



Strongly Performing Python Implementation of the HPC Challenge

Interactive Supercomputing, Inc.

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Star-P enables Python, MATLAB®, and R users to go parallel easily with competitive performance.



Approach to HPC Challenge

- Create Python version of 4 HPCC benchmarks
 - HPL, Stream, Random Access, and FFT
 - Why Python?
- Parallelize with Star-P constructs
- Measure and tune

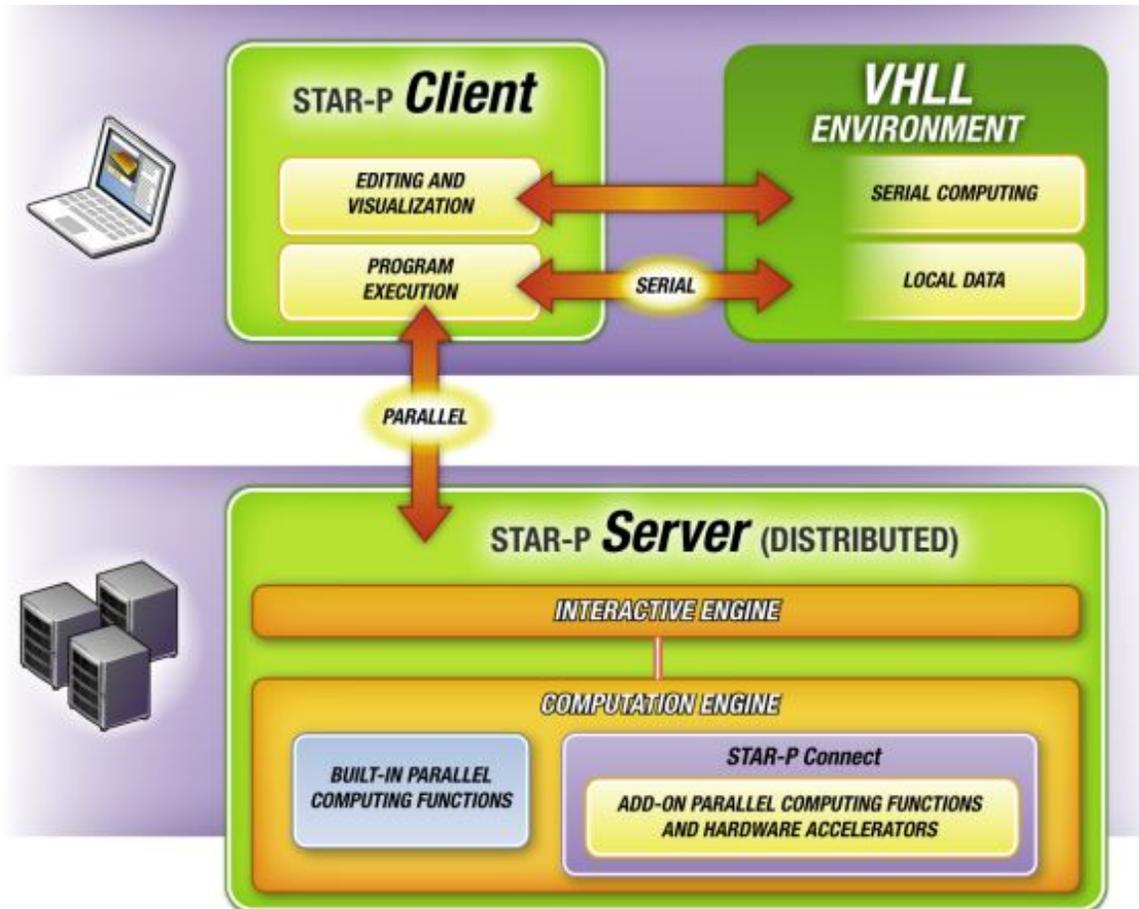
Star-P Basics:

