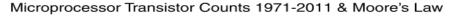
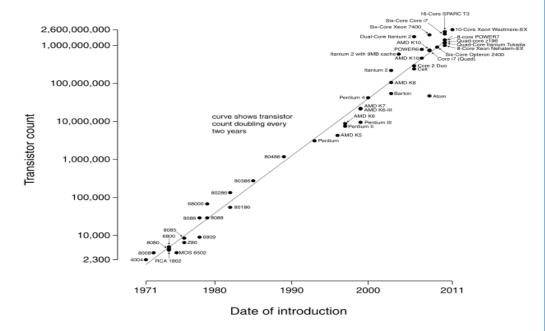


How we started

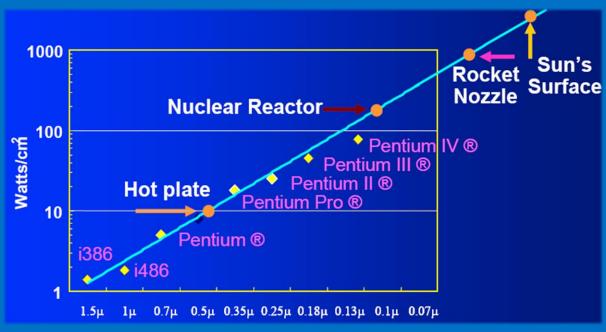




Source: Wikipedia



Tectonic Shift



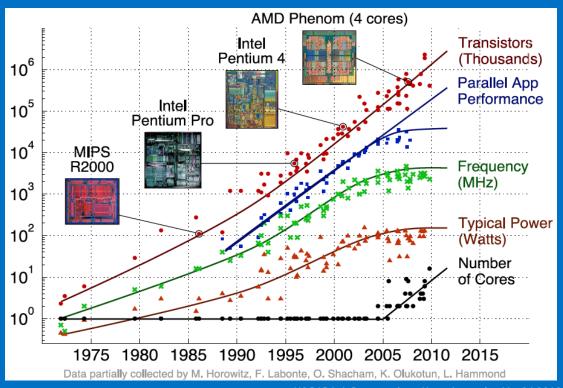
Source: Fred Pollack, Keynote - MICRO'32

Pollack's rule: performance increase due to microarchitecture advances is roughly proportional to square root of increase in complexity [area]

Power Consumption limits single-thread performance



New Computing Paradigm



UC PAR Lab Presentation - Krste Asanovic - May24,2010

Future improvements in performance will require taking advantage of parallelization and specialization techniques.



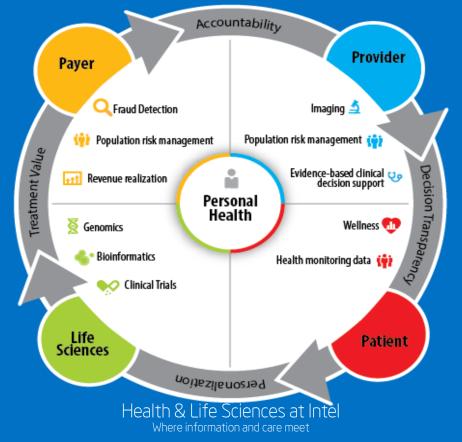
Parallelization and Specialization

- <u>Parallelization</u> Run computation on many low-power cores
- Specialization Run computation on most energy-efficient hardware
- Hardware Repertoire
 - Symmetric Multiprocessing
 - Vector Units Single Instruction, Multiple Data (SIMD)
 - New specialized instructions (e.g., AES-NI)
 - Integrated graphics processor
 - Heterogeneous Computing Co-processor
 - GPUs, Xeon Phi
 - Fixed logic accelerators Offload
 - Programmable logic FPGA

Designing for this complex ecosystem requires deep understanding of workloads and tighter collaboration with domain experts and software developers.



