nb-l27-multiproc

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1 Multiprocessing in Julia

- Add workers using `addprocs`
- Start a function as a Task on an available thread using `remotecall`.
- `fetch(task)` wait for the completion of the taks and retrieve result
- `remotecall_fetch` does `remotecall` and `fetch`

[1]:
```
using Distributed
using LinearAlgebra
using BenchmarkTools

addprocs(4)
```

[1]:
```
4-element Array{Int64,1}:
 2
 3
 4
 5
```

We can also do `addprocs([(hostname,n)])` to work on different hosts

List of workers

[2]:
```
workers()
```

[2]:
```
4-element Array{Int64,1}:
 2
 3
 4
 5
```

Run a function on all threads

[3]:
```
@everywhere println(myid())
```

```
1
      From worker 2:    2
      From worker 3:    3
      From worker 4:    4
      From worker 5:    5
```

Run a function on another worker and return its id multiplied by 10

[4]:
```
@everywhere function run_on()
    return myid()*10
end

remotecall_fetch(run_on,3)
```

[4]:
```
30
```

Distributed Arrays allow to distribute data to all workers

[5]:
```
@everywhere using DistributedArrays
```

Now let us try to calculate a scalar product

Scalar product for two arrays

[6]:
```
@everywhere function mydot(A::Array,B::Array)
    result=0.0
    @inbounds @fastmath for i=1:length(A)
        result+=A[i]*B[i]
    end
    return result
end
```

Scalar product for two distributed arrays

This uses an asynchronous map, where results are collected as they come in

[7]:
```
function mydot(DA::DArray,DB::DArray)
    results=asyncmap(p->remotecall_fetch((DA, DB) -> mydot(localpart(DA),␣
 ↪localpart(DB)),p,DA,DB), workers() )
    reduce(+,results)
end
```

[7]:
```
mydot (generic function with 2 methods)
```

[8]:
```
A=rand(1_000_000)
B=rand(1_000_000)
DA=distribute(A)
DB=distribute(B);

res_s=@btime mydot($A,$B)
```

```
res_p=@btime mydot($DA,$DB)
res_s  res_p
```

```
  392.439  s (0 allocations: 0 bytes)
  428.211  s (418 allocations: 17.58 KiB)
```

[8]: `true`

- Due to communication and data distribution overhead, this is more efficient for coarser grained parallelism

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This notebook was generated using Literate.jl.