ThreadedDenseSparseMul.jl

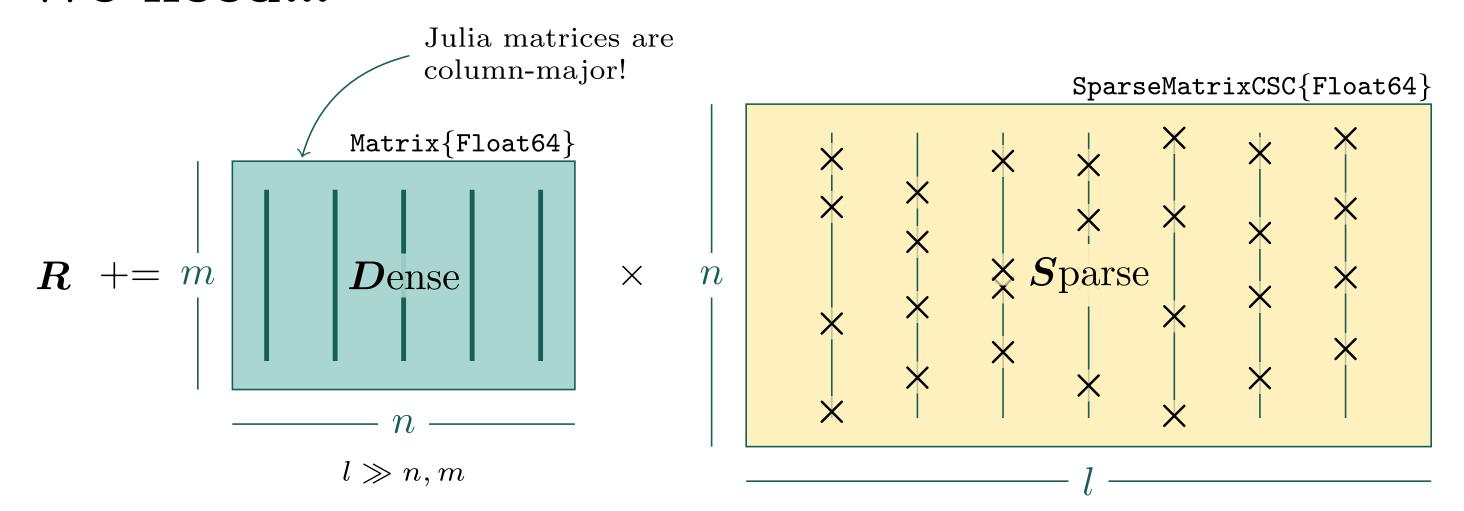
Outperforming MKLSparse in 6 LOC.

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We need...

