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Scientific Programming with Python

Hardware Speedup Exercises

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Exercise 1: Optimizing arithmetic expressions

1. Use script `poly.py` to check how much time it takes to evaluate the next polynomial:

$y = .25x^3 + .75x^2 - 1.5x - 2$

with x in the range $[-1, 1]$, and with 10 millions points.

- Set the 'what' parameter to "numexpr" and take note of the speed-up versus the "numpy" case. Why do you think the speed-up is so large?
2. The expression below:
 $y = ((.25x + .75)x - 1.5)x - 2$
represents the same polynomial than the original one, but with some interesting side-effects in efficiency. Repeat this computation for numpy and numexpr and get your own conclusions.
 - Why do you think numpy is doing much more efficiently with this new expression?
 - Why the speed-up in numexpr is not so high in comparison?
 - Why numexpr continues to be faster than numpy?
 3. The C program `poly.c` does the same computation than above, but in pure C. Compile it like this:
`gcc -O3 -o poly poly.c -lm`
and execute it.

- Why do you think it is more efficient than the above approaches?

Exercise 2: Evaluating transcendental functions

4. Activate the evaluation of the `sin(x)**2+cos(x)**2` expression in `poly.py`, a function that includes transcendental functions and run the script.
 - Why the difference in time between NumPy and Numexpr is so small?

5. In `poly.c`, comment out expression 2) (around line 53) and uncomment expression 4) (the transcendental function).
 - Do this pure C approaches go faster than the Python-based ones?
 - What would be needed to accelerate the computations?

Exercise 3: Using Numba

The goal of Numba is to compile arbitrarily complex Python code on-the-flight and executing it for you. It is fast, although one should take in account the compile times.

6. Edit `poly-numba.py` and look at how numba works.
 - Run several expressions and determine which method is faster. What is the compilation time for numba and how it compares with the execution time?
 - Raise the amount of data points to 100 millions. What happens?

Exercise 4: Parallelism

7. Be sure that you are on a multi-processor machine and activate the:
`y = ((.25*x + .75)*x - 1.5)*x - 2`
expression in `poly-mp.py`. Repeat the computation for both numpy and numexpr for a different number of processes (numpy) or threads (numexpr) (pass the desired number as a parameter to the script).
 - How the efficiency scales?
 - Why do you think it scales that way?
 - How performance compares with the pure C computation?
8. With the previous examples, compute the expression:
`y = x`
That is, do a simple copy of the 'x' vector. What is the performance that you are seeing?
 - How does it evolve when using different threads? Why it scales very similarly than the polynomial evaluation?
 - Could you have a guess at the memory bandwidth of this machine?