Using Xopt for Optimization

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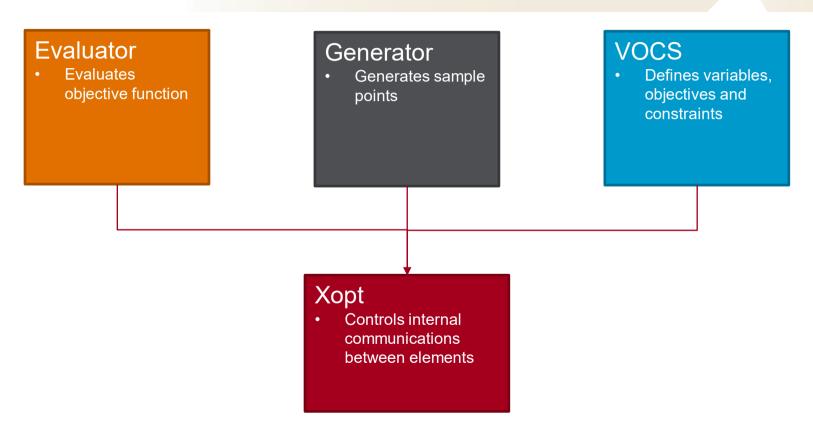


- Flexible framework for optimization of arbitrary problems using python
- Independent of problem type (simulation or experiment)
- Independent of optimization algorithm + easy to incorporate custom algorithms
- Easy to use text interface and/or advanced customized use for professionals



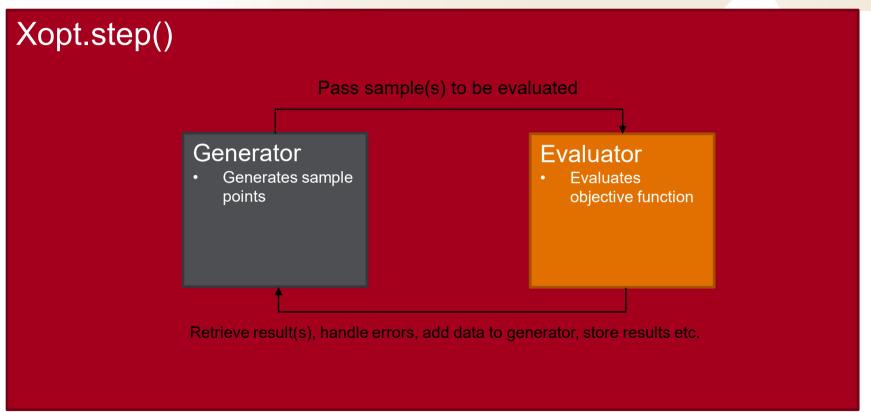
Xopt structure





Synchronous execution





Via YAML file (validated by pydantic):

```
xopt:
    max evaluations: 6400
generator:
    name: cnsga
    population size: 64
    population file: test.csv
    output path: .
evaluator:
    function: xopt.resources.test functions.tnk.evaluate TNK
    function kwargs:
     raise probability: 0.1
vocs:
    variables:
        x1: [0, 3.14159]
        x2: [0, 3.14159]
    objectives: {y1: MINIMIZE, y2: MINIMIZE}
    constraints:
        c1: [GREATER THAN, 0]
        c2: [LESS THAN, 0.5]
    linked variables: {x9: x1}
    constants: {a: dummy constant}
```

Via python code:

```
evaluator = Evaluator(...)
generator = CNSGAGenerator(...)
vocs = MyVOCS(...)
X = Xopt(
        evaluator=evaluator,
        generator=generator,
         VOCS=VOCS
```

- Python function must accept/return dicts
- Input dict must have at least the keys specified in vocs variables/constants (see next slide)
 - You can include extra keyword args if needed!
- Output dict must have at least the keys specified in objectives/constraints (see next slide)
 - The function can output extra keys to be tracked!
- Functions can be defined at the module level and passed via string if they are in PYTHONPATH, they can also be passed inside the same python file (use __main__.my_function)
- Evaluators inherit directly from python concurrent.futures so you can use this for parallel evaluation (see /xopt/docs/examples/basic/xopt_parallel)

```
xopt:
   max evaluations: 6400
generator:
         evaluate(inputs: dict) -> dict
   output path: .
evaluator:
   function: xopt.resources.test functions.tnk.evaluate TNK
   function kwargs:
      raise probability: 0.1
vocs:
    variables:
       x1: [0, 3.14159]
       x2: [0, 3.14159]
   objectives: {y1: MINIMIZE, y2: MINIMIZE}
    constraints:
       c1: [GREATER THAN, 0]
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    linked variables: {x9: x1}
   constants: {a: dummy constant}
```

- Variables: input domain limits and names
- Objectives: objective names and goals (minimize/maximize)
- Constraints: constraint names and conditions (greater than/less than)
- Constants: constant values

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    function: xopt.resources.test functions.tnk.evaluate TNK
    function kwargs:
      raise probability: 0.1
VOCS:
    variables:
       x1: [0, 3.14159]
        x2: [0, 3.14159]
    objectives: {y1: MINIMIZE, y2: MINIMIZE}
    constraints:
        c1: [GREATER THAN, 0]
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    linked variables: {x9: x1}
    constants: {a: dummy constant}
```

- Use built-in generators by name
- · optimization algorithms:
 - o cnsga Continuous NSGA-II with constraints.
 - o bayesian_optimization Single objective Bayesian optimization (w/ or w/o constraints, serial or parallel).
 - o mobo Multi-objective Bayesian optimization (w/ or w/o constraints, serial or parallel).
 - o bayesian exploration Bayesian exploration.
- · sampling algorithms:
 - o random sampler
- Each generator has its own specific options
- Locate the default options in the docs or via

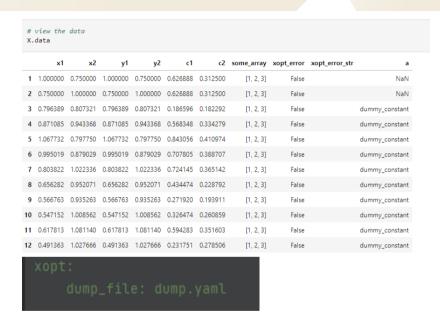
```
from xopt.utils import get generator and defaults
gen, options = get generator and defaults("upper confidence bound")
print(yaml.dump(options.dict()))
acq:
  beta: 2.0
 monte carlo samples: 512
 proximal lengthscales: null
model:
 use conservative prior lengthscale: false
 use conservative prior mean: false
 use low noise prior: false
n initial: 3
optim:
  num restarts: 5
  raw samples: 20
  sequential: true
```

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Data storage



- Data is stored by xopt in the `data` attribute
- Set dump_file in xopt
 options to dump data and
 xopt config to yaml file
 after every evaluation
 step
- Dump file can be used to restart xopt



Tips and Tricks



- Look at the examples in docs/examples !!!!
- Get creative with the evaluate function to track variables/outputs.
- Ask for invite to #xopt channel
- Always looking for help!