Bridges the gap between desktop tools and parallel computing systems



Value proposition

- Rapid, interactive apps development
- Potent high-level parallel abstractions
- Minimize code changes
- High speed and/or large memory
- MATLAB® and Python clients today, R soon
- Scales to 100s of cores, >4TB memory
- Extensible with existing serial or MPI-parallel libraries





Star-P/Python Parallel Constructs

Task-Parallel

- Iterations clearly separable
- Use Star-P's parallel iterator

Data Parallel

- Large monolithic data
- Create distributed arrays
 - Distributed attribute propagates to result variables



HPL Source Code

```
def run_hpl(n, nr, tol=16):
    """
    Run the High-performance LINPACK test on a matrix of size n x n, nr
    number of times and ensures that the the maximum of the three
    residuals is strictly less than the prescribed tolerance (defaults
    to 16).

    This function returns the performance in GFlops/Sec.
    """
    a = random.rand(n, n);
    b = random.rand(n, 1);
    x,t = iterate_func(nr, linalg.solve, a, b)

    r = dot(a, x)-b
    r0 = linalg.norm(r, inf)
    r1 = r0/(eps * linalg.norm(a, 1) * n)
    r2 = r0/(eps * linalg.norm(a, inf) * linalg.norm(x, inf) * n)

    performance = (1e-9 * (2.0/3.0 * n * n * n + 3.0/2.0 * n * n) *
        nr/t)
    verified = numpy.max((r0, r1, r2)) < 16

    if not verified:
        raise RuntimeError, "Solution did not meet the prescribed tolerance
        %d"%tol
    return performance
```



STREAM Source Code

```
def run_epstream(n, nr):  
    """  
    Run the embarrassingly parallel stream benchmark on vectors of size  
    n, nr number of times.  
  
    This function returns the performance of the benchmark in  
    GFlops/second.  
    """  
    s = random.rand(1);  
    a = random.rand(n);  
    b = random.rand(n);  
    c,t = iterate_func(nr, lambda s, a, b: s*a+b, s, a, b)  
  
    performance = (1e-9) * 24.0 * nr * n / t  
  
    return performance
```



Random Access Source Code

```
import time, optparse
import starp as sp

def update_state(ran, idx, table_size):
    sp.runCommand('rng_update_state', ran, idx, table_size);

def update(table_size, n_in, n_out):
    t1 = 0;
    t0 = time.time()
    t = sp.arange(table_size)
    t1 += (time.time() - t0)
    ran = sp.zeros(n_in, )
    idx = sp.zeros(n_in, )

    for outer in xrange(n_out):
        update_state(ran, idx, table_size)
        t0 = time.time();
        t[idx] ^= ran
        t1 += (time.time() - t0);

    return 1.0e-9 * n_in * n_out/float(t1);

def run_random_access(n, nr):
    n_in = 1024;
    n_out = nr/n_in;

    if n_out * n_in != nr:
        raise ValueError("Number of updates must be evenly divisible by %d" %
n_in)

    return update(n, n_in, n_out);
```



FFTE Source Code

```
def run_fft(n, nr, tol=16):
    """
    Run the one-dimensional FFT benchmark on a vector of size n, nr
    number of times and verifies that the inverse transforms recreates
    the original vector upto a tolerance, tol (defaults to 16).

    This function returns the performance in GFlops/sec.
    """
    a = random.rand(n,1)
    b, t = iterate_func(nr, fft.fft, a)

    log2n = math.log(n)/math.log(2)
    performance = 1e-9 * 5.0 * n * log2n * nr/t
    verified = linalg.norm(a - (fft.ifft(b))) / (eps * log2n) < tol

    if not verified:
        raise RuntimeError, "Solution did not meet the tolerance %d"%tol

    return performance
```



Product Scalability: Does this work in any other industry?



