TOWARDS AN AUTO-TUNED AND TASK-BASED SPMV (LASS LIBRARY)

SANDRA CATALÁN, TETSUZO USUI, LEONEL TOLEDO, XAVIER MARTORELL, JESÚS LABARTA, PEDRO VALERO-LARA



SPMV

$$y := \alpha A x + \beta y$$

$$\text{DENSE}$$
SCALAR

SPARSE

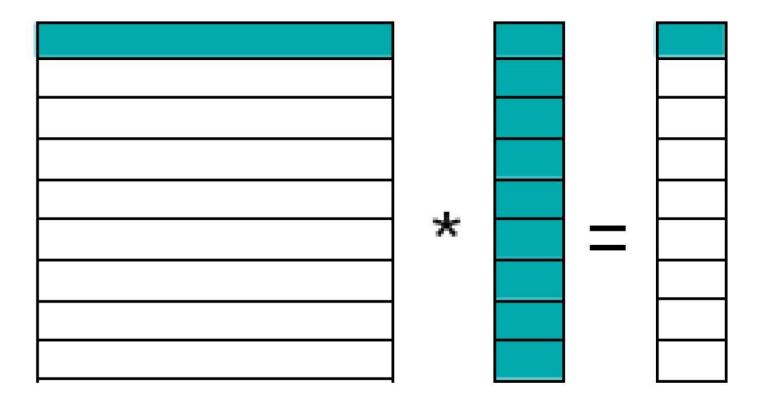
STATE OF THE ART

- Many libraries implement SpMV: MAGMA-Sparse, SuperLU, MUMPS -> LASs
- Many studies about best storage format -> CSR
- There are works about performance and different formats/programming models ->
 OmpSs + CSR +un/balancend matrices

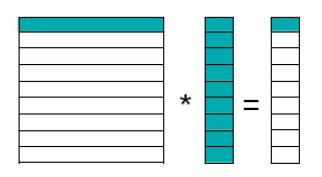
PARALLELIZING SPMV

- One task per row
- Blocking
- Taskloop
- Grouping

ONE TASK PER ROW

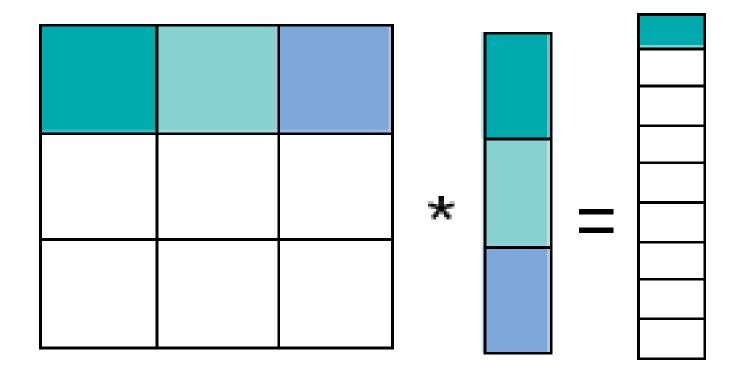


ONE TASK PER ROW

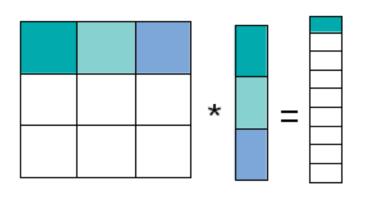


- ✓ Simple and straight-forward
- ✓ No data dependencies
- × Numerous tasks are created
- × Tasks usually very small (low number of non-zeros per row) -> overhead

