Parallel Scalar Multiplication of Elliptic Curve Points

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Motivation

 Elliptic curves are commonly used in public-key cryptography

- Digital Signatures
- Symmetric Key Exchange

•Scalar multiplication of points on a curve is the most costly operation performed

Background – Finite Fields

- •A finite field on pⁿ is the set of integers in {0, pⁿ}, where p is a prime and n is some positive integer
- Two types of finite fields are of interest
 - ► Prime fields, where n=1
 - Uses regular arithmetic, modulo a prime p
 - ► **Binary fields**, where p=2

Uses polynomial arithmetic, modulo an irreducible polynomial p

Background – Polynomial Arithmetic on a Finite Field

- •The binary number $b_{n-1} ||b_{n-2}|| ... ||b_0$ represents the polynomial $\sum_{i=0}^{n-1} b_i x^i$
- •Arithmetic operations defined in terms of polynomials, with coefficients computed modulo 2
- •Squaring is efficiently achieved on binary fields
 - Inserting a 0 between consecutive bits of a number yields its square
 - ►O(n) time compared to O(n²) time for multiplication

Background – Non-Adjacent Forms

•A non-adjacent form (NAF) is an alternate representation for an integer k such that $k = \sum_{i=0}^{l-1} k_i 2^i$ where $k_i \in \{0, \pm 1\}$ and no two consecutive digits are nonzero

•A windowed NAF (wNAF) for k is the representation $_{k=\sum_{i=0}^{l-1}k_{i}2^{i}}$ such that $|k_{i}| < 2^{w-1}$ for a window size w, k_{i} is 0 or odd, and for any w consecutive digits, at most one is nonzero

Elliptic Curves

General elliptic curve equation

$$y^2 + axy + by = x^3 + cx^2 + dx + e$$

- •Two general types of curves are of interest:
 - •Prime curves: $y^2 = x^3 + ax + b$
 - **Binary curves:** $y^2 + xy = x^3 + ax^2 + b$

Binary curve with certain properties called Koblitz curves allows field squaring to replace less efficient point doubling in scalar multiplication, which will be particularly suitable for a parallel implementation

Elliptic Curve Coordinates

•Natural to think of curves and points in terms of affine coordinates (x, y) for geometric intuition and to describe algebraic properties

•Computation often more efficient when projecting on a higher dimensional space

►ie. Projective coordinates (x, y, z) from the affine coordinates (x/z, y/z)

•Compressed coordinates can be used to transmit points with minimal size

The x affine coordinate and a bit signifying the corresponding y value to use

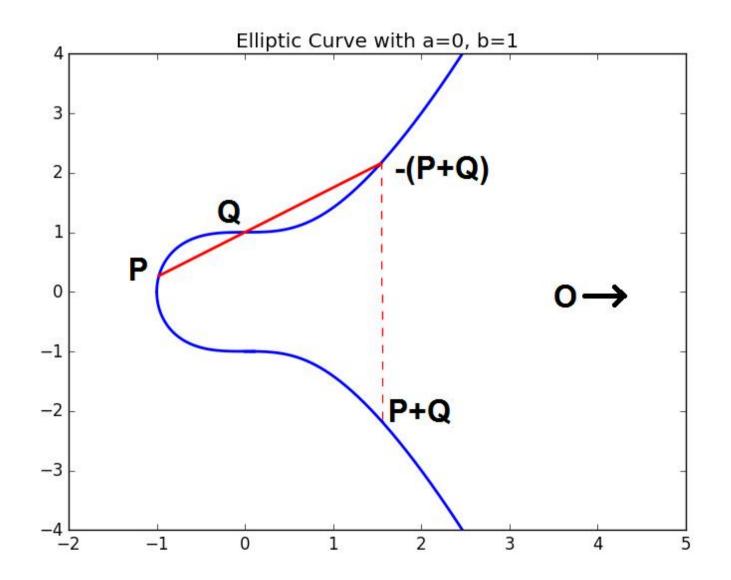
Prime Curves

•For a prime curve, if we have nonzero determinant $4a^3+27b^2 \neq 0 \pmod{p}$ we can define addition of points and form an abelian group:

- ► Closure
- Associativity
- Commutativity
- Identity Element (O, "point at infinity")
- Inverse Element (-P for a point P)

•Two basic point operations: point addition and point doubling

Prime Curves – Geometric Intuition



Prime Curves – Scalar Multiplication

•Basic approach is the "double-and-add" method to compute kP given $k=b_{n-1}||b_{n-2}||...||b_0$ the binary representation of k

```
Input: P, k=b_{n-1}||b_{n-2}||...||b_0

Output: Q = kP

Q=0

For i from 0 to n-1

Q=2Q

If b_i=1 then Q=Q+P

Return Q
```

Prime Curves – Scalar Multiplication

 More efficient by a constant factor to use a wNAF method:

```
Input: P, k
Output: Q =kP
Compute wNAF of k = \sum_{i=0}^{l-1} k_i 2^i
Precompute jP for j={1, 3, ..., 2<sup>w-1</sup>-1}
     Q=O
     For i from I-1 to 0
          Q=2Q
          if k_i > 0 then Q = Q + k_i P
          else if k_i \neq 0 then Q=Q-k_iP
     Return Q
```

Binary Curves

•Binary curves require $b \neq 0$ to define an abelian group

•General binary curves use same algorithms as prime curves to compute scalar multiplication

•Koblitz curves have a property which allows more efficient computation of scalar multiplication

Given a point (x, y) on the curve, (x^2, y^2) is also on the curve, and this can be used to replace point doubling by field squaring

Koblitz Curves – τ Operator

•Define the **T operator** such that $T(x, y)=(x^2, y^2)$ and TO=O

Recall that squaring on a finite field over 2^m can be computed efficiently

•Given a point P, we have $(\tau^2+2)P=\mu\tau P$ where $\mu=(-1)^{1-a}$ where τ^j is the τ operator applied j times

•From the above result, we can consider τ as the complex number satisfying $\tau^2+2=\mu\tau$

►
$$\tau = (\mu + \sqrt{-7})/2$$

Allows a scalar to be expressed in terms of T

Koblitz Curves – wtNAF

•A number $\kappa = r_0 + r_1 \tau$ on the ring $\mathbb{Z}[\tau]$ has a wTNAF representation $\kappa = \sum_{i=0}^{l-1} u_i \tau^i$ where $u_i = \{0, \alpha_{\pm 1}, \alpha_{\pm 3}, ..., \alpha_{\pm (2^{w-1}-1)}\}$

The $\alpha_i = \beta_i + \gamma_i \tau$ for each window size are chosen so that each precomputed point requires at most a single point addition and a single application of τ during precomputation

Koblitz Curves – wtNAF

•Computing the wτNAF representation for a scalar results in a representation that is too long in general – ~2m digits for an m-bit scalar

•To get a suitable length representation, find a complex number ρ' such that $\rho' \equiv k \pmod{\delta}$ where $\delta = (\tau^m - 1)/(\tau - 1)$ using partial modulo reduction

The equivalence ensures that $\rho'P \equiv kP$, where ρ' has a sufficiently short representation bounded in length by m+a+3

High probability of finding p, the shortest representation based on a chosen parameter C

Koblitz Curves – wtNAF Multiplication

•The wτNAF method is as follows:

```
Input: P, \rho' = \sum_{i=0}^{l-1} u_i \tau^i
Output: Q=p'P=kP
    Precompute P_{u} = \alpha_{u} P for u \in \{\pm 1, \pm 3, ..., \pm (2^{w-1}-1)\}
    O=O
    For I from I-1 to 0
         Q=tQ
         If u<sub>i</sub>≠0 then
             Let u be such that \alpha_{i} = u_{i} or \alpha_{i} = -u_{i}
             If u >0 then Q=Q+P
             Else Q=Q-P
    Return Q
```

Securing Against Side Channel Attacks

•The computation methods considered so far depends on the input scalar

- •Adversaries capable of side channel attacks, such as a timing attack, can exploit this to learn secret information
- •Using a Montgomery method modifies multiplication algorithms in a simple way to take fixed time independent of the input scalar size
 - Performance decreased by a constant factor
 - Montgomery ladder used for prime curves
 - Dummy variable used for Koblitz curves

- •Let k be an n-digit long scalar and suppose we have 2^m processors with $2^m \le n$
 - In binary representation for prime curves
 - In wtNAF representation for Koblitz curves
- •We can break k into 2^m parts:

$$k = k_{2^{m}}^{m} ||k_{2^{m}-1}^{m}|| ... ||k_{1}^{m}|$$

Then compute the smaller products in parallel

$$k_{2^{m}}^{m}P, k_{2^{m}-1}^{m}P, \dots, k_{1}^{m}P \Rightarrow Q_{2^{m}}^{m}, Q_{2^{m}-1}^{m}, \dots, Q_{1}^{m}$$