How the coffee industry treats someone ordering a Large



How the computing industry treats VHLL language users wanting a Large



Code Attributes

<i>Benchmark</i>	<i>SLOC (Python/ Star-P)</i>	<i>SLOC (MPI)</i>	<i>Distance to Desktop</i>
<i>framework</i>	63	?	2
<i>HPL</i>	13	15608	0
<i>STREAM</i>	6	658	0
<i>RandomAccess</i>	46 (+71 C++)	1883	6 (+71 C++)
<i>FFTE</i>	8	1747	0

- Implemented with Star-P 2.5.1 (currently shipping product)
- Developed on small in-house system, scaled directly to 128-core system at SDSC
- Difference from desktop, in framework for HPL/Stream/FFTE

```
if nproc == 0:
    from numpy import *
else:
    from starp import *
```

 - Can exert greater control with more code changes
- RandomAccess:
 - Not a good match for current Python/Star-P
 - Used custom 27-line C++ kernel

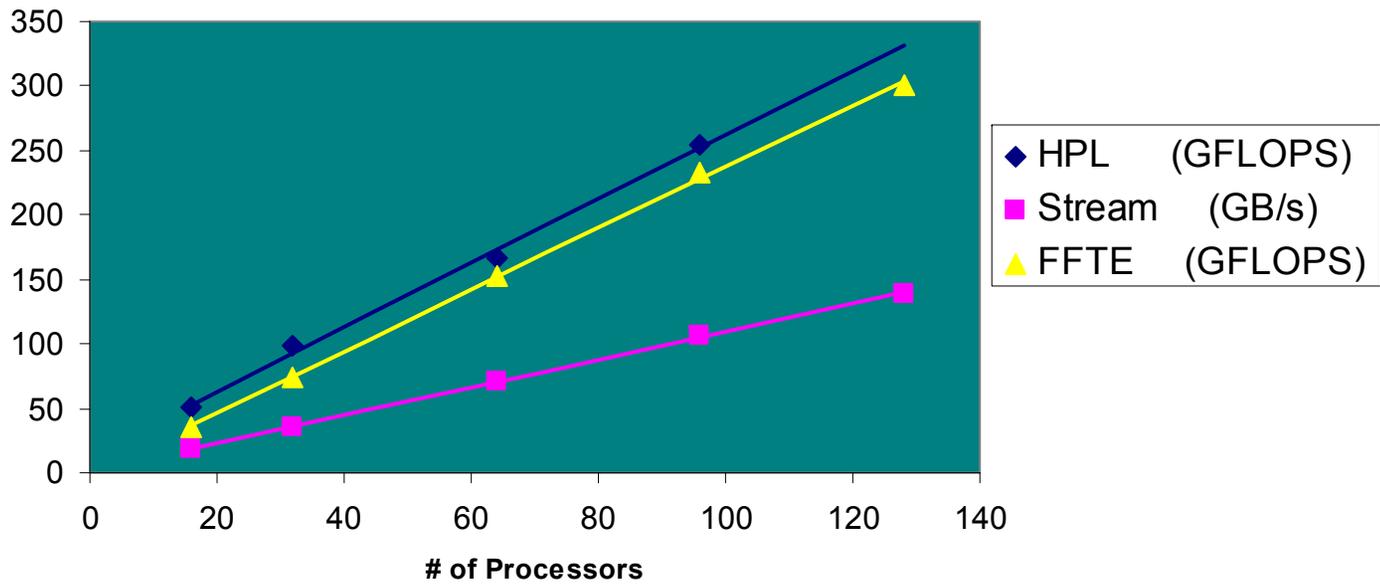


Performance

#cores	HPL (GFLOPS)	Stream (GB/s)	FFTE (GFLOPS)	RandomAccess (GUPS)
16	50.7	17.894	35.627	0.00118
32	98.901	35.626	74.037	0.00194
64	165.969	70.769	152.574	
96	254.221	106.504	232.475	
128		139.475	299.976	

- Strong absolute performance
- Strong scalability

Scaling Results





Relevance for General HPC

- HPCC benchmarks (except RandomAccess) lend themselves to trivial task- or data-parallel expression
 - Data analysis codes are similar, and need rapid development
- Typical HPC apps have more complex data sharing patterns and depend more on many simpler functions, not one large function



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For our full description, go to
www.InteractiveSupercomputing.com/applibrary/