Personalized Medicine = Complex Big Data and Compute Ecosystem





Life Sciences:: Key Industry Challenges and Solutions

- Many (most) applications are single-threaded, single address space
 Intel is delivering optimizations working with open source
 community, developing NGS+HPC curriculum
- Some algorithms scale poorly with the size of the problem. Large data sets exceed available memory and storage
 - Innovations in acceleration, compute, storage, networking, security, and *-as-a-service.
- International collaboration is an imperative, bioinformatics expertise is scarce
- Intel is working closely with the ecosystem to address enterprise to cloud transmission of terabyte payloads
- Databases are distributed, data is siloed and will likely stay that way
 Tools like Hadoop, Lustre, Graphlab, In-Memory Analytics,
 Security etc.



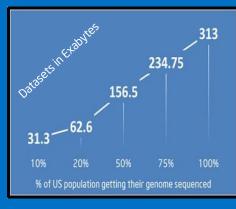


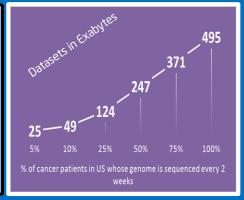






Genomics - Big Data Problem





313 Exabytes if everyone in the US has their genes sequenced

495 Exabytes

if every cancer patient in the US has their genes sequenced every 2 weeks.

Source: Knights Cancer Institute, Oregon Health Sciences University & Intel

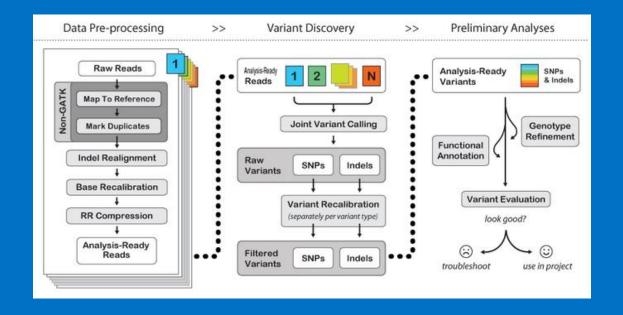
This is a key area with a large growth potential.

Goal is to anticipate demand for compute, provide efficient solutions, and help grow the market.

Energy and Total Cost of Operation are key



DNA Pipeline - GATK Best Practices







DNA Pipeline: BWA+GATK

Whole Genome Sample: ~65x Coverage

Step Tool	# of Threads	Runtime (hours)
Read Alignment (bwa)	16	8
Sampe (bwa)	1	24
Import (samtools)	1	11
Sort + Index (samtools)	1	14.5
MarkDuplicates	1	11.5
(picardtools) + Index UnifiedGenotyper* (GATK)	16	7.5
SomaticIndelDetector (GATK)	1	3
RealignerTargetCreator (GATK)	16	0.8
IndelRealigner* (GATK) + Index	1	17.5
BaseRecalibrator*(GATK)	1	62
PrintReads* (GATK) + Index + Flagstat	1	25
TOTAL (hours)		177



Step	# of Threads	Runtime (hours)
Read Alignment (bwa mem)	24	7
View (samtools)	24	2
Sort + Index (samtools)	24	3
MarkDuplicates (picardtools) + Index	1	11
RealignerTargetCreator (GATK)	24	1
IndelRealigner* (GATK) + Index	24	6.5
BaseRecalibrator*(GATK)	24	1.3
PrintReads* (GATK) + Index + Flagstat	24	12.3
TOTAL (hours)		44