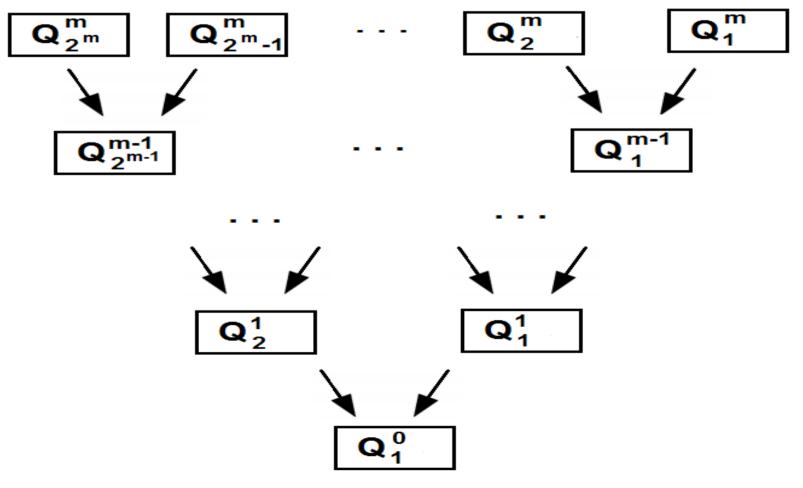
- •From these smaller products, we can then recursively recombine the Q values to obtain kP
 - For prime curves, we recombine via doubling $Q_{j/2}^{i} = 2^{|k_{j-1}^{i+1}|} Q_{j}^{i+1} + Q_{j-1}^{i+1}$
 - For Koblitz curves, we recombine via τ

$$Q_{j/2}^{i} = \tau^{|k_{j-1}^{i+1}|} Q_{j}^{i+1} + Q_{j-1}^{i+1}$$

► We have $Q_1^0 = kP$

► In general denote the recombination function as $Q_{j/2}^{i} = f(Q_{j}^{i+1}, Q_{j-1}^{i+1})$

•The recombination steps can be represented as a tree:



•Putting this together, the algorithm for parallel scalar multiplication is:

```
Input: P, k = d_{2^n}^n || d_{2^n-1}^n || ... || d_1^n

Output: Q = kP

Q = O

for i=1 to 2<sup>n</sup>, in parallel

Q_i^n = d_n^j P

For i=n-1 to 0

For j=i+1 to 1, in parallel

Q_{j/2}^i = f(Q_j^{i+1}, Q_{j-1}^{i+1})

Return Q_0^1
```

•Hypercube and tree topologies naturally suited

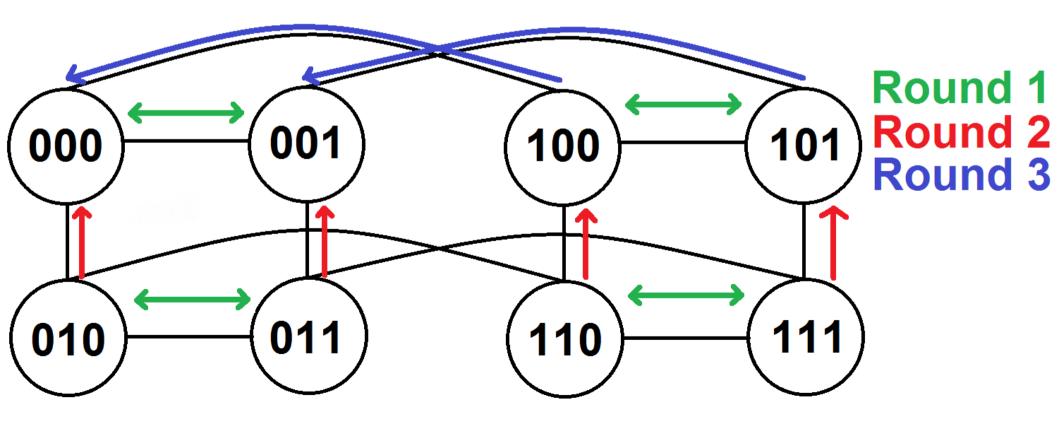
- Tree suitable for pipelining
- Hypercube could interweave multiple multiplications together

•A linear structure can also be used, but has worse running time than a hypercube or tree

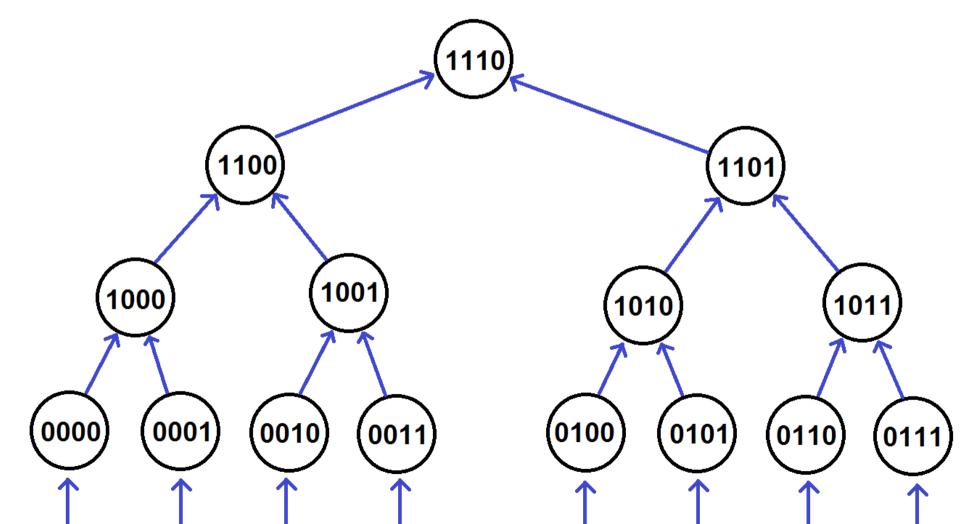
Better asymptotic throughput than a tree

•Higher throughput with no speedup can also be achieved by a simple division of processors, with results distributed across processors

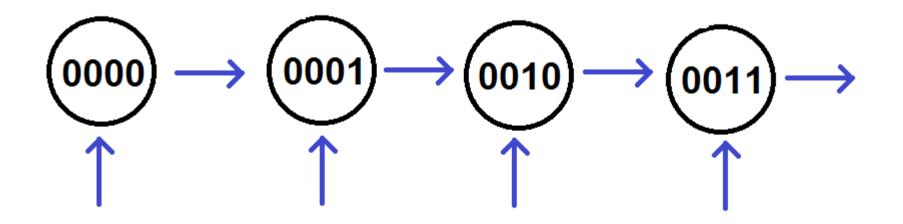
•Messages exchanged in a hypercube with 2 interweaved multiplications and 8 processors



•Messages exchanged while pipelining multiplications in a tree



•Messages exchanged while pipelining multiplications in a linear array



Asymptotic Running Time -Sequential

•In terms of point additions (A), point doublings (D), field size (m), and processors (p)

The tau operator is asymptotically more efficient than other point operations

•For a prime curve, **m** point doublings and on average **m/(1 + w)** point additions are required for a window size of w with 2^{w-2} precomputation work

•Asymptotic running time is thus:

- ►General: O(mD+mA)
- ►Koblitz: O(mA)

Asymptotic Running Time -Hypercube & Tree

- •First round computes multiplication of size m/p sequentially, requiring O(m/p D + m/p A) time
- •The i-th (of log p total) recombination round requires 2^{i} m/p point doublings and one addition
- •Theoretical optimal speedup using m/4 processors
- •Asymptotic parallel running time is thus:
 - ► General: $O(mD + (m/p + \log p)A)$ when $2^n < m/4$ $O(mD + (\log m)A)$ when $2^n \ge m/4$
 - ► Koblitz: $O((m/p + \log p)A)$ when $2^n < m/4$ $O((\log m)A)$ when $2^n \ge m/4$

Asymptotic Running Time -Linear

- •Each processor computes in parallel a sequential multiplication of size m/p, requiring O(m/p) time
- •Recombination requires O(m/p) point doublings per processor, except the last one, and a single point addition
- •Asymptotic parallel running time is thus:
 - General: O(mD + (m/p + p)A)
 - ►Koblitz: O((m/p + p)A)

Asymptotic Throughput

•Throughput in a tree is determined by the maximum of the root's computation time and the leaves' computation time:

►General: O(1 / max(m/p (D + A), m D)

►Koblitz: O(1 / (m/p A))

- •Throughput in a linear array is determined by the computation time in a single node:
 - ►General: O(1 / (m/p D + m/p A))

►Koblitz: O(1 / (m/p A))

Practical Running Time & Throughput

- •Parallel overhead O(log p) time for a tree or hypercube and O(p) time for a linear array
 - Network delays (MPI)
 - Packing/unpacking overhead (MPI)
 - Synchronization delays (OpenMP)
- •Constant factors impact running time

Window sizes vary based on subscalar size, limiting speedup for regular multiplication

Practical Running Time & Throughput

•Sequential portion of multiplication – point doubling or tau operator and scalar conversion

Large sequential portion due to point doubling cost for general curves limits speedup

More efficient tau operator reduces sequential portion, but sequential portion becomes more significant with many processors

Sequential portion more significant for regular multiplication, further limiting speedup

Experimental Parameters

- •10 standard NIST curves: P-192, P-224, P-256, P-384, P-521, K-163, K-233, K-283, K-409, K-571
- •Number of cores varied from 1-128
- •Input form of scalar NAF or binary
- •Number of simultaneous multiplications varied from 1-16 (hypercube)
- •Multiplication type Montgomery or regular
- •Logical topologies Hypercube, Tree, Linear
- •OpenSSL used to handle basic point operations
- •GMP/MPFR to handle large rationals/floats

