispectrum for detection of non-Gaussianity in the CMB.

Education

Drexel University

PHILADELPHIA, PA

PhD. in Physics

2008 – 2015

Probing AGN Accretion Physics through AGN Variability: Insights from *Kepler*

Virginia Commonwealth University

RICHMOND, VA

M.S in Physics & Applied Physics

2005 – 2007

CAFM Studies of Epitaxial Lateral Overgrowth GaN Films

Certifications

deeplearning.ai

COURSERA.ORG

Deep Learning Specialization*Certificate earned on June 24, 2018.***Neural Networks and Deep Learning***Certificate earned on November 4, 2017.***Improving Deep Neural Networks: Hyperparameter tuning, Regularization and Optimization***Certificate earned on November 26, 2017.***Structuring Machine Learning Projects***Certificate earned on December 17, 2017.***Convolutional Neural Networks***Certificate earned on January 22, 2018.***Sequence Models***Certificate earned on June 24, 2018.*

Skills

Technical expertise: My expertise lies in the research, design & implementation of high-performance scientific/numerical software on exotic, highly-parallel, and distributed hardware. Current interests include hardware-software co-design, particularly in the area of AI accelerators. I earned my doctorate in Astrophysics by applying mathematical & statistical analysis and machine learning to complex time-domain data. While I am familiar with developing Deep Learning applications using TensorFlow & PyTorch, my speciality lies in knowledge of the innards of these frameworks & their execution on various accelerators such as GPUs, DPUs, etc.... Used to working with(in) teams, I am fond of using Agile methodologies (Scrum) and continuous integration (Jenkins) to manage & deliver on-time & bug-free software. I enjoy writing C-For-Metal, OpenCL, FlowGraph, C, C++, & Python, and am learning Julia/CUDA/Go/IO/Prolog. I have excellent knowledge of parallelization technologies: OpenMP, MPI, C++11 threads, & POSIX threads, & hardware architecture. Over the years, I have gained extensive experience programming on novel computing platforms such as Intel's Xe-HPC accelerators, Wave Computing's Dataflow Processing Units (DPU) & Intel's Xeon Phi (Knight's Corner & Knight's Landing). I have extensive knowledge & experience with various programming toolchains including the Intel toolchain, GNU toolchain, LLVM toolchain, PGI toolchain, AOCC toolchain, Valgrind, gdb, make, SCons etc.... My preferred development platform is Linux (~15 years of development experience) although I have also developed on Windows (1 year of dev-ex), & Mac OSX (7 years of dev-ex). I am very comfortable with writing in L^AT_EX (~15 years of experience) and have a good understanding of the UNIX programming environment and tools (memory-management, process spawning, etc...)

Public speaking: With years of experience delivering highly technical talks to both expert & general audiences, I am comfortable with public speaking & outreach.

Natural languages: English (*native language*) and Hindi (*native language*).

Service

The National Science Foundation Served on a grant review panel for the Division of Astronomical Sciences.

The Astrophysical Journal Peer reviewed multiple publications.

Monthly Notices of the Royal Astronomical Society Peer reviewed a publication.

Publications

A Performance-Based Comparison of C/C++ Compilers Colfax Research, 2017
Science-driven Optimization of the LSST Observing Strategy arXiv, 2017
Large Synoptic Survey Telescope Galaxies Science Roadmap arXiv, 2017
Extracting Information from AGN Variability MNRAS, 470, 3, 3027-3048, 2017
The LSST Data Management System Proceedings of ADASS XXV, 2015
Do the Kepler AGN light curves need reprocessing? MNRAS, 453, 2075, 2015
Are the variability properties of the Kepler AGN light curves consistent with a damped random walk? MNRAS, 451, 4328, 2015
Thirty Meter Telescope Detailed Science Case: 2015 <http://arxiv.org/abs/1505.01195>, 2015
AFM and CAFM studies of ELO GaN films Proc. SPIE 6473, 647308, 2007
Local electronic and optical behaviors of a-plane GaN grown via epitaxial lateral overgrowth Appl. Phys. Lett., 90, 011913, 2007

Grants

Kepler Guest Observer Program Co-Investigator on Kepler Guest Observer Program accepted proposals K2 GO16088, K2 GO14088, K2 GO12013, K2 GO8052, & K2 GO10052
NASA Grant NNX14AL56G Helped write proposal for awarded NASA Grant NNX14AL56G. Grant was used to fund my Ph.D. research.

Presentations

Webinar on Intel & A Day in the Life of a Deep Learnin Software Engineer HiCounselor, September 24th, 2019, Sunnyvale, CA
Applications of High Performance Computing in Artificial Intelligence Rajasthan Student Startup Exposure Program 2018 (RSSEP2018), September 8th, 2018, San Jose, CA
Intel Advisor Stanford University ME344: Introduction to High Performance Computing, July 20th, 2017, Stanford, CA
Optical Variability Signatures from Massive Black Hole Binaries 229th Meeting of the American Astronomical Society, 2017, Grapevine, TX
Extracting Information From AGN Variability: an LSST AGN Collaboration Proposal 2017 LSST AGN Science Collaboration Roadmap Development Meeting, 2017, Grapevine, TX
Extracting Information from AGN Variability 2016 KARL LSST Workshop, November 2016, Louisville, KY.
Surveying the Dynamic Sky with the LSST 2016 KARL LSST Workshop, November 2016, Louisville, KY.
AGN Variability: Insights from Kepler 2016 Hotwiring the Transient Universe V Meeting, October 2016, Villanova, PA.
Probing Accretion Processes through Variability 2016 TMT Science Forum 'International Partnership for Global Astronomy', May 2016, Kyoto, Japan.
AGN Variability: Insights from Kepler Princeton HSC Science Discussion Series, March 2016, Princeton, NJ.
AGN Variability on Short Timescales: What does Kepler tell us about AGN Variability? 2015 TMT Science Forum 'Maximizing Transformative Science with TMT', June 2015, Washington, DC.
What can Kepler tell us about AGN variability? 225th Meeting of the American Astronomical Society, January 2015, Seattle, WA.
Do Kepler AGN Light Curves Exhibit a Damped Random Walk? 24th Meeting of the American Astronomical Society, June 2014, Boston, MA.
The Bispectrum of Galactic Dust: Implications for Microwave Background non-Gaussianity 204th Meeting of the American Astronomical Society, May 2004, Denver, CO.

Conferences

HotChips 2020 on behalf of Intel Corporation August 16th - 18th, 2020 San Jose, CA
The Next AI Platform 2020 on behalf of Intel Corporation June 25th, 2020 San Jose, CA
The Next FPGA Platform 2020 on behalf of Intel Corporation Jan 22nd, 2020 San Jose, CA
The Next AI Platform 2019 on behalf of Intel Corporation May 9th, 2019 San Jose, CA
Super Computing 17 (SC17) on behalf of Colfax International November 12th - 17th, 2017 Denver, CO