Algorithmic Improvement Thread-level Parallelism

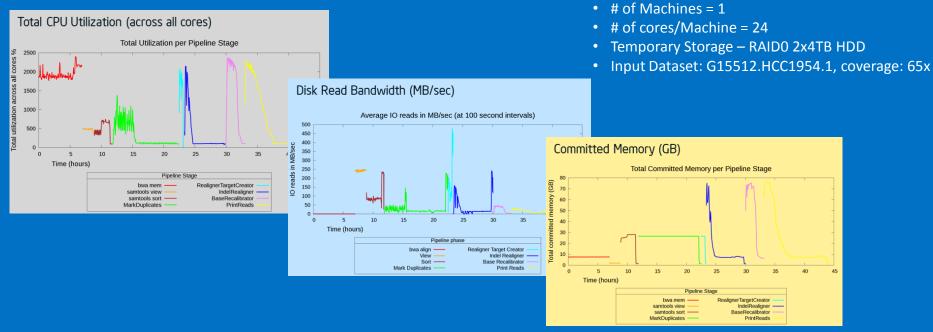
Cluster-level Parallelism





Profiling: Single Instance Run of GATK

GATK: Genome Analysis Toolkit



Average CPU utilization is very low. Most cores not being used

Average I/O bandwidth is very low. Application not I/O bound

Average memory footprint is small. Application not using memory available in newer systems

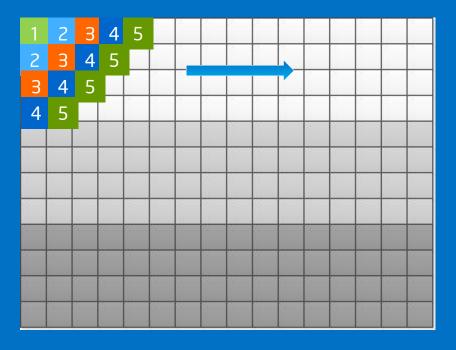


PairHMM Computation Kernel in Java

```
* Updates a cell in the HMM matrix
     * The read and haplotype indices are offset by one because the state arrays have an extra column to
hold the
     * initial conditions
     * @param indI
                                row index in the matrices to update
     * @param indJ
                               column index in the matrices to update
     * @param prior the likelihood editing distance matrix for the read x haplotype * @param transition an array with the six transition relevant to this location
    protected void updateCell( final int indI, final int indJ, final double prior, final double[]
transition) {
        matchMatrix[ind]][ind]] = prior * ( matchMatrix[indI - 1][indJ - 1] * transition[matchToMatch] +
                                                     insertionMatrix[indI - 1][indJ - 1] *
transition[indelToMatch] +
                                                     deletionMatrix[indI - 1][indJ - 1] *
transition[indelToMatch] );
        insertionMatrix[indI][indJ] = matchMatrix[indI - 1][indJ] * transition[matchToInsertion] +
                                                     insertionMatrix[indI - 1][indJ] *
transition[insertionToInsertion];
        deletionMatrix[indI][indJ] = matchMatrix[indI][indJ - 1] * transition[matchToDeletion] +
                                                     deletionMatrix[indI][indJ - 1] *
transition[deletionToDeletion];
```



PairHMM Wave-Front Computation in AVX





Improvements in GATK 3



Introducing GATK 3.0: New features and enhancements





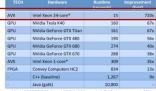
Mauricio Carneiro, Laura Gauthier, Bertrand Haas, Ami Levy-Moonshine, Ryan Poplin, David Roazen, Valentin Ruano Rubio, Khalid Shakir, Joel Thibault, Geraldine A. Van der Auwera, Eric Banks

Genomics Platform, Broad Institute of MIT and Harvard, Cambridge, MA, USA



Faster processing through heterogeneous compute

The haplotype caller is the best variant caller available, but it has been used only for small projects due to it's very long run time. We have re-implemented the main routines using alternative hardware: AVX, GPU and FPGA and are happy to report that speed is no longer an issue!



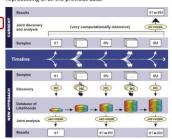
80x NA12878 PCR Free WGS 2x250 reads (all of chromosome 20)

- A deep whole genome can be "variant called" with the haplotype caller in under 1h in a single machine!
 SSE/AVX improvements are already available in most
- SSE/AVX improvements are already available in most machines older than 2007.
 We are evaluating the use of GPU, AVX and FPGA
- implementations for new and existing GATK tools.

This is just a preview of the three main features of the new GATK release. There are many more new features and tools that will be made available in the 3.x release cycle.

