

Massively parallel read mapping on GPUs with the q-group index and PEANUT Johannes Köster, Sven Rahmann Genome Informatics, University of Duisburg-Essen



We present the q-group index, a q-gram index variant with a particularly small memory footprint, and efficient parallel algorithms for building and querying the data structure. On top of the q-group index, we developed **PEANUT**, a massively parallel GPU-based solution to the read mapping problem. PEANUT is 2 to 10 times faster than its competitors while maintaining a comparable accuracy. The software is available at http://peanut.readthedocs.org.

Read Mapping Problem

Occurs with next-generation sequencing of DNA or RNA:



partitioned into streaming multiprocessors (SM)



- millions of small DNA or RNA reads are produced
- information about their origin in the genome is lost
- read mapping problem: find the likely origin of each read in a known reference genome
- optimal solution: calculating optimal alignments with Smith-Waterman algorithm (infeasible due to quadratic run-time)
- state of the art: use index data structures to find anchor points for alignment (BWT/FM-Index, q-gram index)

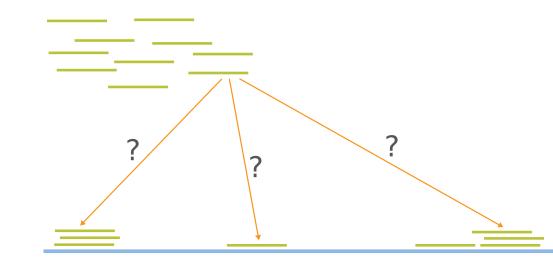


Fig. 1: The read mapping problem.

- each SM executes the same instruction of 32 threads in parallel • branching (if-else) breaks parallelism
- memory small (e.g. 3GB) and rather slow (because of small caches) • memory transfer between host and GPU slow

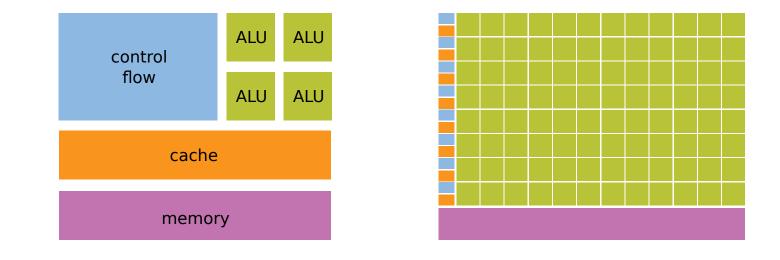


Fig. 2: CPU vs. GPU.

Q-Gram Index

- Search for q-gram matches between reads and reference. A q-gram is a subsequence of length q. • encode q-grams as integers: $ACGT = 11 \ 10 \ 01 \ 00 = 228$
- for text T, q-gram index consists of arrays (S, O)such that k-th occurrence position of q-gram g is

O[S[g] + k].

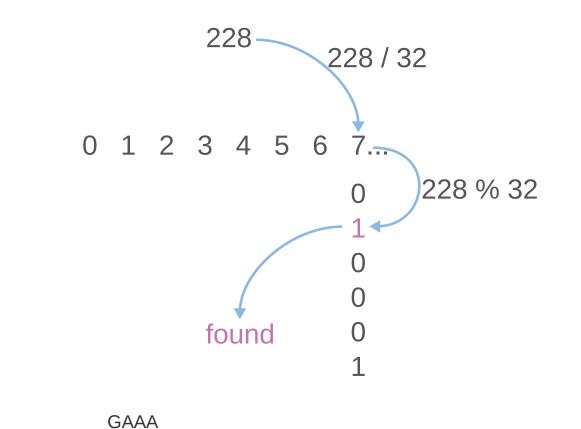
Q-Group Index

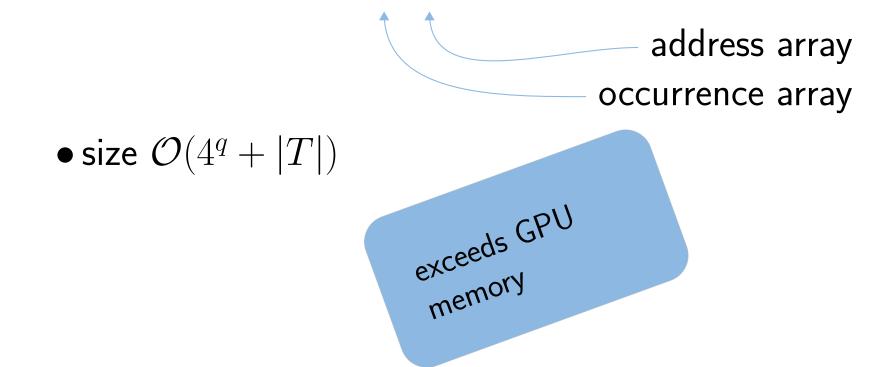
We introduce the q-group index, a variant of the q-gram index with small memory footprint.

• assign each q-gram g to a q-group $i = \lfloor g/w \rfloor$

• for text T, q-group index consists of arrays (I, S, S', O) such that k-th occurrence position of q-gram g is

> $O[S'[S[i] + popcount(I[i]\&(2^{j} - 1))] + k].$ — array of bit vectors





q-group address array q-gram address array occurrence array

• build and query in parallel using *population counts* and *prefix sums* (avoiding branching)

• worst case size: $2/w \cdot 4^q + \min\{4^q, |T|\} + |T|$



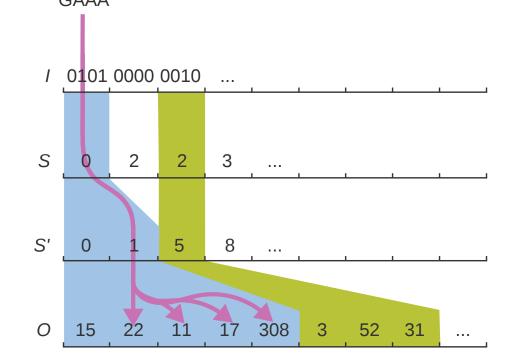


Fig. 3: Q-group index.