Experimental Setup

•16 core machines utilized for all tests at UB CCR:

- ►Intel E5-2660 Xeon (dual 8 core)
- Infiniband Network (when using >16 cores)
- •MPI Thread Safety for Hybrid Approach
 - Tree/hypercube: MPI_THREAD_SERIALIZED
 - Linear: MPI_THREAD_MULTIPLE
- •Points and scalars generated at random
- •50,000 total multiplications performed for each experiment

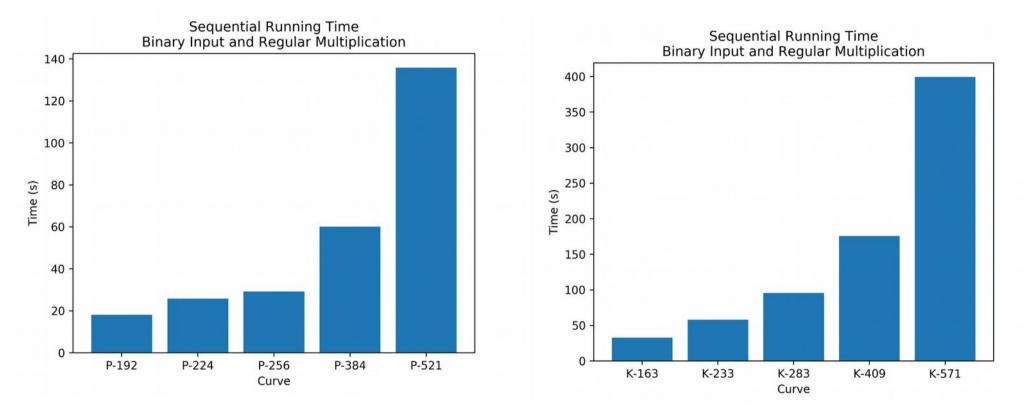
Experimental Setup

•Linear and tree running time is not measured directly, but estimated

Tree running time estimated by estimated by summing average running time at each tree level excluding the time spent waiting for other processors

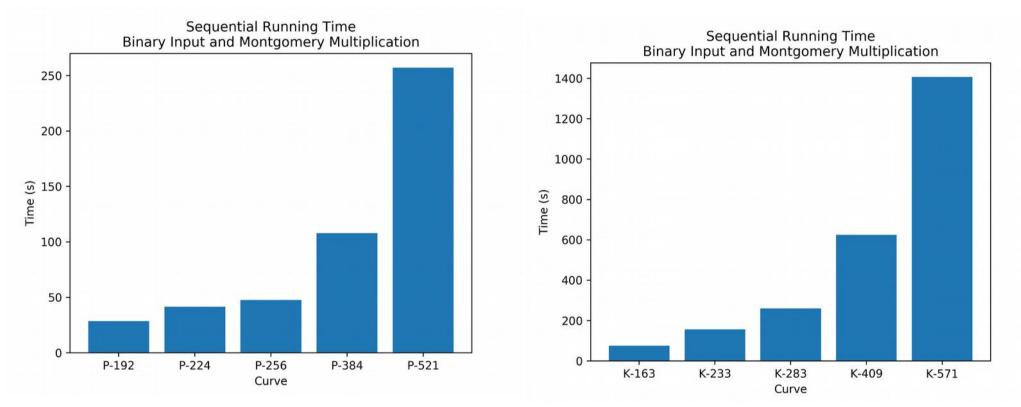
Linear running time estimated by summing the the time spent in each node sequentially plus the time spent in parallel

Sequential Running Time



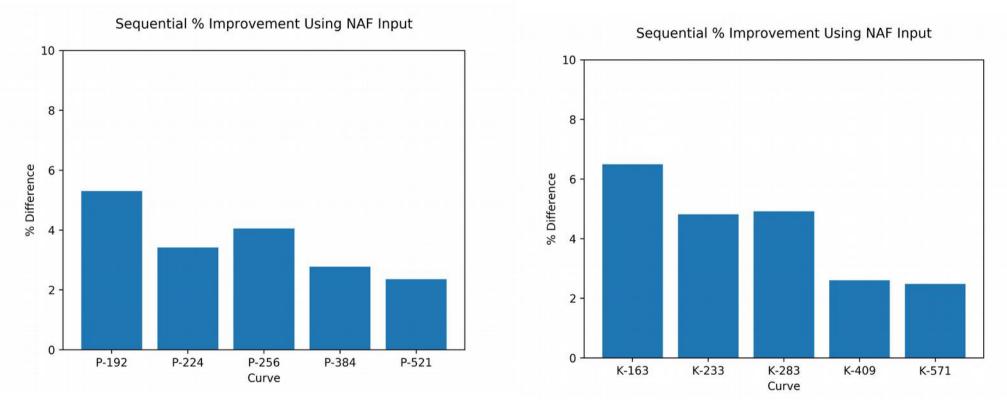
•Koblitz curves (right) exhibit slower running times due to less support in OpenSSL and binary curves in general being better suited for hardware implementations

Sequential Running Time



- •Montgomery methods up to 3.5 slower than regular multiplications (previous)
- Performance hit worse for Koblitz curves

Sequential Running Time

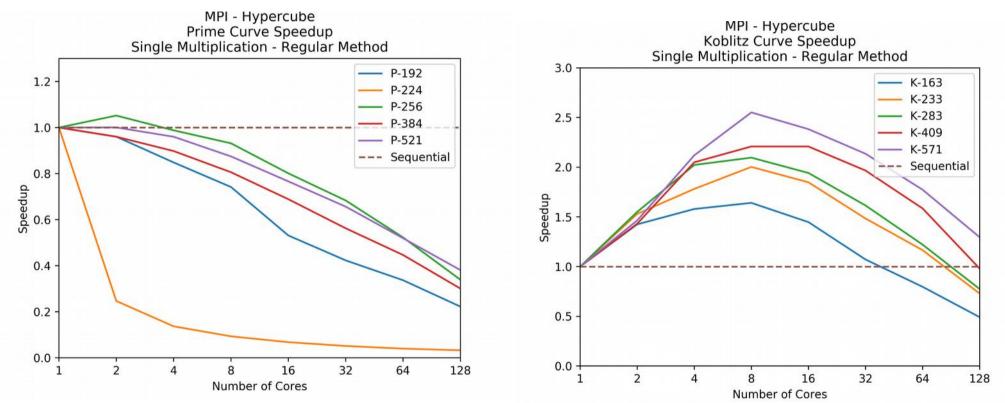


•Small improvement using NAF input

·Going forward, only binary input is presented

Results for NAF input show slight improvement

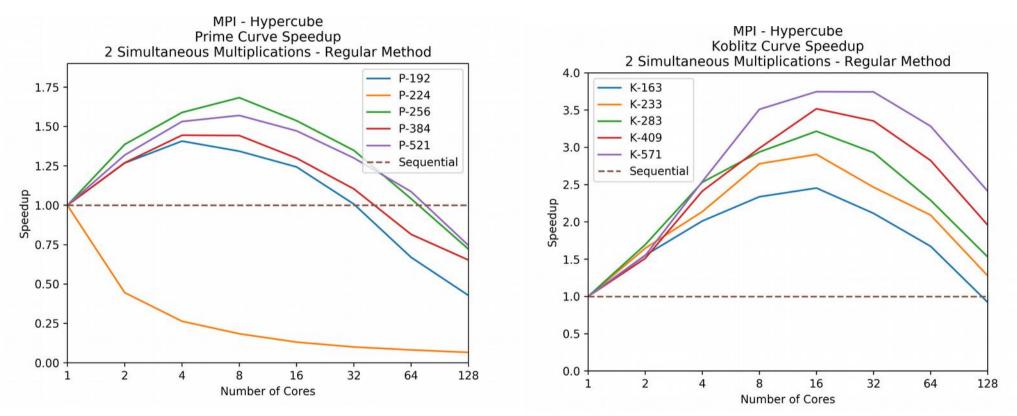
Hypercube Speedup



 Large parallel overhead limits speedup for prime curves in particular

Worse than sequential except P-256 using 2 cores

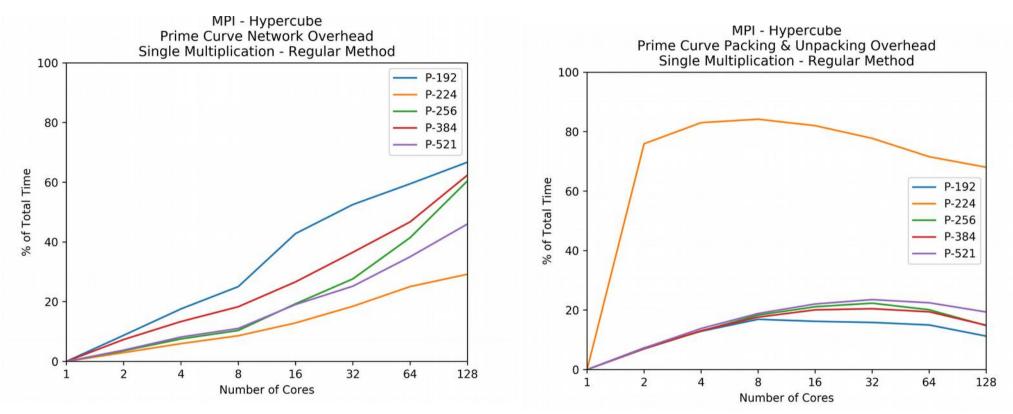
Hypercube Speedup



Interweaving worse than dividing processors

Same holds for other configurations – further graphs on simultaneous multiplications omitted

