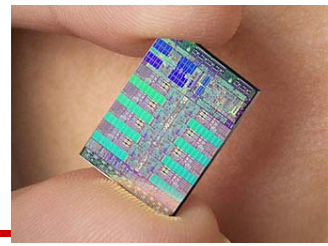


Experiments with the Cell for Linear Algebra Operations

Jack Dongarra

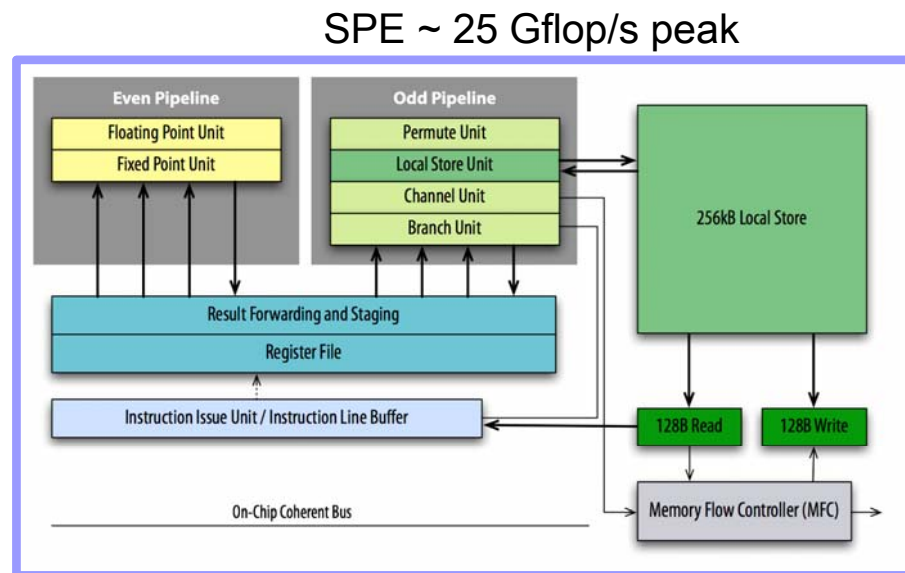
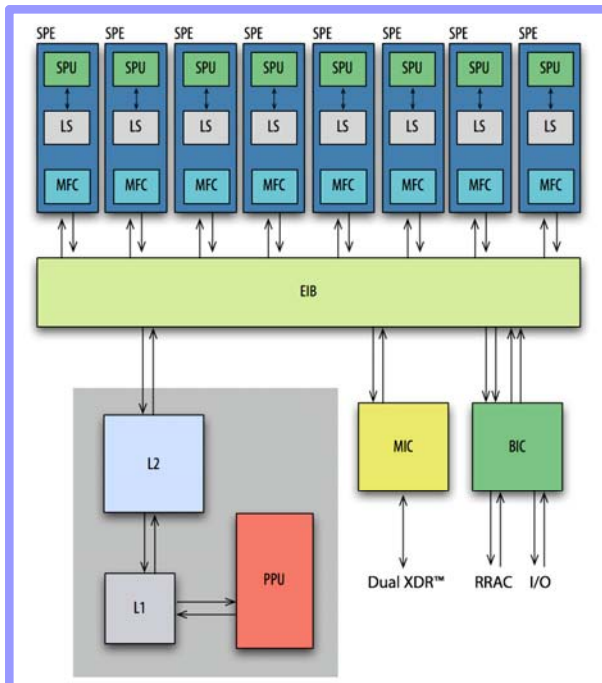
INNOVATIVE COMPUTING LABORATORY

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Exploiting Mixed Precision

- Current Version of the Cell has > a factor of 10 between single precision and double precision performance (204 GFlop/s to 14 GFlop/s)!
 - Next version this will narrow to a factor of 2 (as in most common processors today)
- We became interested in looking for ways to exploit the speed of SP but still retain the accuracy of DP.