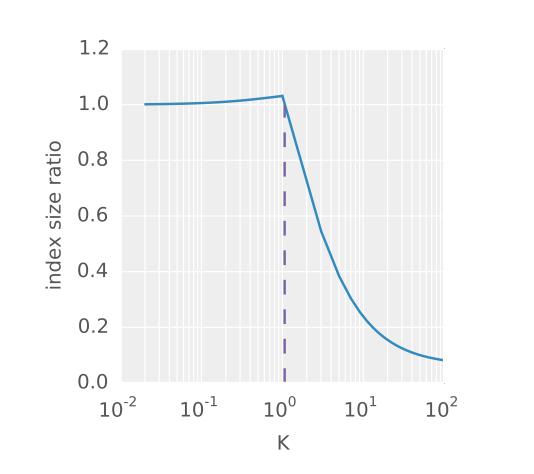


Fig. 4: Ratio between the size of the q-group index and the q-gram index for factor $K = \frac{4^{q}}{|T|}$ describing ratio between number of possible q-grams and text size.

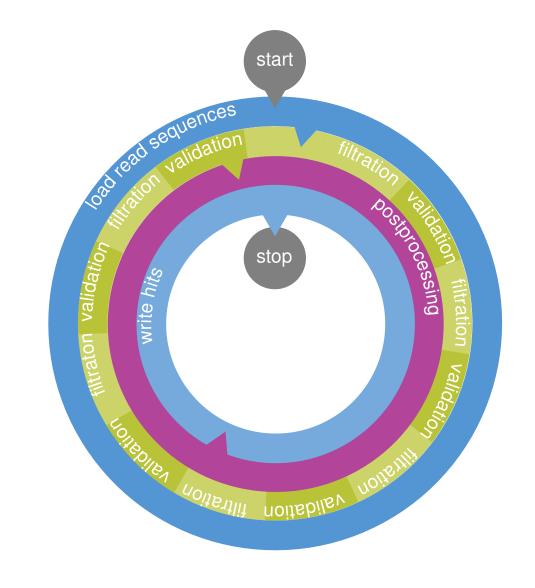


Fig. 5: PEANUT algorithm. Different layers run in parallel and are

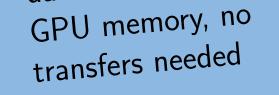
connected by queues.

PEANUT Algorithm

- 1. load reads into buffer
- 2. *build* a q-group index
- 3. *filter* hits between reference and index
- 4. validate hits with bit-parallel alignment algorithm (Myers G, 1999)
- 5. post-process hits
- 3. write hits

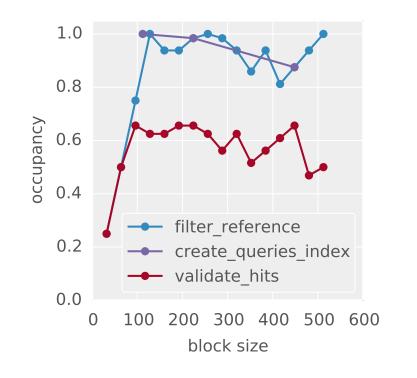
IO
GPU
CPU

step 2 to 4: data resides entirely in



Occupancy

• occupancy measures the saturation of GPU cores • high occupancy: ability to hide memory latency



Accuracy

PEANUT provides at least comparably good precision and

recall as other mappers.

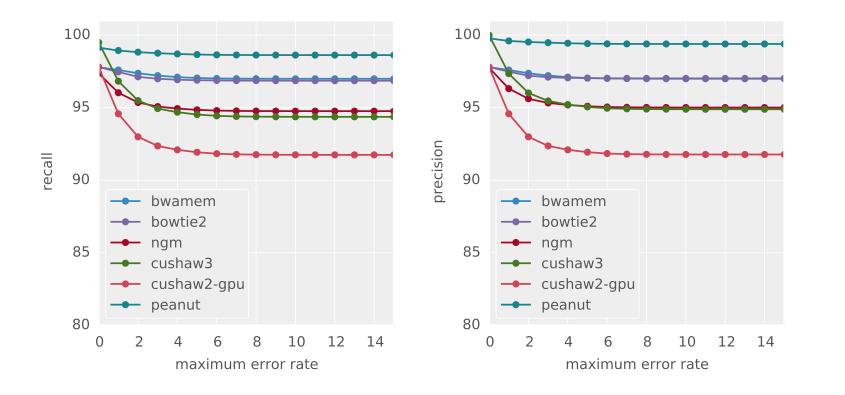


Fig. 7: Recall and precision vs. error rate.

Performance

• 50 million human paired-end Illumina HiSeq 2000 reads • 4-core Intel Core i7, NVIDIA Geforce 780, 16GB RAM

mapper	mode	time [min:sec]		
PEANUT	best-stratum	18:22	18:36	18:31
BWA-MEM	best-hit	36:46	36:33	36:35
Bowtie 2	best-hit	54:38	54:22	55:51
CUSHAW3	best-hit	390:20	390:15	390:41
CUSHAW2-GPU	best-hit	30:23	30:30	30:34
PEANUT	all-hits	254:43	254:49	254:19
RazerS 3	all-hits	900:27	901:33	900:50

Fig. 6: Occupancy vs. thread block size.

₽T_E**X** TikZ**poster**