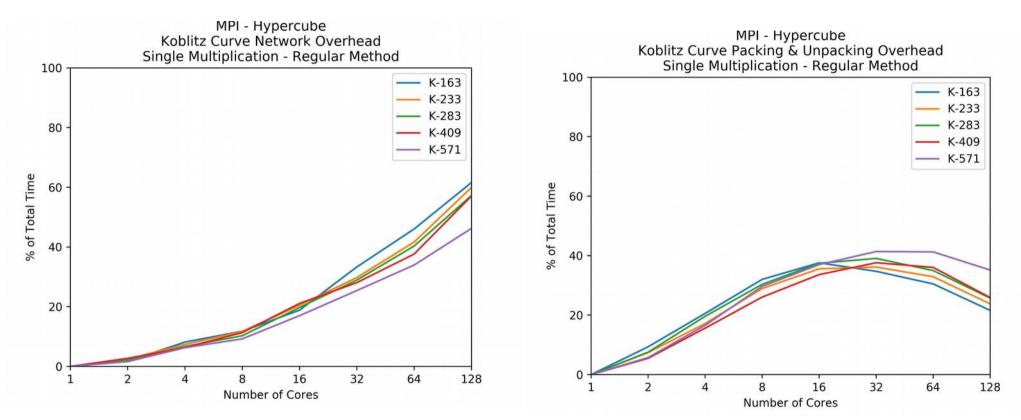
Hypercube Overhead



•Overhead grows with number of cores

•OpenSSL optimizations for P-224 at expense of packing/unpacking time explain its results

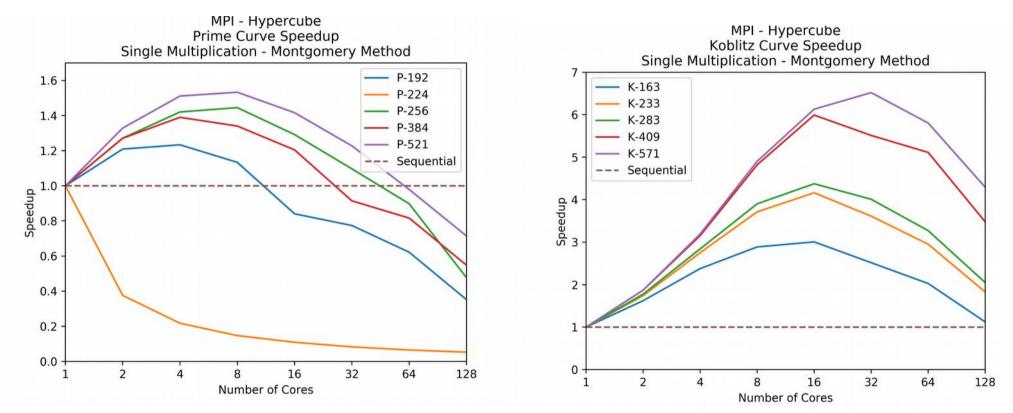
Hypercube Overhead



•More time spent on packing/unpacking overhead for Koblitz curves

Generally less networking delays for Koblitz curves

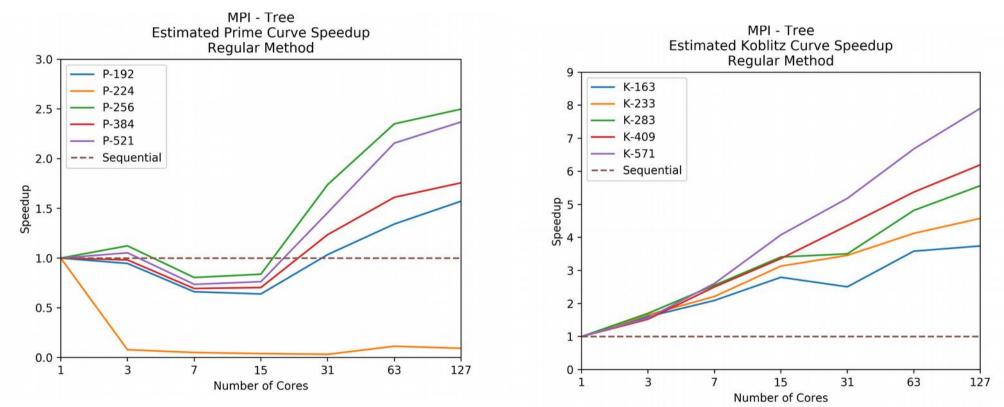
Hypercube Speedup



•Better speedup using a Montgomery method

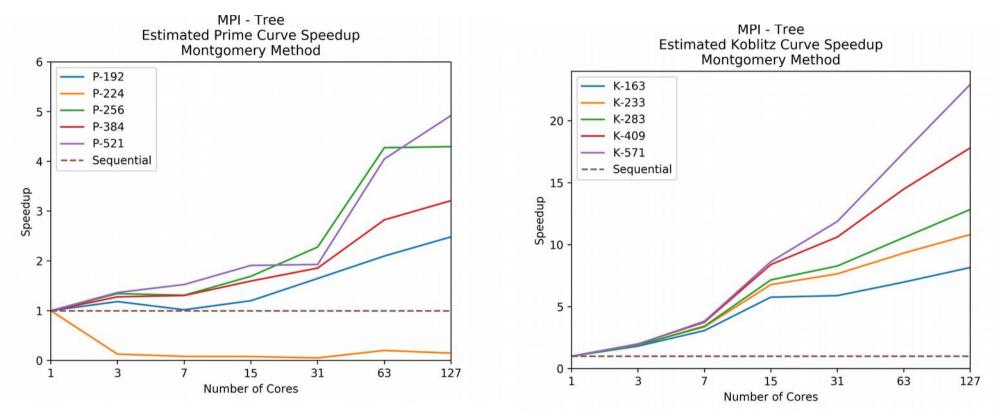
 Prime curves show limited speedup due to larger sequential portion

Tree Speedup



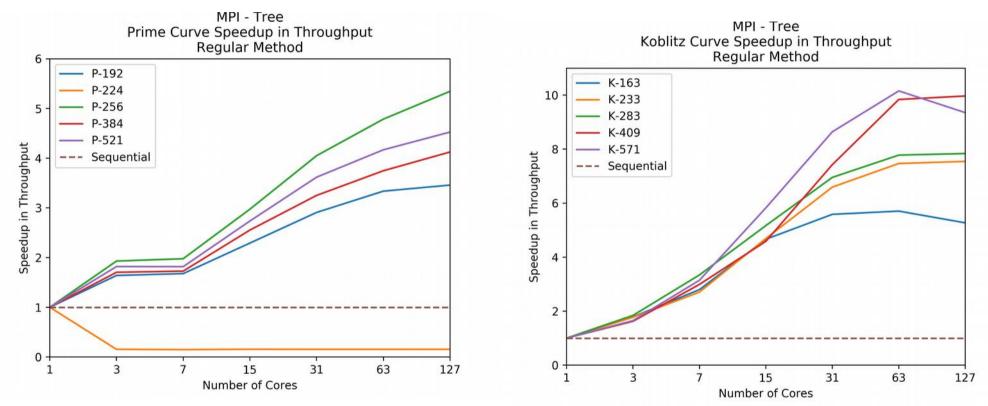
- •Better speedup than equivalent hypercube as communications spread out over more time
- •Overhead/constant factors outweigh parallel benefits for prime curves with <15 processors