BLOCKING

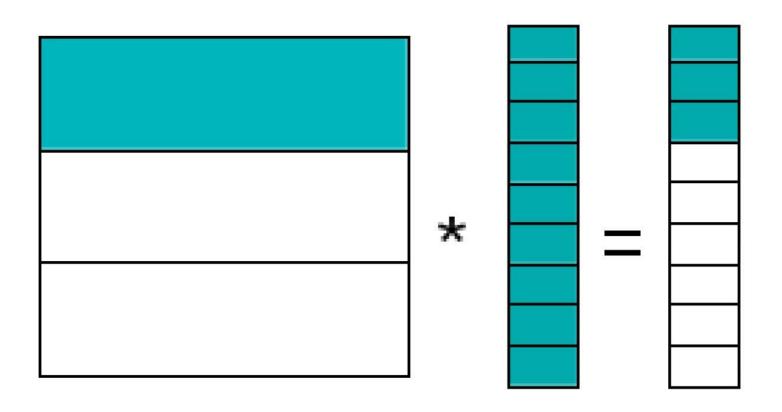


BLOCKING

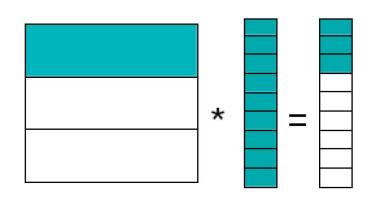


- ✓ Reuse of y entries within a task
- ✓ Successfully tested in other operations (CG)
- × Preprocess required
- × Data dependencies
- × Non-negligible changes in the code

TASKLOOP

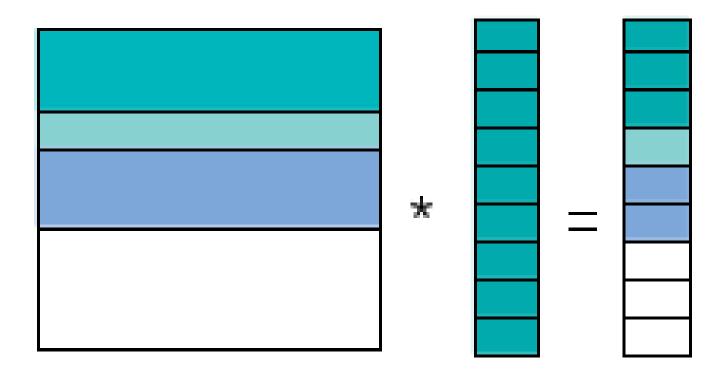


TASKLOOP

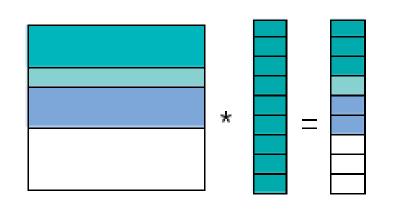


- ✓ Simple
- ✓ Allows coarser tasks
- × Grainsize needs to be computed
- × Potential unbalance

GROUPING



GROUPING



- ✓ Balanced tasks
- ✓ Successfully tested in other operations (Batched GEMM)
- × Extra calculations required
- × One core devoted to create groups

TEST PLATFORMS

Marenostrum

- Single node of Marenostrum 4 (2 sockets Intel Xeon Platinum 8160)
- 24 cores each at 2.10GHz (48 cores in total)
- Memory hierarchy: each core has 32KB L1 and 1MB L2 caches, and 33MB L3 (cache shared)

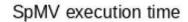
Dibona

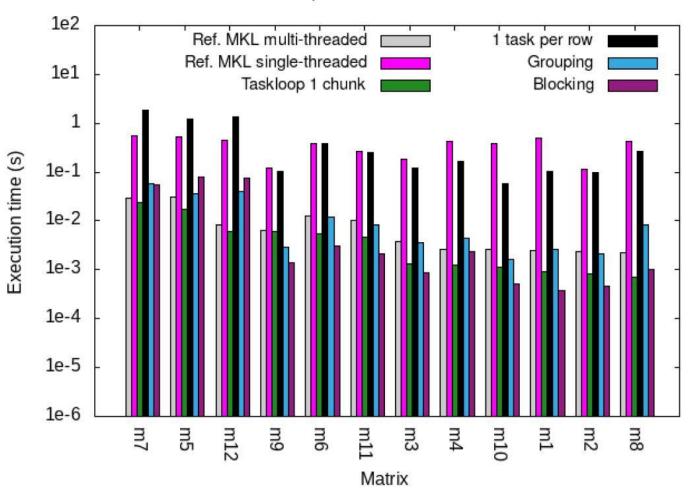
- Single node two sockets ARM Thunder X2 (ARMv8 NEON) CPU with
- 32 cores each at 2.0GHz (64 cores in tota)
- Memory hierarchy: each core has 32KB of L1 cache, 256KB L2 cache, and 32MB L3 cache