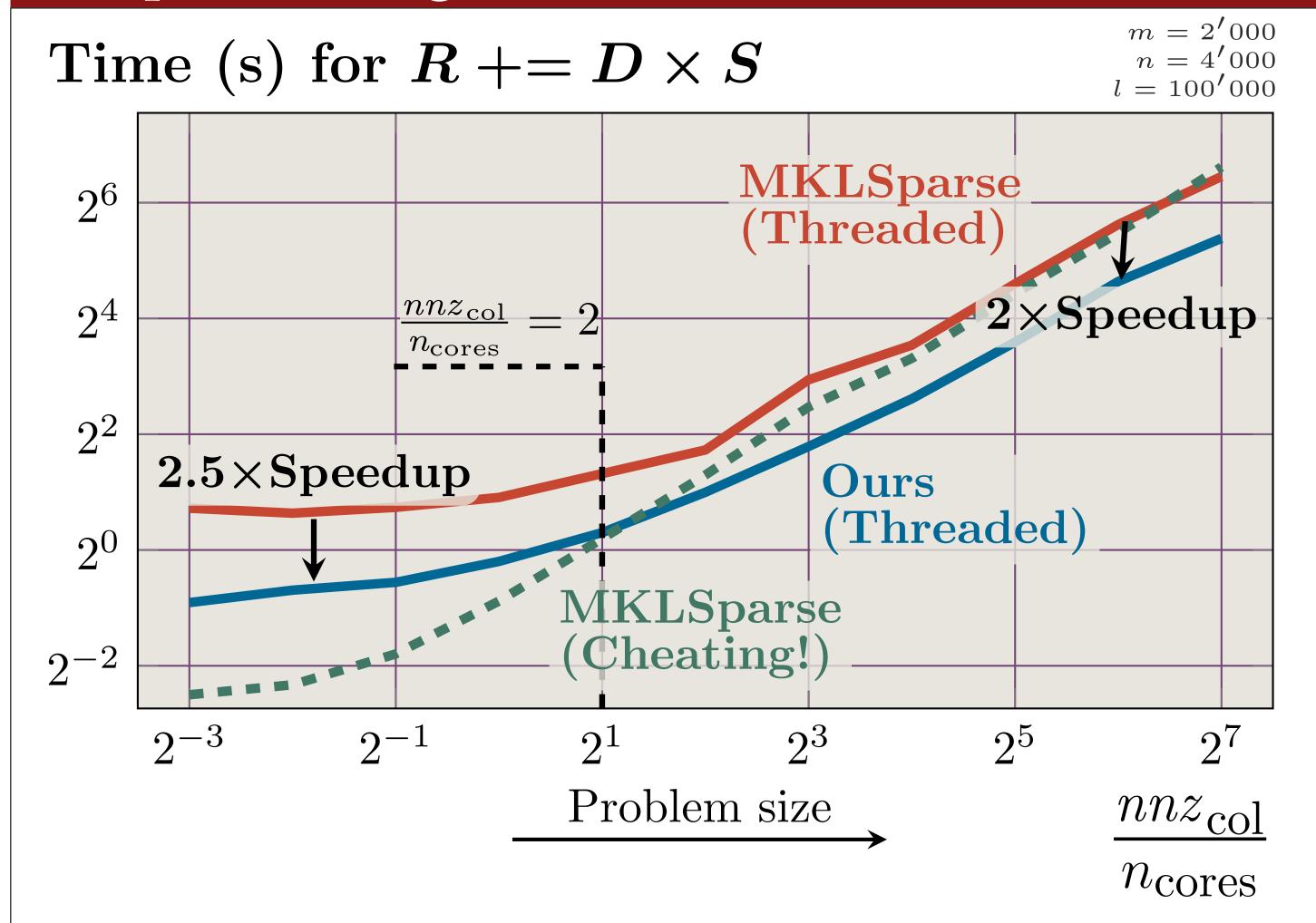


i.e., threaded dense times sparse matmul.

Our options:

• ThreadedDenseSparseMul.jl (this poster)

Outperforming MKLSparse: Benchmark results.



Ours outperforms MKLSparse with speedup $2\times!$ For large $nnz_{\rm col}$, Ours even outperforms MKLSparse (Cheating!) which computes sparse×dense directly.

Why can we win this?

Better memory layout. MKLSparse transposes (twice). Let's count <u>memory movements</u> of mul!(R', S', D')'.

- Transpose: $\approx 2 \times \Theta(m \cdot l)$
- Multiply: $\Theta(nnz_{\text{col}} \cdot m \cdot l)/n_{\text{cores}}$

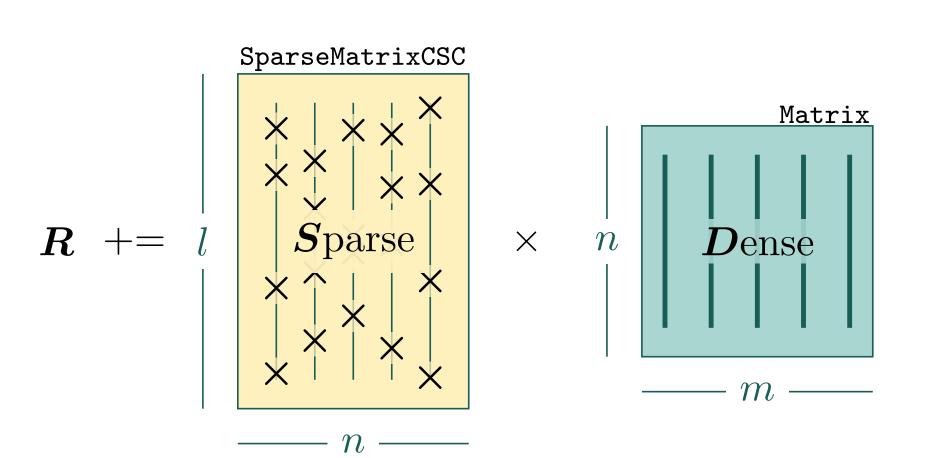
$$\Rightarrow \left(2 + \frac{nnz_{col}}{n_{cores}}\right) \times \Theta(m \cdot l)$$

 \Rightarrow For $(nnz_{\rm col} > 2 \times n_{\rm cores})$, transposing dominates!

Takeaways

- Good memory layout lets us outperform MKL.
- Don't be afraid to "hand-roll" tight loops.
- Know when you are memory or compute bound.
- Benchmark, benchmark, benchmark!

...but everyone else does



i.e., threaded <u>sparse times dense</u> matmul.