Single-sample calling pipeline for joint variant calling

Joint calling greatly empowers variant discovery but performance doesn't scale. Calling 65,000 exomes has proven to be an extremely difficult project and bumped into many of the limitations of the current framework. In addition every new sample sequenced requires a full reprocessing of all the previous data.



The <u>single-sample pipeline</u> enables fast and <u>incremental</u> variant calling while maintaining the power of joint analysis. Thanks to a new reference model.

- Discovery is done independently using the new reference model of the haplotype caller and can be done incrementally.
- Genotyping is done jointly and because it is a very cheap operation, it can use all previous data ever sequenced!

http://www.broadinstitute.org/gatk

Building a Best Practices pipeline for RNA-seq analysis

aatk 🖭

@gatk_dev

The GATK team has made an incredible contribution to the scientific community by providing a Best Practices pipeline for DNA sequencing. The pipeline became the standard around the world, and we are very proud of that. Unfortunately, RNA sequencing has no equivalent set of recommendations.



sequencing experiments. • We are using the same rigorous approach to RNA sequencing

- we are using the same rigorous approach to KNA sequencing data as we did for DNA.
- First draft of our Best Practices for RNA-seq publicly available in version 3.0.
- Development of new and better tools specifically designed for RNA-seq data processing and analysis will be unveiled in the next releases of the GATK. Stay tuned!
 - Acknowledgements



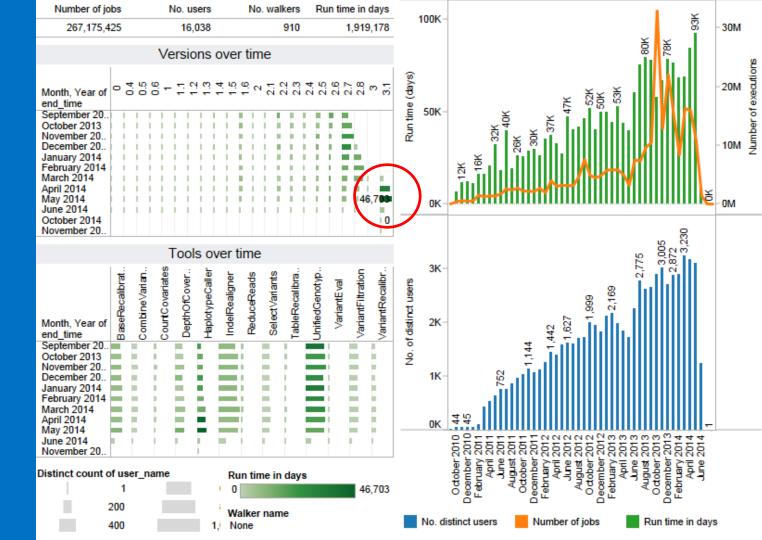


- Pair HMM Acceleration using Intel® AVX resulted in 970x speedup
 - Computation kernel and bottleneck in GATK Haplotype Caller
 - AVX enables 8 floating point SIMD operations in parallel

	Time (seconds)	Speedup C++/Java
Serial C++	1540	1x / 9x
1 core with AVX (Intra)	340	4.5x / 40.7x
1 core with AVX (Inter)	285	5.4x / 48.6x
24 cores with AVX (Inter)	14.3	108x / 970x
24 cores hybrid (Inter)	15.7	98x / 882x

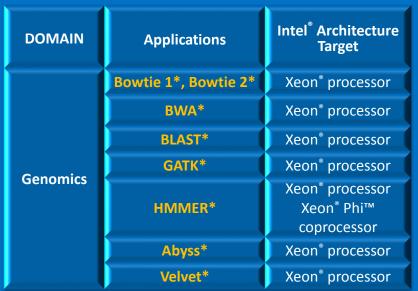


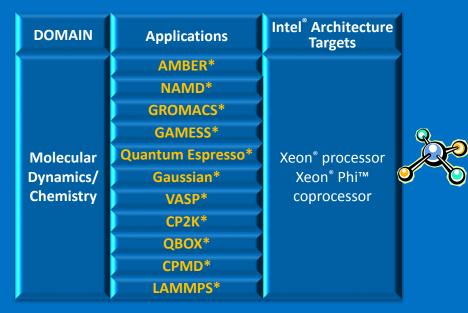
GATK downloads over time.