Idea Goes Something Like This...

- Exploit 32 bit floating point as much as possible.
 - Especially for the bulk of the computation
- Correct or update the solution with selective use of 64 bit floating point to provide a refined results
- Intuitively:
 - Compute a 32 bit result,
 - Calculate a correction to 32 bit result using selected higher precision and,
 - Perform the update of the 32 bit results with the correction using high precision.

Mixed-Precision Iterative Refinement

- Iterative refinement for dense systems, $Ax = b$, can work this way.

$L U = \text{lu}(A)$	$O(n^3)$
$x = L \setminus (U \setminus b)$	$O(n^2)$
$r = b - Ax$	$O(n^2)$
WHILE $\ r \ $ not small enough	
$z = L \setminus (U \setminus r)$	$O(n^2)$
$x = x + z$	$O(n^1)$
$r = b - Ax$	$O(n^2)$
END	

- Wilkinson, Moler, Stewart, & Higham provide error bound for SP fl pt results when using DP fl pt.

Mixed-Precision Iterative Refinement

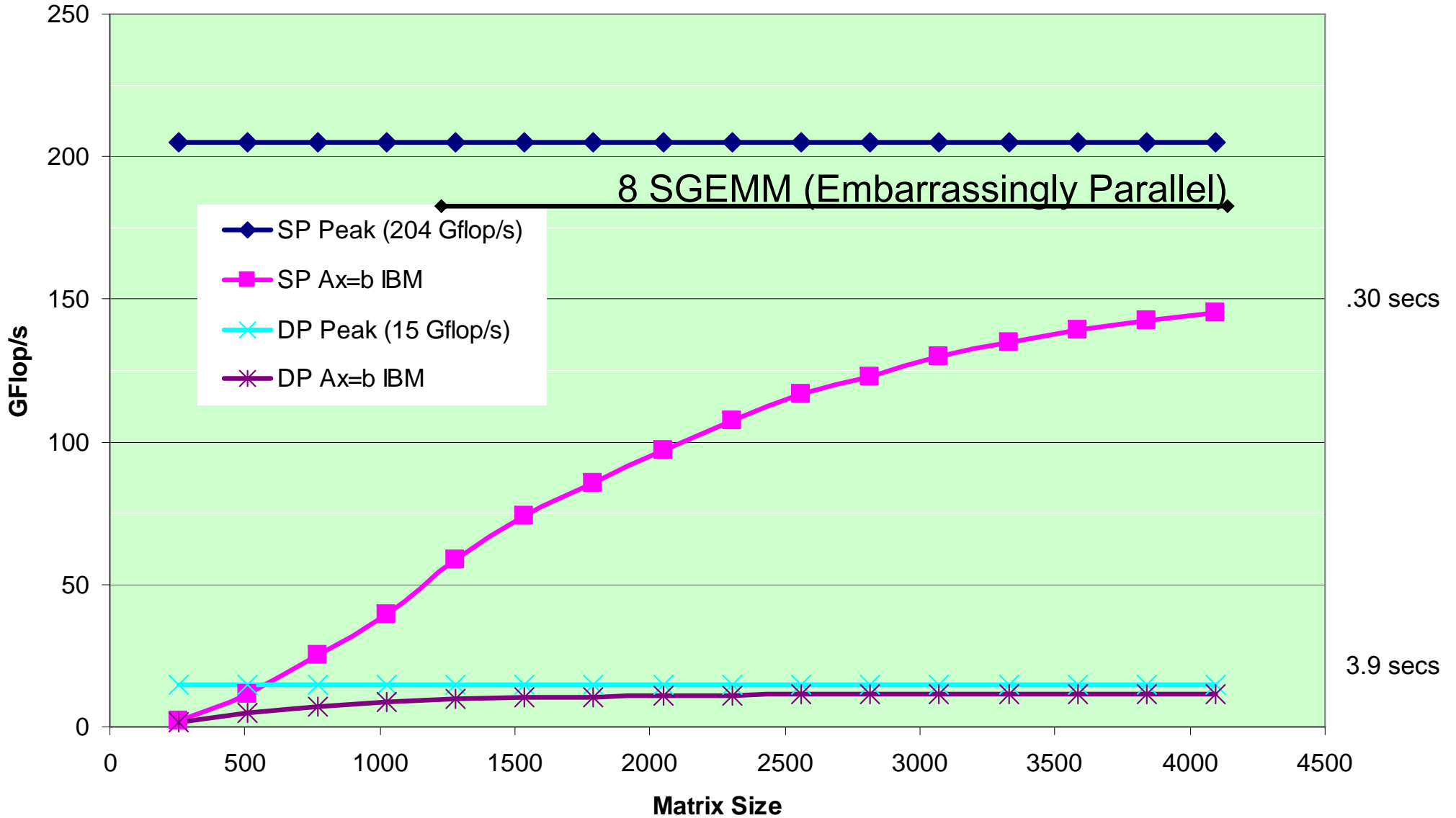
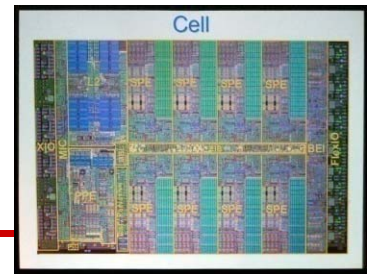
- Iterative refinement for dense systems, $Ax = b$, can work this way.

