

Evaluating MapReduce for Multicore and Multiprocessor Systems

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Google's MapReduce

□ A general-purpose environment for large-scale data processing

- Programming model (API) and runtime system for large clusters
- Functional representation of data parallel tasks

□ Easy to use & very successful within Google

 Indexing system, distributed grep & sort, document clustering, machine learning, statistical machine translation, ...

□ MapReduce supports

- Automatic parallelization and distribution
 - Abstracts parallelization, synchronization, and communication issues
- Fault tolerance
 - Task monitoring and replication
- I/O scheduling



WordCount Example [OSDI'04]





The Phoenix System

□ Question: is MapReduce applicable to multicore programming?

- What is the performance?
- Is the performance scalable & portable?
- Does it help with locality and fault management?
- How does it compare to other parallel programming approaches?

□ *Phoenix*: a shared-memory implementation of MapReduce

- Uses threads instead of cluster nodes for parallelism
- Communicates through shared memory instead of network messages
 - Works with CMP and SMP systems
- Current version works with C/C++ and uses P-threads
 - Easy to port to other languages or thread environments



The Phoenix API

□ System-defined functions

- int **phoenix_scheduler** (scheduler_args_t *args)
 - Initializes the runtime system
- void emit intermediate (void *key, void *val, int key size)
- void emit (void *key, void *val)

User-defined functions

- void (*map_t) (map_args_t *args)
 - Map function applied on each input element
- void (*reduce_t) (void *key, void **buffer, int count)
 - Reduce function applied on intermediate pairs with same key
- int (*key_cmp_t) (const void *key1, const void *key2)
 - Function that compares two keys
- int (*splitter_t) (void *input, int size, map args t *args)
 - Splits input data across Map tasks (optional)

Simple & narrow API. Similar to Google's API



The Phoenix Runtime

Orchestrates program execution across multiple threads

- Initiates and terminates threads (workers)
- Assigns map & reduce tasks to workers
- Handles buffer allocation and communication

□ Key runtime features

- Dynamic scheduling of tasks for load balancing
- Communication through pointer exchange (when possible)
- Locality optimization through granularity adjustment
- Support for failure recovery

Details of parallel execution are hidden from programmer

• Low-level threading, communication, scheduling, ...



Phoenix Execution Overview





Input Data Splitting



Divides input data to chunks for map tasks

- Small chunk \rightarrow map overhead; large chunk \rightarrow locality issues
- □ Phoenix: chunk size determined by cache size for locality
 - Or programmer provides custom splitter



Map Stage



Each task applies map function to an input chunk

- Phoenix: typically 100s of tasks multiplexed to available workers
- No reduce tasks are started before map tasks complete
- □ Intermediate pairs partitioned to reduce queues based on keys
 - Partitioning can introduce significant overhead!



Reduce Stage



□ Each task processes a key set from the reduce queues

- Dynamic scheduling used for reduce tasks as well
- □ A poor partition function can lead to significant imbalance
 - Default partition function based on key hashing
 - Programmer can provide custom partition function



Merge Stage



Combines reduce output queues to single sorted output

- May be unnecessary for some applications
- But merge time tends to be small compared to map/reduce time
- □ Phoenix: binary merging of reduce queues into single queue
 - Overhead increases with number of reduce tasks



Potential Performance Detractors

Significant detractors

- **<u>Partitioning overhead</u>**: communication and grouping requirements
- **Model overhead**: particularly due to calls to emit/emit_intermediate
- **<u>Key management</u>**: some apps do not naturally associate keys with data
- **<u>Repeated Map/Reduce invocations</u>**: necessary for some apps
- Practically insignificant issues
 - **Final merging & sorting**: insignificant compared to other tasks
 - **<u>Buffer management</u>**: extensive (re-)use of pre-allocated buffers
 - **<u>Reduce imbalance</u>**: handled through dynamic scheduling
 - Serialized input splitting: most map tasks involve non-trivial work



Phoenix Fault Tolerance

□ Focus: transient or permanent errors in workers

Error detection: worker time-out

• Execution time of similar tasks used as yardstick

Error recovery

- Restart or potentially re-assign affected tasks
- Handle input/output buffer management

□ Future work

- Fault tolerance for the scheduler
- Error detection and isolation through worker sandboxing



Evaluation Methodology

□ Shared-memory systems

- CMP: Niagara-based Sun Fire T1200 (8 CPUs, 4 threads/CPU)
- SMP: Sun Ultra E6000 (24 CPUs)
 - SMP results similar to CMP \rightarrow portable performance
 - See paper for details

□ 8 applications

- Domains: enterprise, scientific, consumer
- Three code versions: sequential, MapReduce (Phoenix), P-threads
 - Optimized independently

Experiments

- Talk: performance & scalability, Phoenix Vs. P-threads
- <u>See paper for</u>: dependency to data-set size, dependency to input task granularity, (soft & hard) fault injection experiment



Applications

- □ **<u>Word count</u>** determine frequency of words in documents
- String match search file with keys for an encrypted word
- Reverse Index build reverse index for links in HTML files
- □ Linear regression find the best fit line for a set of points
- □ <u>Matrix multiply</u> dense integer matrix multiplication
 - MapReduce version introduces coarse-grain coordinate variables
- □ <u>Kmeans</u> clustering algorithm for 3D data points
 - Multiple MapReduce invocation with translation step
- □ <u>PCA</u> principal component analysis on a matrix
 - MapReduce version introduces coordinate variables
- □ <u>Histogram</u> frequency of RGB components in images
 - There is no need for keys in original algorithm







Good scalability across all applications

- Absolute speedup depends on importance of various detractors
- Note for CMP: 1 core = 4 threads (4 workers)

□ Improved locality leads to significant improvements for some apps

At high core counts: some bandwidth saturation or load imbalance



Phoenix Vs. P-threads



Phoenix equal to P-threads if algorithm matches MapReduce model

- Note that P-threads' low-level API is more flexible
- □ Anecdote: we looked at Phoenix behavior to tune some Pthreads codes
- P-threads is better for algorithms that do not fit MapReduce model
 - Does not use keys, requires multiple MapReduce iterations, …



Conclusions

□ Phoenix: a shared-memory implementation of MapReduce

- MapReduce API and runtime system for C/C++
 - Uses threads instead of cluster nodes for parallelism
 - Communicates through shared memory instead of network messages
 - Dynamic scheduling, locality management, fault recovery, ...
- Scalable & portable performance, compares well to P-threads

Future work

- Improve queue structures
- Automatic configuration detection
- Better fault detection
- Experiment with more systems & apps



□ Want a copy of Phoenix?

• Will post the source code at *http://csl.stanford.edu/~christos*