Tree Speedup



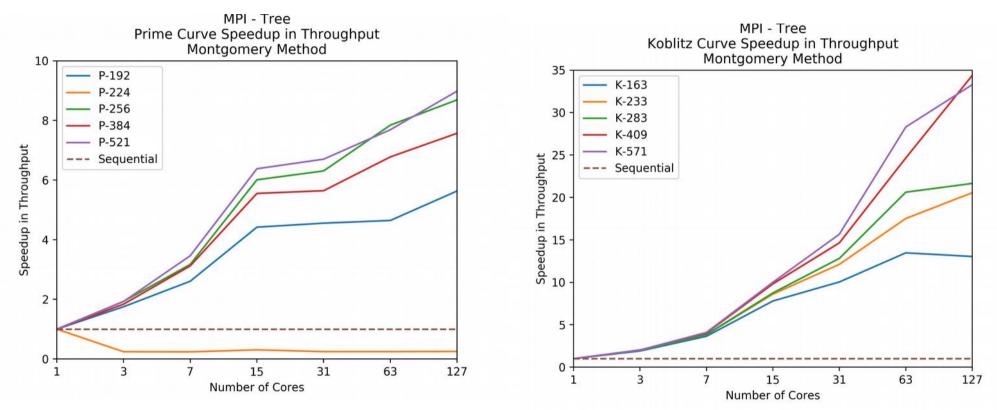
•Better speedup using Montgomery method

Tree Throughput



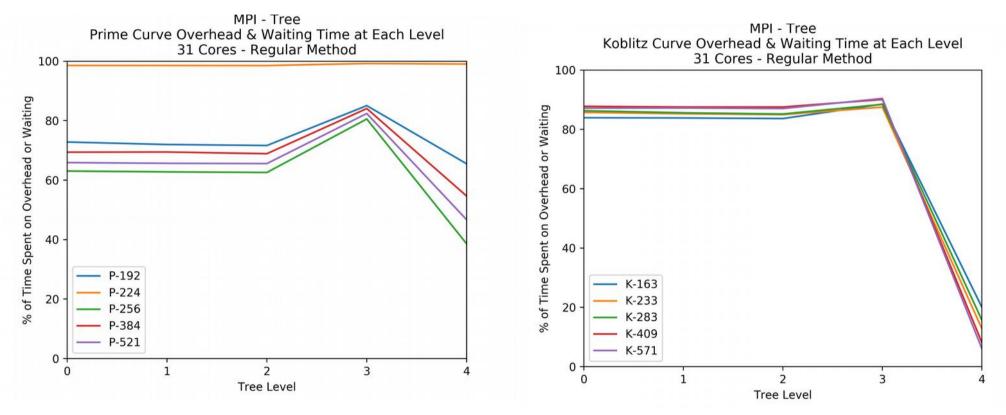
- Throughput continues to improve (except P-224) as number of cores increased
- •Better throughput by using processors sequentially, but worse speedup in some cases

Tree Throughput



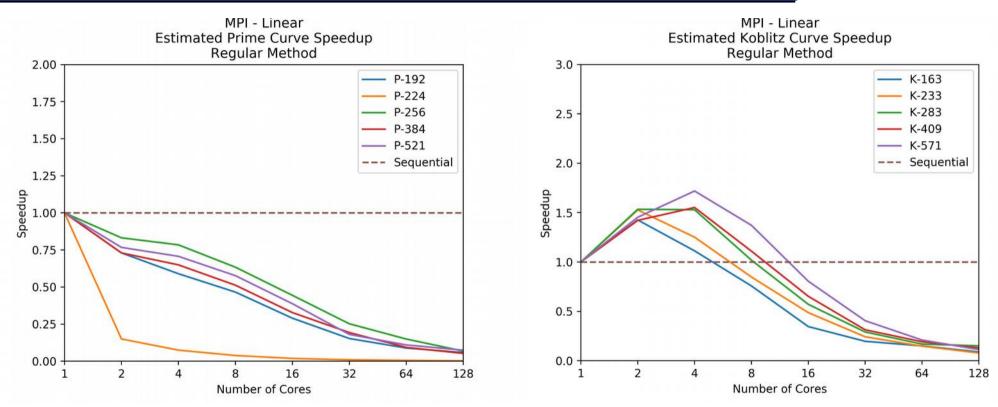
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Time Spent Waiting or on Parallel Overhead in Tree



- Large amount of idle time, waiting for other processors at non-leave levels
- •Similar results for other configurations

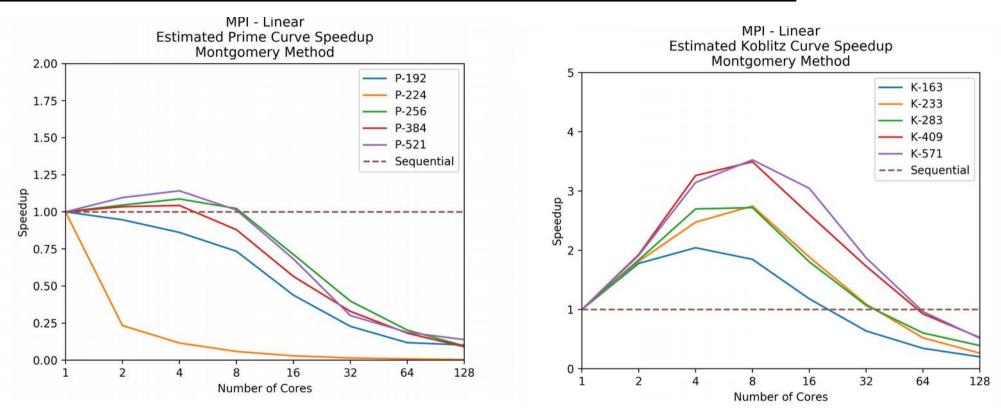
Linear Speedup



Strictly worse than sequential for prime curves

•For Koblitz curves, 2 cores give speedup comparable to 2 core hypercube or 3 core tree and worse otherwise

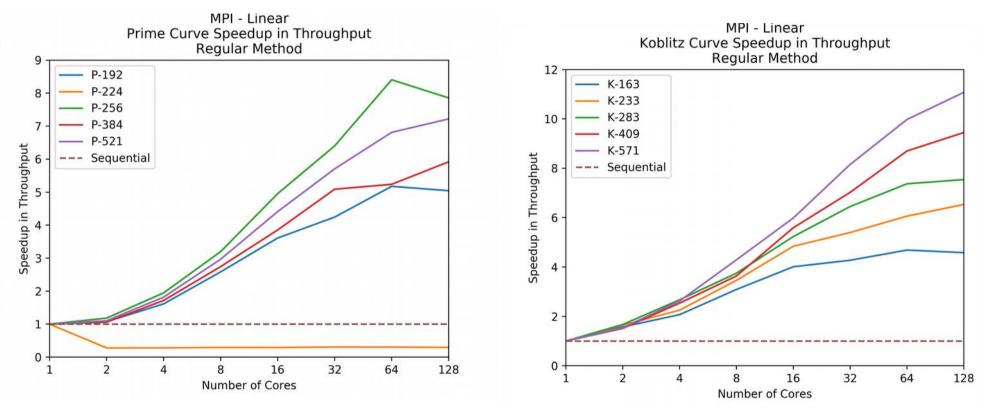
Linear Speedup



•Montgomery method shows marginal speedup for prime curves, worse than hypercube or tree

•Better speedup for some Koblitz curves for 2-4 cores compared to 2-4 core hypercube or 3-7 core tree

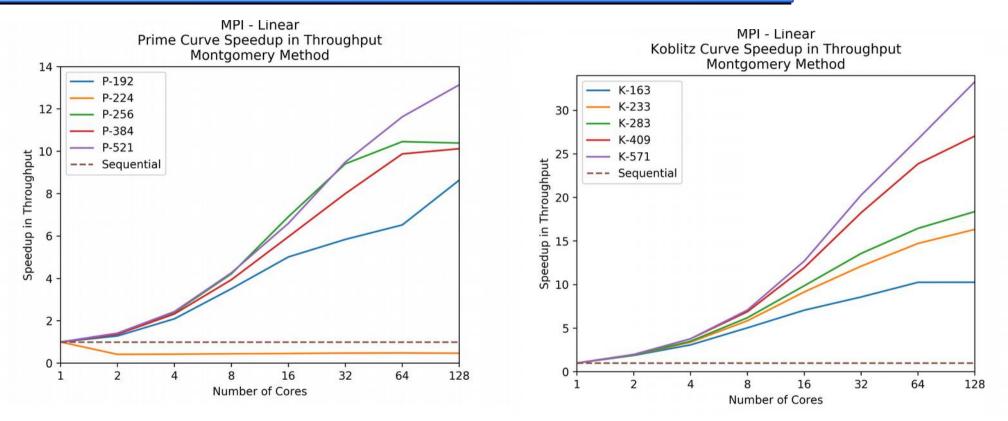
Linear Throughput



•Throughput is generally a bit better than a tree

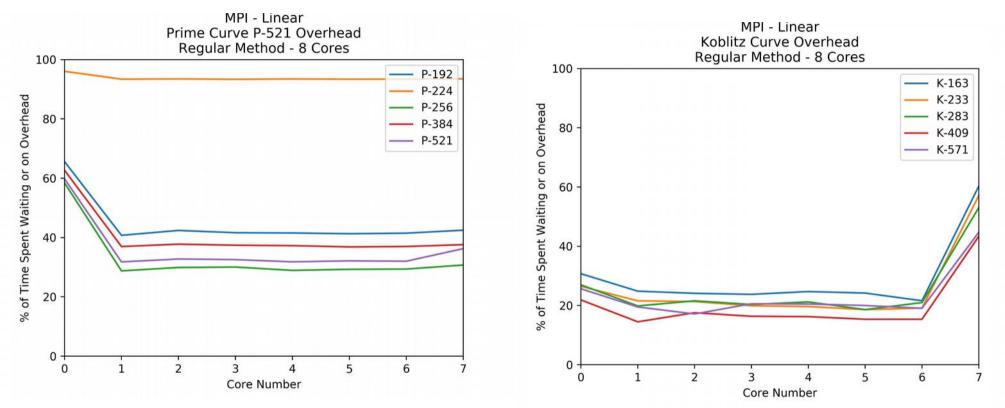
Strictly better to distribute multiplications sequentially on prime curves using since no speedup advantages and worse throughput