$L U = \text{lu}(A)$	SINGLE	$O(n^3)$
$x = L \setminus (U \setminus b)$	SINGLE	$O(n^2)$
$r = b - Ax$	DOUBLE	$O(n^2)$
WHILE $\ r \ $ not small enough		
$z = L \setminus (U \setminus r)$	SINGLE	$O(n^2)$
$x = x + z$	DOUBLE	$O(n^1)$
$r = b - Ax$	DOUBLE	$O(n^2)$
END		

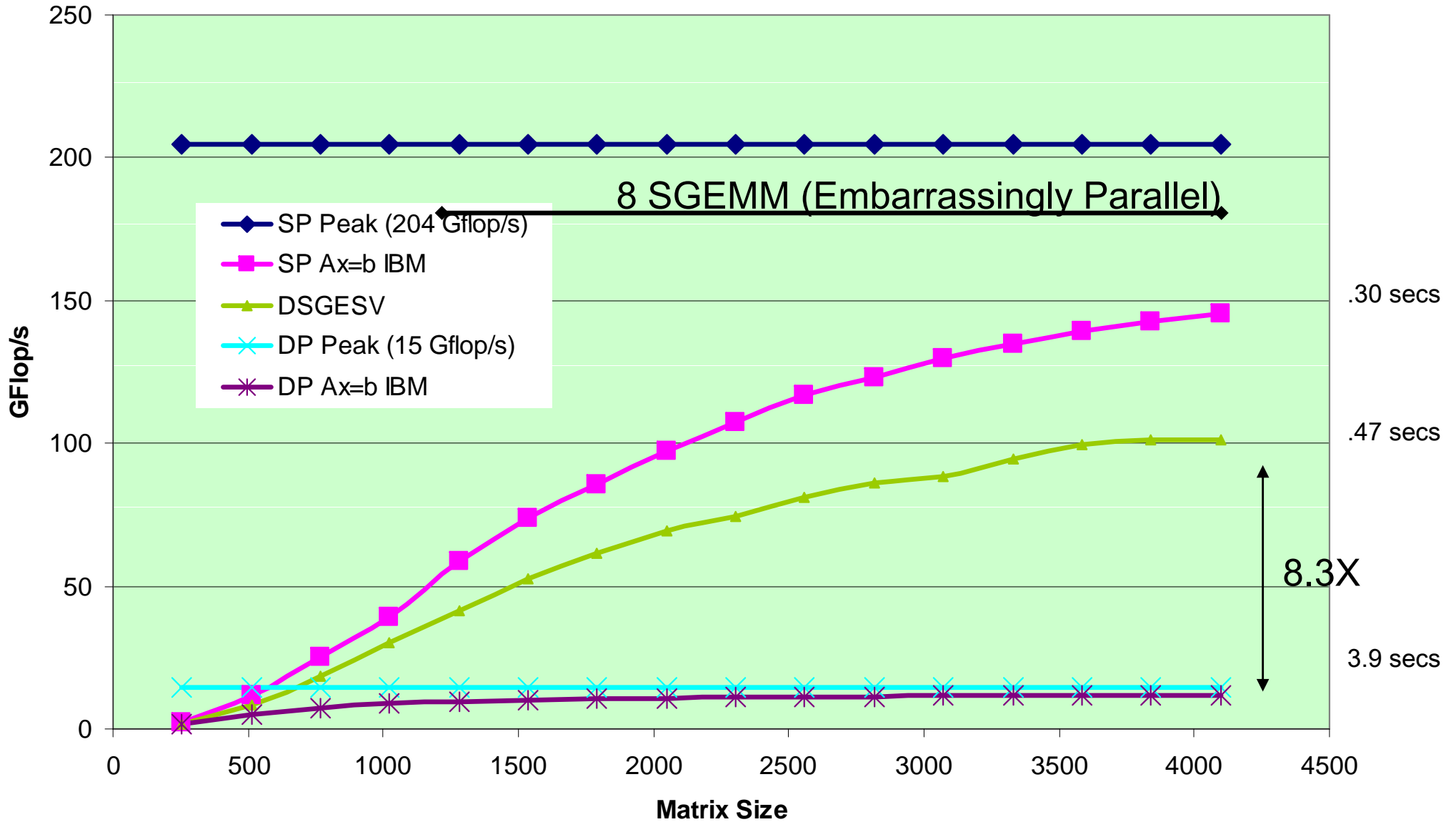
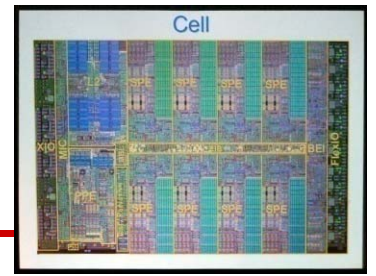
- Wilkinson, Moler, Stewart, & Higham provide error bound for SP fl pt results when using DP fl pt.
- It can be shown that using this approach we can compute the solution to 64-bit floating point precision.

- Requires extra storage, total is 1.5 times normal;
 - $O(n^3)$ work is done in **lower precision**
 - $O(n^2)$ work is done in **high precision**
- 5
- Problems if the matrix is ill-conditioned in sp; $O(10^8)$

