

High Performance Computing & Big Data Services hpc.uni.lu hpc@uni.lu weulhpc

Uni.lu HPC School 2020

PS08: Advanced Distributed Computing with Python

Uni.lu High Performance Computing (HPC) Team E. Kieffer

University of Luxembourg (UL), Luxembourg

http://hpc.uni.lu





Latest versions available on Github:



UL HPC tutorials:

UL HPC School:

PS08 tutorial sources:

https://github.com/ULHPC/tutorials

http://hpc.uni.lu/hpc-school/

ulhpc-tutorials.rtfd.io/en/latest/python/advanced/























Summary

- Introduction
- 2 Parallel machine learning with ipyparallel
- 3 Parallel evolutionary computing with scoop





Main Objectives

- How to parallelise your python code?
- Mereafter, we are going to see two alternatives :
 - → High-level approach with ipyparallel for scikit-learn
 - \hookrightarrow Low-level approach with scoop





Parallel machine learning with ipyparallel

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Scikit-learn + ipyparallel

- Scikit-learn is a well-known python scientific package:
 - \hookrightarrow Machine learning algorithms (e.g. SVM)
 - → Data analysis approaches (e.g. PCA)
 - → Data mining techniques (e.g. Clustering)
- Scikit-learn algorithms can be parallelised
- Especially useful for Hyper-parameters search
- Scikit-learn relies on ipyparallel and joblib libraries to parallelism algortihms





Ipyparallel

- Originally designed under lpython
- IPython's architecture for parallel and distributed computing
- Support many different styles of parallelism:
 - → Single program, multiple data (SPMD) parallelism
 - → Multiple program, multiple data (MPMD) parallelism
 - \hookrightarrow Message passing using MPI
 - → Task farming
 - → Hybrid approaches combined the above ones
- Ipyparallel can detect a job scheduler (e.g. Slurm) when started on a HPC platform





Practical session

Please go to https://ulhpc-tutorials.readthedocs.io/en/latest/python/advanced/scikit-learn/





Parallel evolutionary computing with scoop

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Scoop + deap

- Deap
- Python evolutionary computing library:
 - Genetic algorithms
 - \hookrightarrow Particle swarm algorithms
 - \hookrightarrow Evolutionary strategies
 - \hookrightarrow Estimation of Distribution algorithms
- Deap relies on scoop

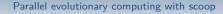




Scoop

- SCOOP => Scalable COncurrent Operations in Python
- Applications of SCOOP:
 - \hookrightarrow Evolutionary algorithms
 - → Monte Carlo simulations
 - → Data mining
 - → Data processing
 - Graph traversam
- Very simple to use
- Override default map (reduce) function







Practical session

Please go to https://ulhpc-tutorials.readthedocs.io/en/latest/python/advanced/scoop-deap/





Thank you for your attention...



Questions?

High Performance Computing @ Uni.lu



University of Luxembourg, Belval Campus Maison du Nombre, 4th floor 2, avenue de l'Université L-4365 Esch-sur-Alzette mail: hpc@uni.lu Introduction

Parallel machine learning with ipyparallel

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