Linear Throughput



Slightly better throughput than a tree when using few cores

Linear Overhead



Generally linear overhead takes up less overall time

•Similar results for other configurations

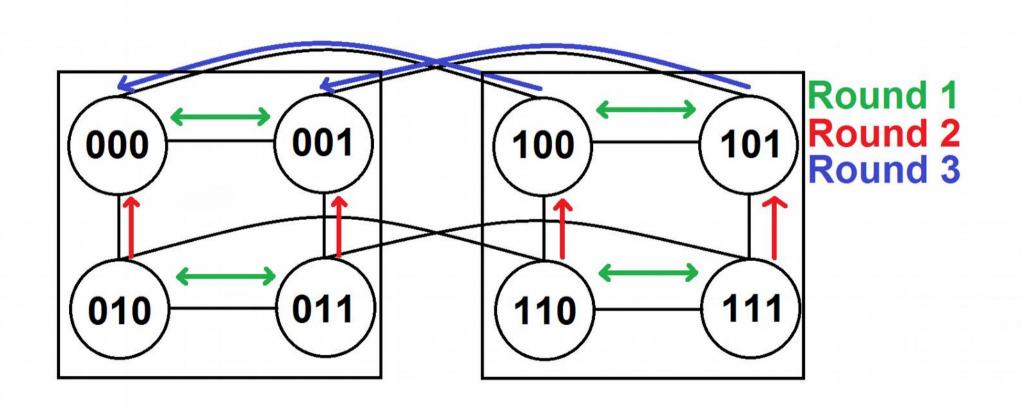
MPI Conclusions

- •Packing/unpacking time for some curves and network delays limit achievable speedup and throughput
- •Simultaneous communication can cause congestion limiting speedup, as seen with a tree achieving better speedup than an equivalent hypercube
- •Trees generally offer good balance between speedup and throughput
- •Linear array never good for prime curves, and better than a tree for Koblitz curves with a small number of cores available

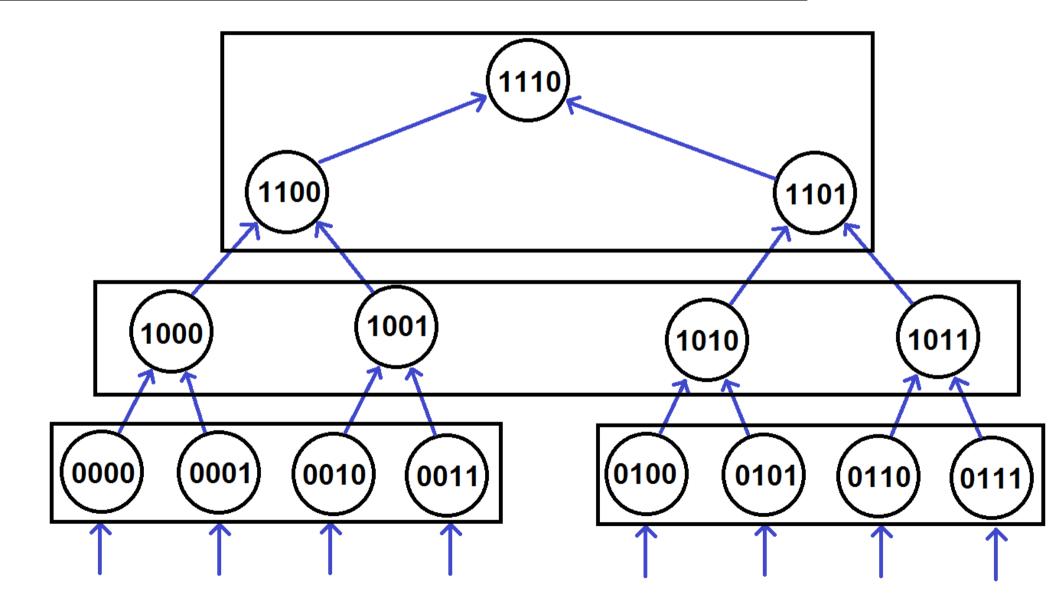
Challenges Moving to a Hybrid Approach

- Explicit synchronization required in OpenMP
- •Results from MPI indicate limiting MPI calls could be beneficial
 - Where possible, MPI calls are merged, but this requires additional synchronization
- •Where to use OpenMP vs MPI?
 - Based on rounds in hypercube topology
 - Based on level in tree topology
 - Based on neighbors in linear topology

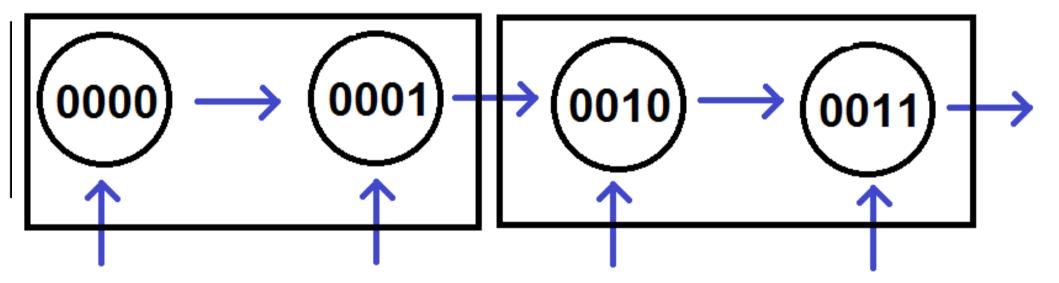
Hybrid Hypercube with 2 MPI nodes and 4 threads



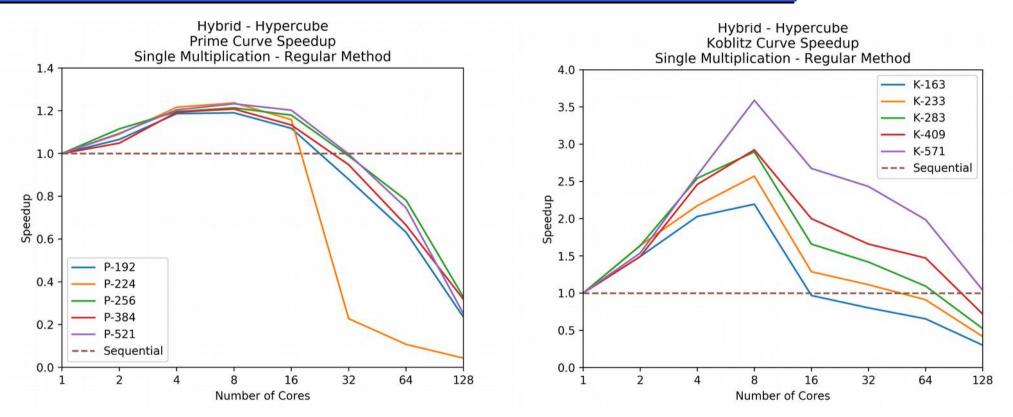
Hybrid Tree with 4 MPI nodes and 4 threads



Hybrid Linear with 2 MPI nodes and 2 threads



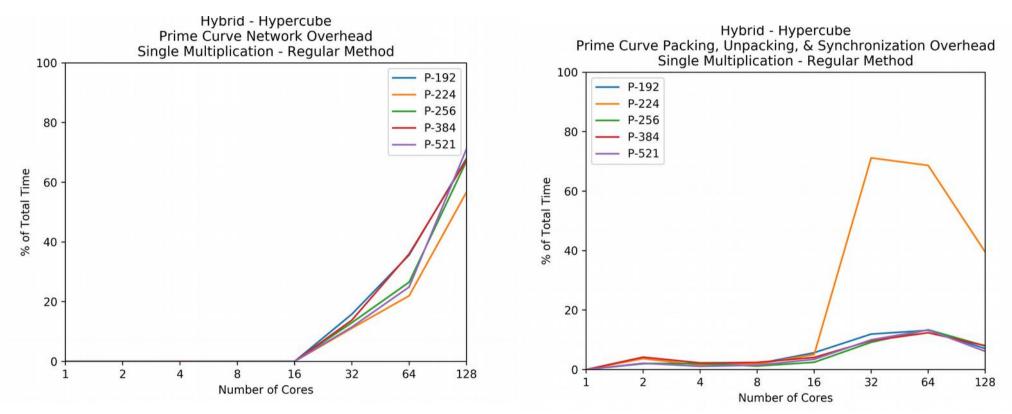
Hypercube Speedup



•Better speedup than MPI until 16 cores for prime curves and 8-16 cores for Koblitz curves

Performance impact for >8 cores may be due to frequent cache misses between processors