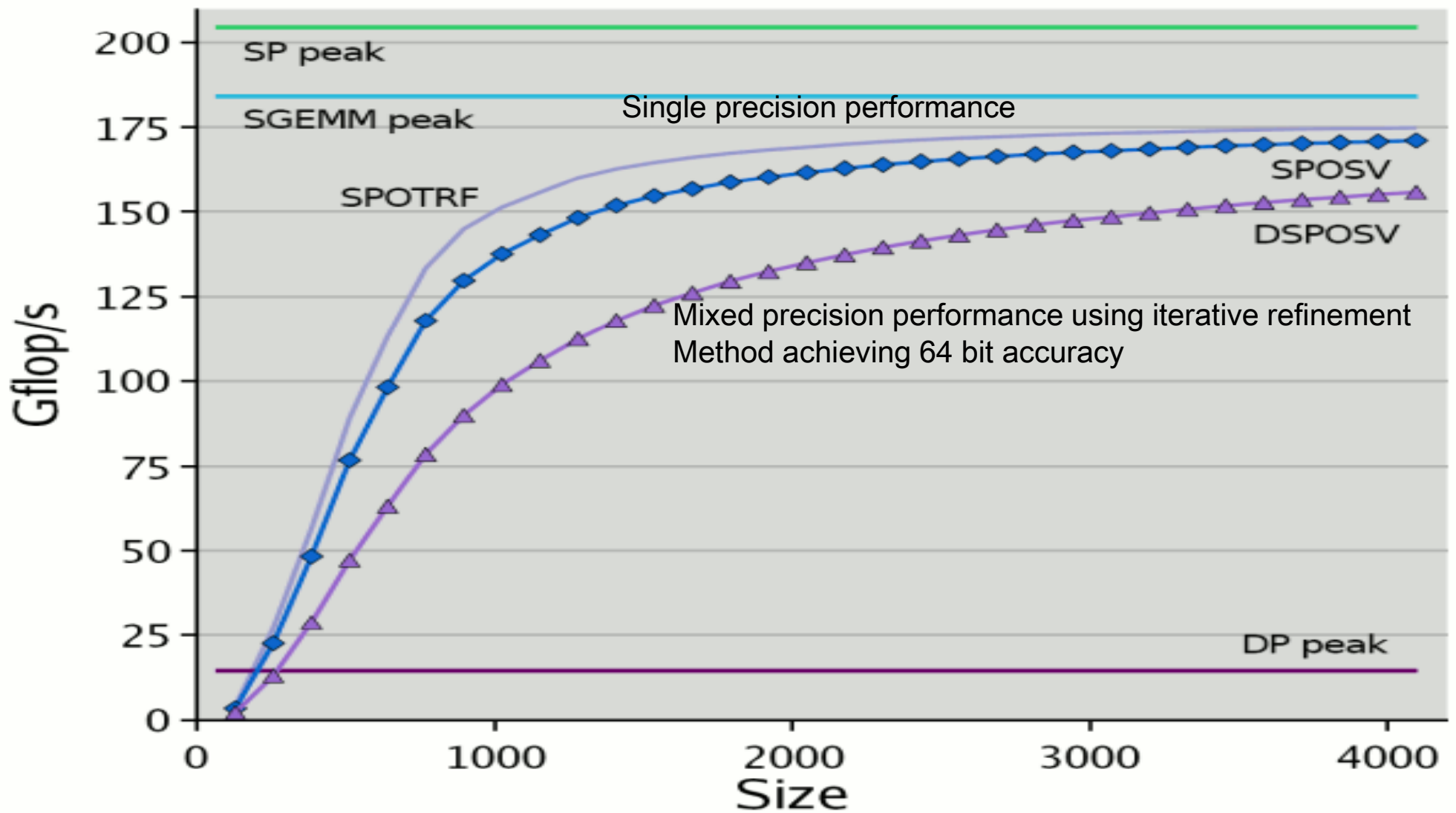
IBM Cell 3.2 GHz, $Ax = b$



IBM Cell 3.2 GHz, $Ax = b$

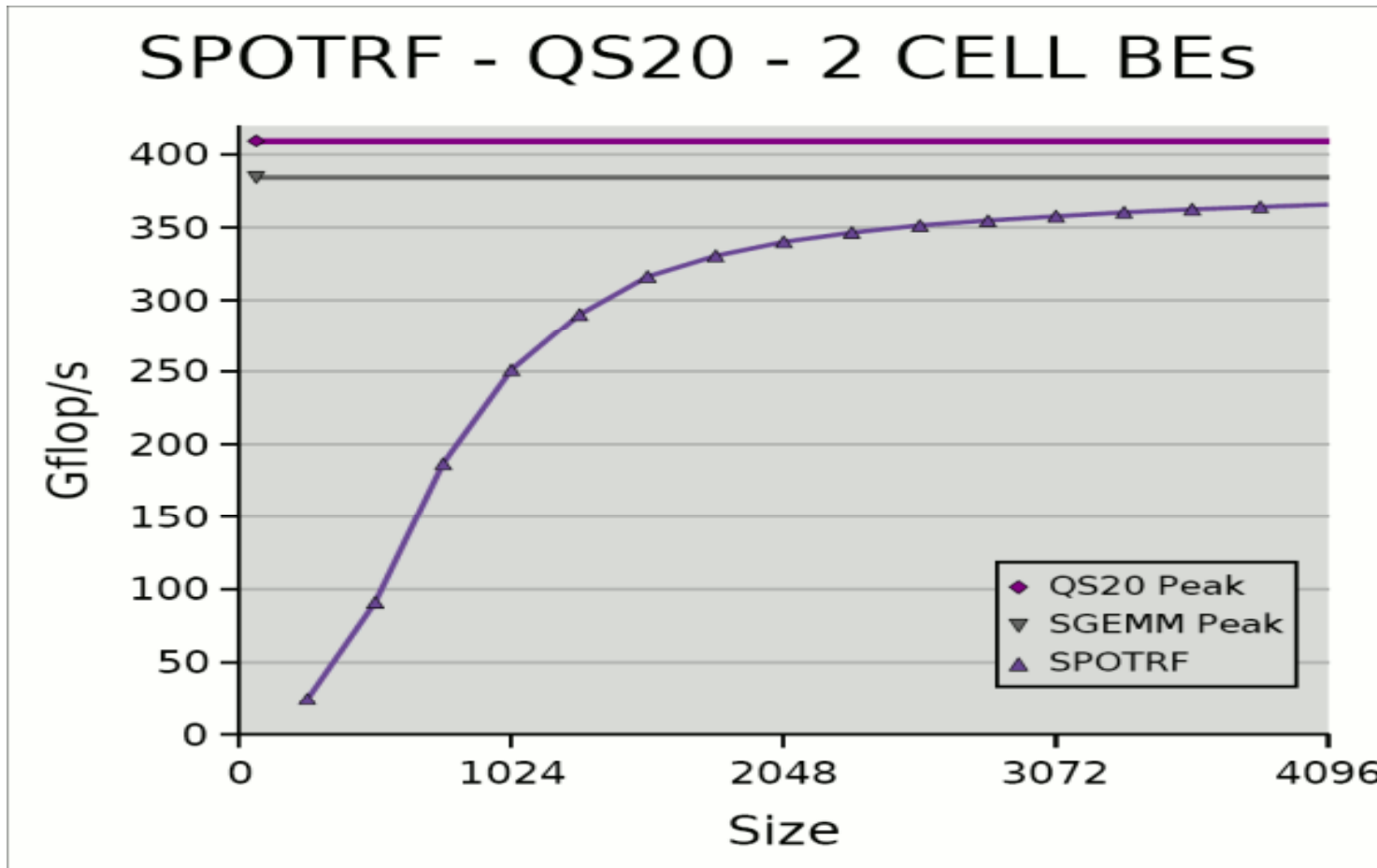


Cholesky on the Cell, $Ax=b$, $A=A^T$, $x^T Ax > 0$



For the SPE's standard C code and C language SIMD extensions (intrinsics)

Cholesky - Using 2 Cell Chips



Intriguing Potential

- Exploit lower precision as much as possible
 - Payoff in performance
 - Faster floating point
 - Less data to move
- Automatically switch between SP and DP to match the desired accuracy
 - Compute solution in SP and then a correction to the solution in DP
- Potential for GPU, FPGA, special purpose processors
 - What about 16 bit floating point?
 - Use as little you can get away with and improve the accuracy
- Applies to sparse direct and iterative linear systems and Eigenvalue, optimization problems, where Newton's method is used.

$$x_{i+1} - x_i = - \frac{f(x_i)}{f'(x_i)}$$

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

Correction = - A\b(b - Ax)