Applications and Workloads Optimized on Intel Architecture

- Focus on improving genomics, molecular dynamics pipelines
- Optimize individual applications (node and cluster); Work with code authors to release optimizations









 Challenge: Ayasdi Cure[™] analyzes highly complex, large data sets and relies on fast computation times to provide real-time output.

Solution:

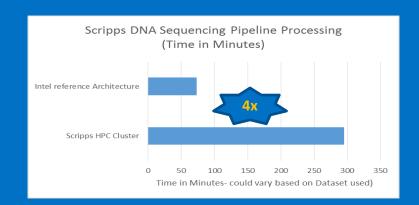
- Intel® AVX instructions four double-precision floating-point operations in parallel vs. one.
- Intel® MKL library accelerate filter computations
- Benefits: 400% performance increase in distance computation.





Scripps DNA Sequencing Pipeline

- Challenge: Processing times, Logistical Delays, Cluster complexity
- Solution: Intel® Xeon® E7-4800 series using SSDs
- Benefits: ~4x Improvement on processing times











Ultra High-Speed Networking Optimizations

 Challenge: Improving big data transfer to and from the backend data center

Solution:

- Optimize ultra high-speed (10+ Gbps) data transfer solutions built on Aspera's FASP ™ technology
- Intel® Xeon® E5-2600 (DDIO, SR-IOV)

Benefits:

- 300% improvement in transfer throughput
- Physical or virtual, LAN or WAN same transfer speeds

High Performance Scale-out Storage Challenge:

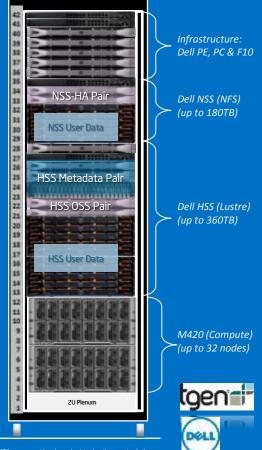
- Challenge: 10-15TB data added weekly, small fraction of overall storage capacity and need a system to scale, be flexible and efficient
- Solution: HPC-class storage, powered by Intel[®]
 Enterprise Edition for Lustre* software

Benefits:

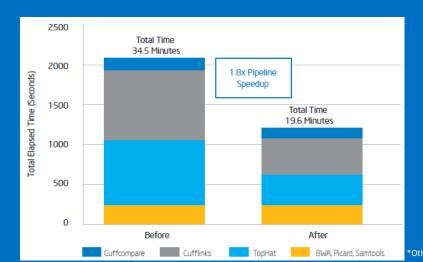
- Openess, global namespace
- Performance of upwards of 1 TB/s
- Virtually unlimited file system and per file sizes, and management simplicity



HPC Appliances for Life Sciences



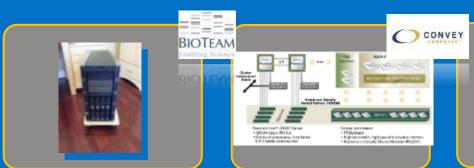
- Challenge: Experiment processing takes 7 days with current infrastructure. Delays treatment for sick patients
- Solution: Dell Next Generation Sequencing Appliance
 - Single Rack Solution; 9 Teraflops, Lustre File Storage; Intel SW tools
- Benefits: RNA-Seq processing reduced to 4 hour
- Includes everything you need for NGS compute, storage, software, networking, infrastructure, installation, deployment, training, service & support



** 2-socket Intel(R) Xeon(R) CPU E5-2687W / 3.1 GHz

*Other names and brands may be claimed as the property of others

Genomics & Clinical Analytics Appliances













Let us all make Personalized Medicine mainstream by 2020 ...

- www.intel.com/healthcare/bigdata
 - Paolo.Narvaez@intel.com