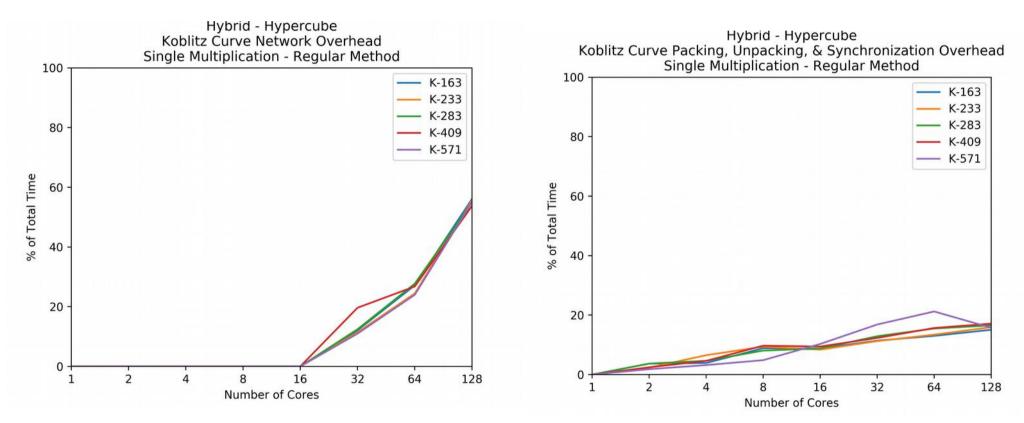
Hypercube Overhead



OpenMP has less overhead compared to MPI

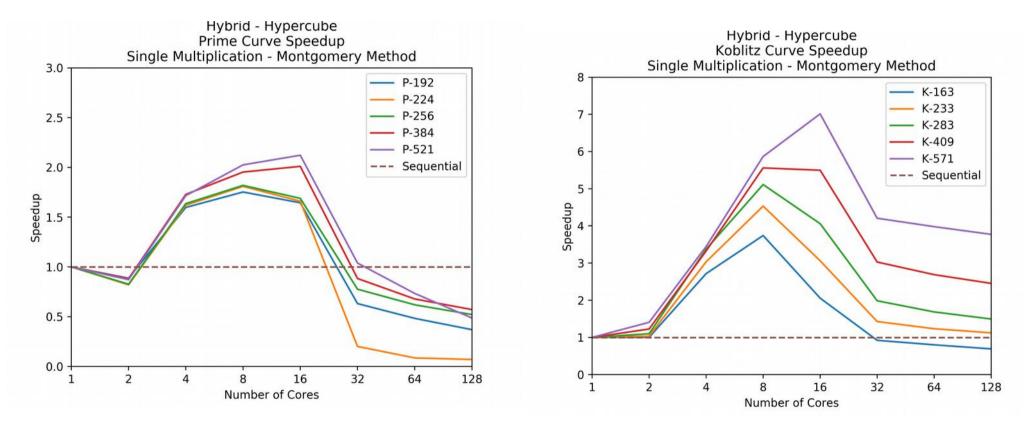
•Network delays with hybrid approach (>16 cores) quickly become significant

Hypercube Overhead



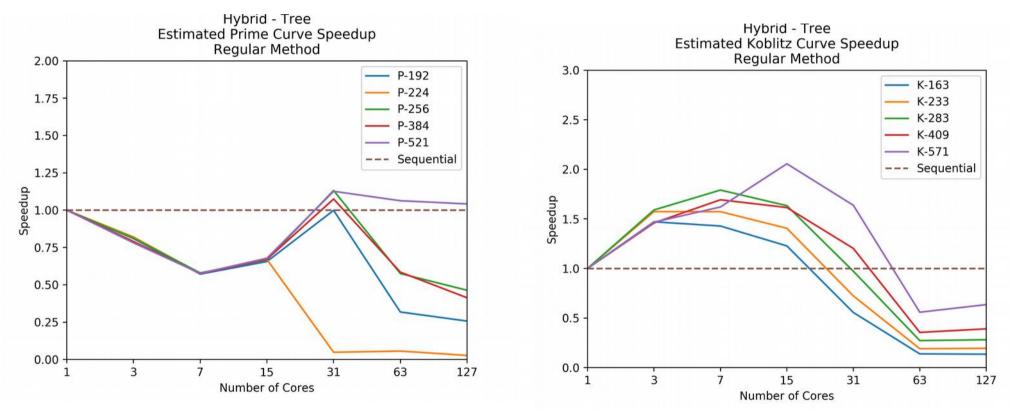
 Montgomery method shows less networking overhead, and more time spent on other overhead

Hypercube Speedup



•Montgomery methods offer better speedup up to 8-16 cores with an initial performance hit at 2 cores compared to MPI

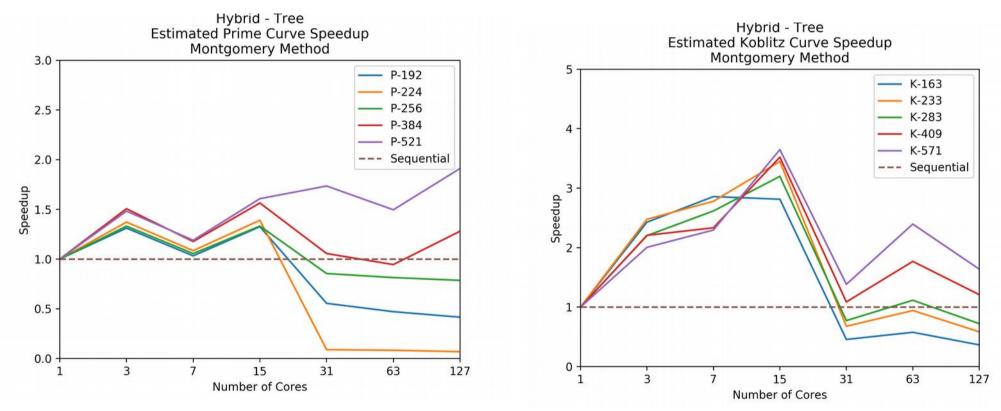
Tree Speedup



•Tree performs worse than in MPI

Synchronization costs for a tree greater than speedup attainable from the parallel algorithm

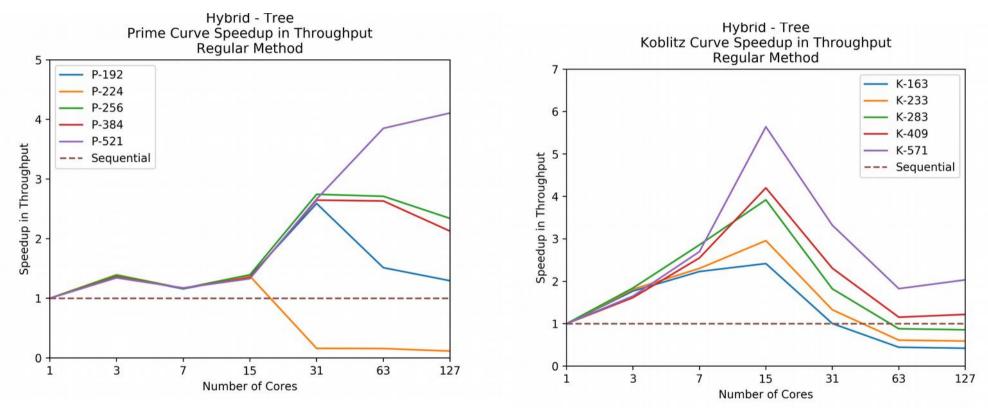
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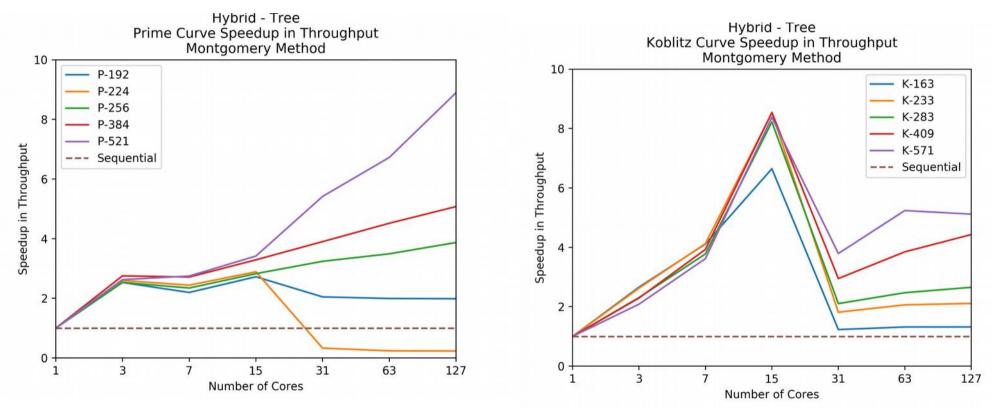
Tree Throughput



•Throughput for some curves comparable to throughput in MPI up to 15 cores

 Synchronization delays with >15 cores limits throughput

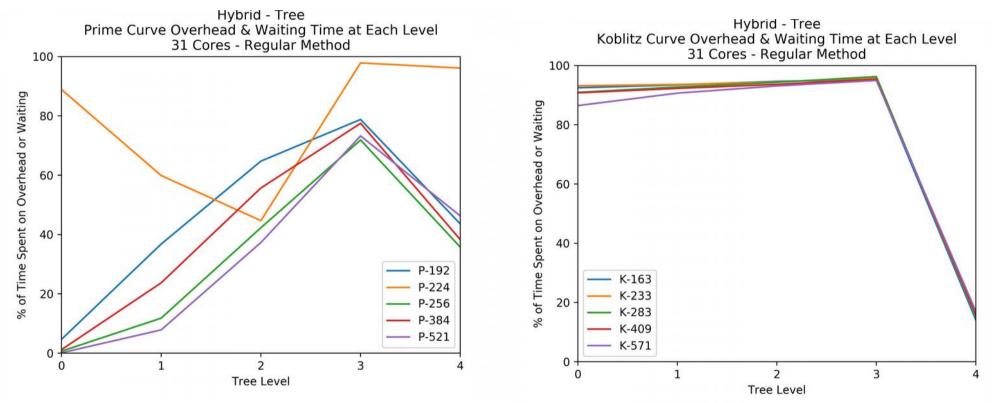
Tree Throughput



•Throughput for some Koblitz curves comparable to throughput in MPI up to 15 cores

