$(\triangle) EasyVVUQ Bringing your UQ workflows () E C M A$ to the exascale!



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What is **EasyVVUQ**? https://github.com/UCL-CCS/EasyVVUQ

EasyVVUQ is a Python 3 library designed to facilitate verification, validation and uncertainty quantification (VVUQ) for a wide variety of simulations. It was conceived and developed within the EU funded VECMA (Verified Exascale Computing for Multiscale Applications) project.

Large scale execution

Direct submission of a large group of jobs to a scheduling system can result in long aggregated time to finish as each single job is scheduled independently and waits in the queue. To ensure efficient allocation, we create large container jobs within which we can then allocate and schedule the individual simulation runs. To do this, we can use either the QCG PilotJob^[3] component in the VECMA Toolkit or Dask JobQueue^[5]. QCG PilotJob Manager is designed to schedule and execute many jobs inside a single allocation. It provides a container for sub-jobs, for which necessary resources are then assigned efficiently. In order to simplify the use of QCG PilotJob from EasyVVUQ workflows, a lightweight integrating code EASYVVUQ-QCGPJ^[4] is provided (e.g. for Fusion and UrbainAir applications^[7]). Users can also choose to use Dask JobQueue, which is useful for smaller runs. For two applications (migration and CFD^[7]) we combined complex workflows, using automation, remote access and curation features from FabSim3^[6], with VVUQ offered by EasyVVUQ. This integration can be made either by using the FabSim3 API in a script, or by incorporating EasyVVUQ in a FabSim3 plugin.

We use the existing Python scientific computing infrastructure and extend it with tools for executing the jobs in HPC environments. We also automate away most of the tedious details of UQ usage scenarios.

Why You Need EasyVVUQ?

UQ typically requires a very large number of jobs to be run and analysed. Most existing UQ tools are not designed with large scale execution in mind. Therefore, we have designed EasyVVUQ to augment them with the capabilities to run on modern HPC machines. The framework is designed to be open-ended with regards to the execution of simulations. It is currently possible to execute jobs locally or on HPC clusters using QCG pilot job or Dask JobQueue.

Existing simulations can make use of EasyVVUQ by extending several of the classes that make up the framework. In most cases these would just be the Encoder and Decoder responsible for parsing the inputs and outputs of your simulation respectively.

The end goal of EasyVVUQ is to allow you to test your UQ procedures on your laptop and then take it directly to a supercomputer. It is designed to be flexible, fault tolerant and resumable. In the code example below, we outline a typical use scenario. It involves a toy simulation of a cooling coffee cup and goes through all the main stages of a UQ run with EasyVVUQ.



Conceptual Model

Partners



Overall workflow

Campaign

my campaign = uq.Campaign(name='cooling cup')

Parameter Space Description

vary = $\{$ "kappa": cp.Uniform(0.025, 0.075), "t env": cp.Uniform(15, 25)

Encoder

encoder = uq.encoders.GenericEncoder(template_fname='tutorial_files/cooling.template', delimiter='\$', target filename='cooling in.json')

Decoder

decoder = uq.decoders.SimpleCSV(target filename="output.csv", output_columns=["te"], header=0)

Sampling

my sampler = uq.sampling.PCESampler(vary=vary, polynomial order=3) my campaign.draw_samples()

Model evaluation

my_campaign.apply_for_each_run_dir(uq.actions.ExecuteLocal(app))

Aggregation

my_campaign.collate()

Dask JobQueue

my_campaign = uq.CampaignDask(name='coffee pce') cluster = SLURMCluster(job extra=['--cluster=mycluster'], queue='myqueue', cores=48, memory='64 GB')

client = Client(cluster)

cluster.scale(96)

OR

QCG-PJ qcgpjexec = Executor() qcgpjexec.create_manager(dir=my_campaign.campaign_dir) qcgpjexec.add_task(Task(TaskType.EXECUTION, TaskRequirements (cores=Resources (exact=1)), application=app qcgpjexec.run(

Analysis

analysis = uq.analysis.PCEAnalysis(sampler=my sampler, qoi cols=output columns)

Descriptive Statistics

results = my_campaign.get_last_analysis()

campaign=my_campaign, submit order=SubmitOrder.EXEC ONLY) qcgpjexec.terminate_manager()

References & Acknowledgements

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[7] VECMAtk application tutorials <u>https://www.vecma-toolkit.eu/tutorials/</u>



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