INPUT MATRICES

ID	Name	Domain	#rows	NNZ	Max.	Min.	Avg.	Matrix
m1	cant	FEM Cantilever	62,451	2,034,917	40	1	32	
m2	conf5_4-8x8-05	Quantum chromody- namics	49,152	1,916,928	39	39	39	
m3	consph	FEM concentric spheres	83,334	3,046,907	66	1	36	
m4	cop20k_A	Accelerator cavity design	121,192	1,362,087	24	0	11	
m5	eu-2005	Small web crawl of .eu domain	862,664	19,235,140	6,985	0	22	
m6	Ga41As41H72	Real-space pseudo potential method	268,296	9,378,286	472	1	34	1
m7	in-2004	Small web crawl of in domain	1,382,908	16,917,053	7,753	0	12	
m8	mac_econ_fwd500	Macroeconomic model	206,500	1,273,389	44	1	6	
m9	mip1	Optimiation problem	66,463	5,209,641	713	1	78	
m10	pdb1HYS	Protein data bank 1HYS	36,417	2,190,591	184	1	60	
m11	Si41Ge41H72	Real-space pseudo potential method	185,639	7,598,452	531	1	40	
m12	webbase1-M	Web connectivity matrix	1,000,005	3,105,536	4,700	1	3	

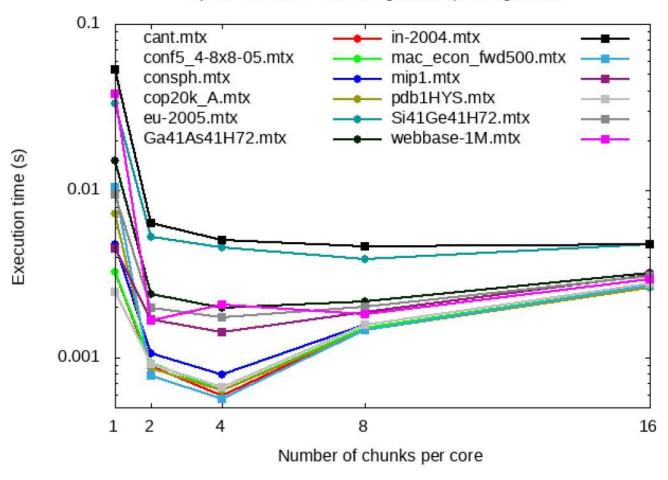
PERFORMANCE RESULTS





THE RIGHT GRAINSIZE IN TASKLOOP

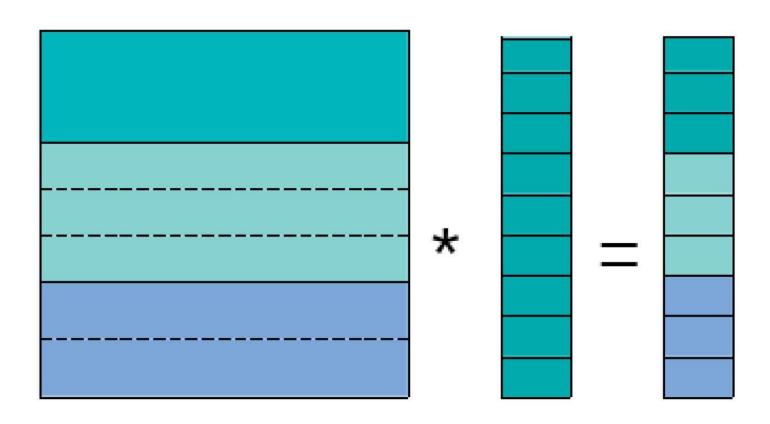




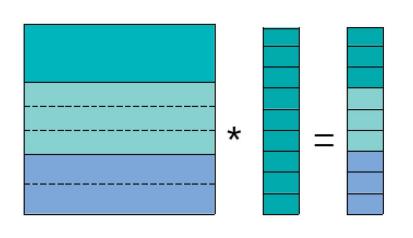
HOW GRAINSIZE AFFECTS TASK DISTRIBUTION



ADAPTED NUMBER OF TASKS: TASKLOOP + NESTING

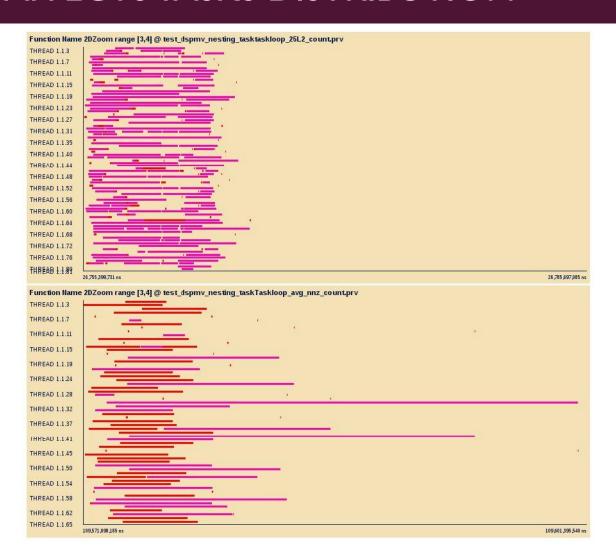


ADAPTED NUMBER OF TASKS: TASKLOOP + NESTING

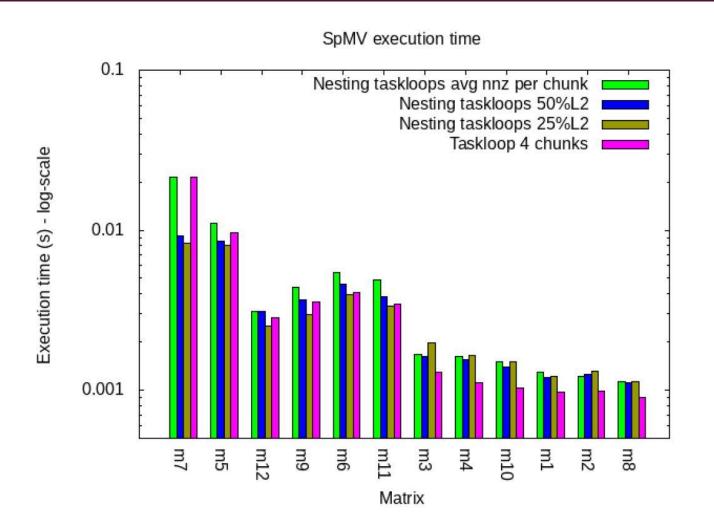


- ✓ No data dependencias
- ✓ Balanced nnz per task
- × More complex code
- × A threshold needs to be choosen

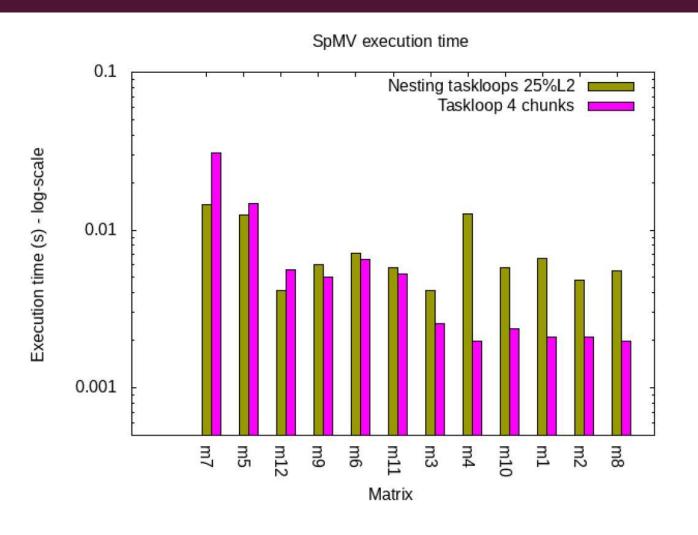
HOW NESTING AFFECTS TASKS DISTRIBUTION



RESULTS ON INTEL



RESULTS ON ARM



CONCLUSIONS

- Taskloop worked well for balanced sparse matrices
- Taskloop+nesting can do better in unbalanced sparse matrices
- Detecting unbalance is important

FUTURE WORK

- Combining both strategies
- Include the result in LASs

THANKS!!

Any question?