 Synchronization delays with >15 cores limits throughput

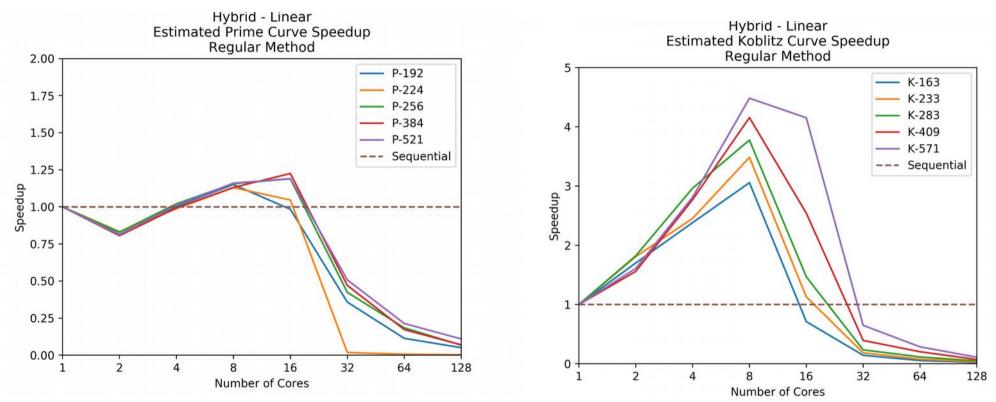
Time Spent Waiting or on Parallel Overhead in Tree



•Significant overhead costs and idle time (Koblitz curves)

 Additional costs incurred from setting locks used for synchronization

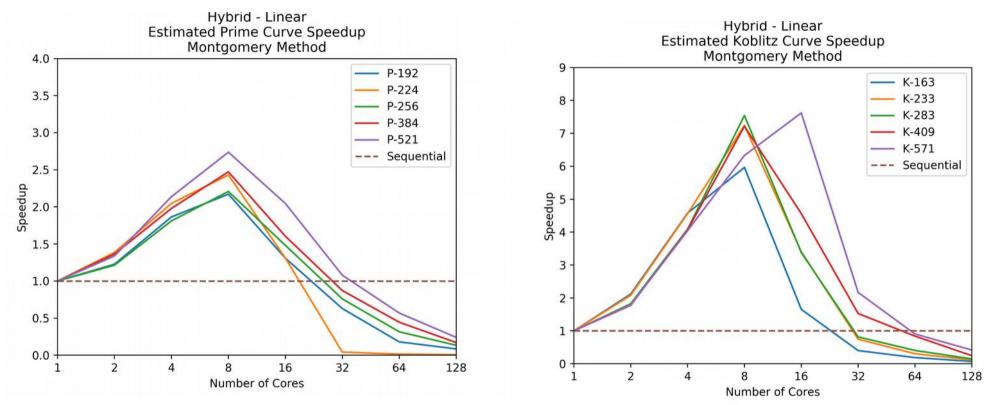
Linear Speedup



•Better speedup than in MPI with <16 cores for prime curves and 8 cores for Koblitz curves

For prime curves, parallel overhead overwhelms algorithm's speedup when using 2-4 cores

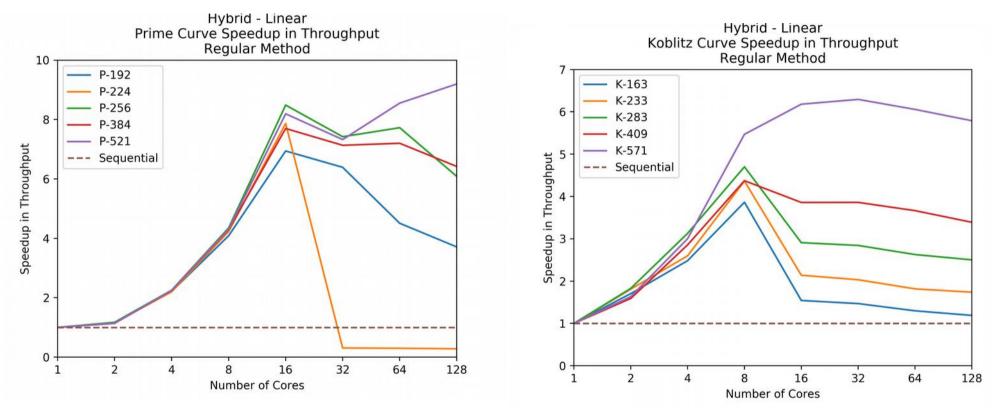
Linear Speedup



Surprisingly better speedup than a hypercube

- Less synchronization costs
- Performance hit at >8 cores

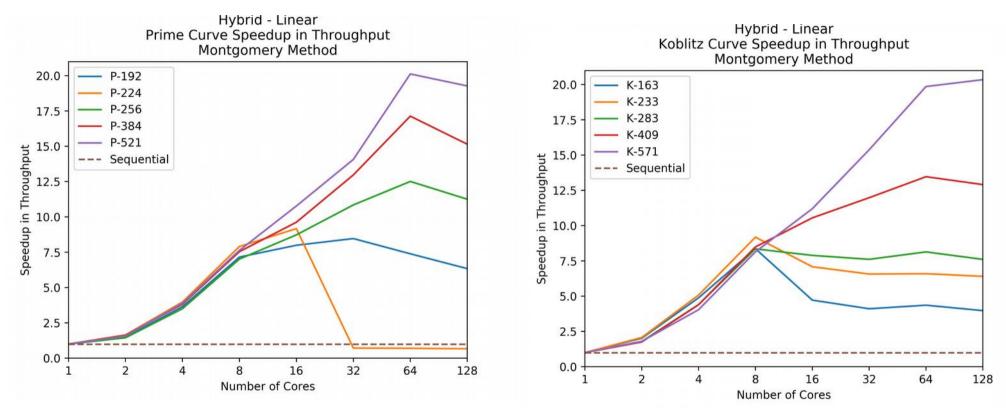
Linear Throughput



•Generally better throughput than when using MPI with linear array

Performance hit when hybrid approach is used and when two processors per compute node used

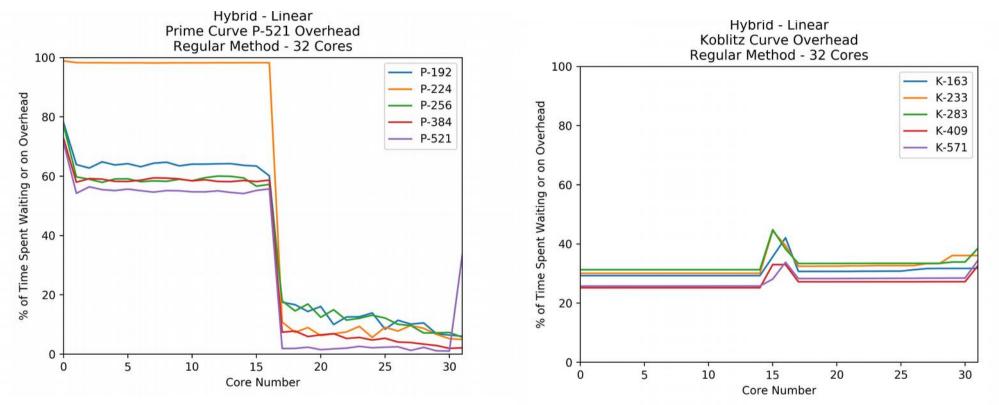
Linear Throughput



•Better throughput when using <8-16 cores than in MPI

Performance hit when hybrid approach is used and when two processors per compute node used

Linear Overhead



 Large overhead when utilizing multiple MPI nodes for prime curves corresponding to network delays

•Koblitz have nearly constant overhead for all cores with spikes near MPI node boundaries

Hybrid Conclusions

- •Synchronization delays can be worse than networking delays in MPI in some cases
- •Observed performance moving to 16 cores significantly impacted the hybrid approach
 - Frequent cache misses using multiple processors may be the cause for these results
- •Linear array showed better speedup than other structures, but worse throughput than in MPI
- Less overhead compared to other structures
 Merging MPI calls may not have been beneficial

Overall Conclusions

•Best logical structure depends on number of cores available, desired throughput, desired speedup, and curve type

Koblitz curves better suited for parallelization

Splitting cores sequentially best for maximizing throughput

MPI tree gives generally good balance between speedup and throughput, for many cores

 OpenMP linear array gives generally good balance between speedup and throughput for few cores

Future & Related Work

- •Large amount of time in a tree is spent waiting for other processors for non-leaves, and it may be possible to merge some non-leave nodes
- •Combining topologies may yield better throughput results in some cases
- •Parallelism at the point or field level is also possible using a fixed number of processors
- •Multiple multiplications on the same point can use globally precomputed values for better performance
 - Key generation

Future & Related Work

•Better results can likely be achieved if suspected frequent cache misses due to dual-processor compute nodes are accounted for

One method to account for this is to use 2 MPI nodes per server (1 per processor), with 8 threads used per MPI node so MPI takes care of it

•Not merging MPI calls may be better suited for hypercubes and trees in the hybrid approach

References

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