Our options:

MKLSparse.jl, ThreadedSparseCSR.jl, ThreadedSparseArrays.jl

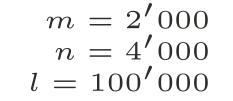
The Secret Sauce: Chris Elrod's Polyester.jl

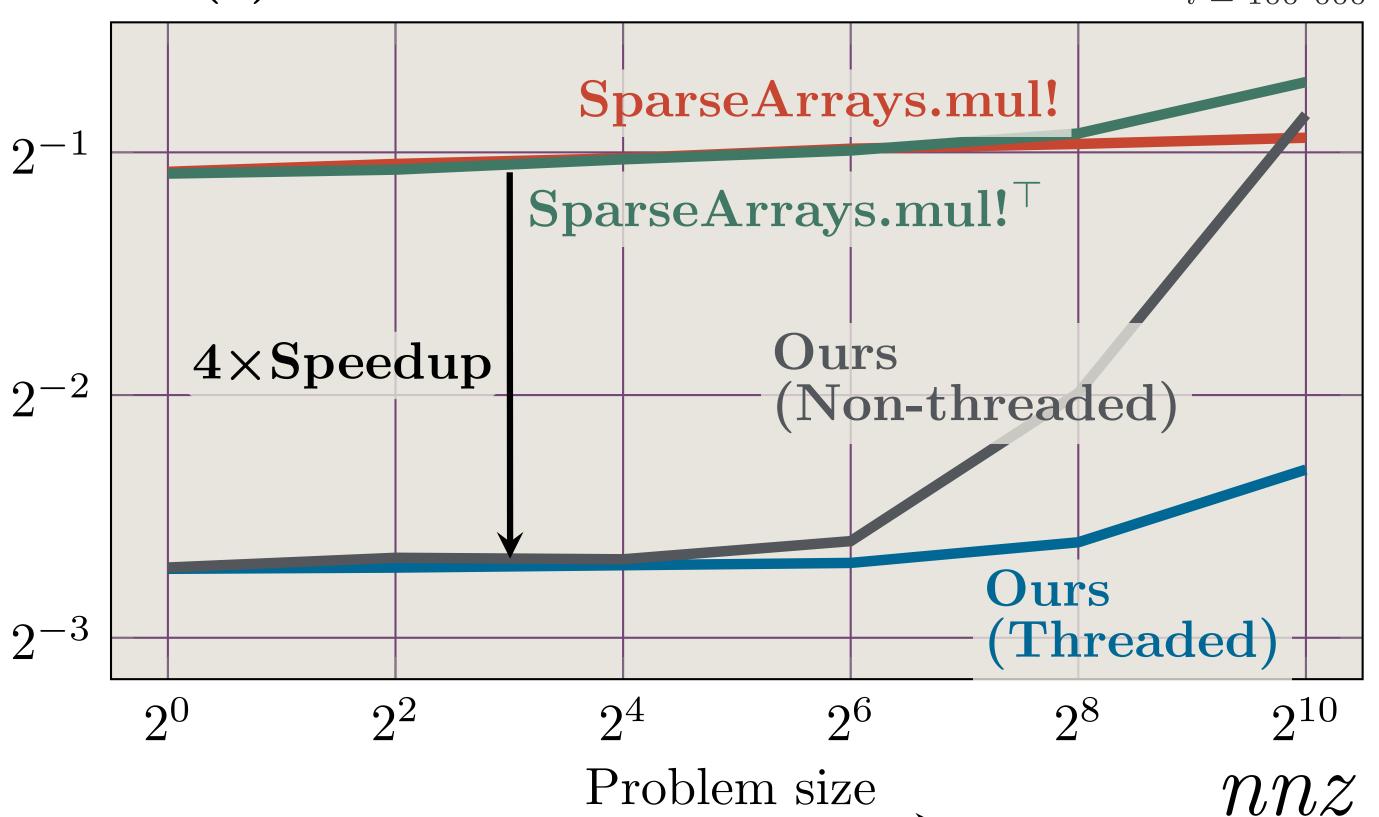
Hand-rolling loops has never been so easy! Also, memory layout is actually perfect for dense×sparse!

Things get even better: outer product.

MKLSparse doesn't have an outer product at all! And we can outperform SparseArrays.

Time (s) for
$$R += \vec{d} \times \vec{s}^{\top}$$





Ours beats SparseArrays.mul!/mul!^{\top} with speedup $4 \times \text{in 5 LOC}$ (single-threaded, no Polyster.jl), or 9 LOC (multi-threaded, w/ Polyster.jl).

All benchmarks collected on mobile workstation with 11th Gen Intel(R) Core(TM) i7-11800H @ 2.30GHz, 32GB RAM, using the Chairmarks.jl library.

