Data Units and Graph-Based Inputs to defineQuanta

This page is a design proposal to address the problems in the baseline SuperTask noted in the last section of SuperTask Status: Science Pipelines Perspective. In this design, we replace the camera-dependent dictionary data IDs currently used by the Butler with a more structured set of classes that are camera-independent. Camera-specific data ID keys will still be relevant in the user-provided expressions that describe that output datasets should be produced and what input datasets should be considered, but these expressions will be explicitly transformed into to a new data structure containing only camera-independent classes before they come into contact with any SuperTask.

Units and Datasets

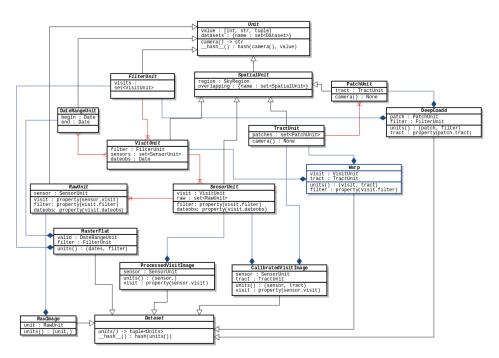
A Unit is a conceptually similar to a data ID key (i.e. "visit", "ccd", "patch"), and a particular unit may (for some cameras) map exactly to an existing data ID key. Units, however, are considered to be generically important quantities from an algorithmic perspective, and hence they are camera-generic. A Dataset is analogous to a combination of a datasetType and its data ID; a Dataset class corresponds to a datasetType, while an instance adds a tuple of Unit instances that represent the data ID in a camera-independent way.

In code, Unit is an abstract base class that holds an immutable unique identifier whose type is defined by the subclass, such as an integer ID for visits, a tuple of integers for patches, or a string name for filters. The actual types and values of these IDs are unimportant, and may in fact be camera-dependent, as long as the values are globally unique identifiers within a camera. A Unit also holds a dictionary of all the datasets that are defined on that unit (i.e. all of the ProcessedVisitImages for a particular visit), and has an abstract camera() method that must be implemented by concrete derived classes. Un it subclasses define attributes that link one Unit to its related Unit instances (e.g. the visit that a sensor belongs to, and vice versa).

The set of **all** Units that may be used to define a Dataset is shown in the the diagram below. Some of these classes (with italics) are themselves abstract, and need to be subclassed in an obs package to create a concrete Unit class. Only the TractUnit and PatchUnit classes are used directly by all cameras. Datasets and SuperTasks may interact only with the classes defined here, however; per-camera specializations will be used only to make globally-unique identifiers and to map Units to the per-camera keys we use in data IDs today. It is, of course, quite likely that I've actually missed one or two important kinds of Unit, but it's highly unlikely that number is > 5, and the important point is that the set of all Unit types is itself an important part of this abstraction.

Some Units inherit from the intermediate base class SpatialUnit, which represents a Unit that occupies a specific area on the sky. It holds a conservative (inclusive) approximation of that region as an attribute (probably something like a sphgeom.RangeSet in a particular pixelization), and a dicti onary linking it to all of the SpatialUnits whose regions overlap its own. Like the other link attributes, this dict may be empty or have some possible relationships missing.

The diagram below only attempts to show a few examples of concrete Datasets. Unlike the set of all Unit types, the set of all Dataset types is very much dynamic: some Datasets may be defined only when a SuperTask is configured; the input and output datasets of a SuperTask will typically be defined by a special pex_config field that has a user-configurable name and a SuperTask-defined set of units. It may even be better to not define a separate derived type for every Dataset, and instead have a single flexible data structure with e.g. __getattr__ overloading for attributes that allow it to be used for any Dataset; I consider that an implementation detail. I think it's easier for the conceptual discussion to think of them as classes.



DatasetGraph and the new SuperTask signatures

Because of the links between them, a set of related Unit and Dataset instances naturally forms a graph-like structure. We define the DatasetGraph class as a top-level entry point to that graph:

class DatasetGraph:

```
def __init__(self):
self.datasets = {} # dict of {name: set<Dataset>}
self.units = {} # dict of {name: set<Unit>}
```

A DatasetGraph could be used to represent all of the content in a DataRepository (including its parent repositories), but it can also be bounded in two ways:

- It may not contain all possible Units (which would require some link attributes on included Units to be None).
- It may not contain all possible Datasets (which just makes some set attributes smaller).

However, a well-formed DatasetGraph may not have any Datasets with Unit attributes set to None.

A DatasetGraph can also be used to represent a superset of the contents of a DataRepository: new datasets can be added to represent outputs that have not yet been generated.

As a result, the right DatasetGraph represents everything defineQuanta needs: given a DatasetGraph that is bounded to the intersection of the input and output data ID expressions, defineQuanta can iterate over its input datasets and add instances of its output dataset to the graph, using the link attributes to form groups between related datasets and build Quanta. This also works for a collection of SuperTasks in a SuperTaskComposite: we can iterate forward (i.e. the same direction as processing flow) through the SuperTasks, adding new output datasets. Because the DatasetGraph is bounded by both the input and output data ID expressions, no Quanta for unnecessary outputs will be generated. We thus propose the following signatures for SuperTask's core methods:

```
class Quantum:
   def __init__(self, inputs, outputs):
        ""Construct from dicts of input and output datasets.
       Parameters
        _ _ _ _ _ _ _ _ _ _ _ _
       inputs : dict of {name: set<Dataset>}
           All inputs used by a single invocation of a SuperTask.
       outputs : dict of {name: set<Dataset>}
           All outputs produced by a single invocation of a SuperTask.
            . . .
       self.inputs = inputs
       self.outputs = outputs
class SuperTask:
   def defineQuanta(self, datasets):
        ""Return a list of Quantum objects representing the inputs and outputs
       of a single invocation of runQuantum().
       Parameters
        _____
       datasets : DatasetGraph
           A DatasetGraph containing only Units matching the data ID
           expression supplied by the user and Datasets that should be
           present in the repository when all previous SuperTasks in the same
           pipeline have been run. Any Datasets produced by this SuperTask
           should be added to the graph on return.
        . . .
       raise NotImplentedError()
   def runQuantum(self, guantum, butler):
        """Run the SuperTask on the inputs and outputs defined by the given
       Quantum, retrieving inputs and writing outputs using a Butler.
       raise NotImplentedError()
   def getDatasetClasses(self):
       """Return a dict containing all of the concrete Dataset classes used
       by this SuperTask.
        ....
       result = \{\}
        for fieldName in self.config:
           cls = getattr(self.ConfigClass, fieldName)
           if subclass(cls, DatasetField):
               p = getattr(self.config, fieldName)
                result[p.name] = p.type
       return result
   def getUnitClasses(self):
        """Return a dict containing all of the concrete Unit classes used
       by this SuperTask.
        . . .
       result = {}
       for DatasetClass in self.getDatasetClasses().values():
            for UnitClass in DatasetClass.UnitClasses:
               result[UnitClass.name] = UnitClass
       return result
```

The getDatasetClasses and getUnitClasses methods provide introspection into the Dataset and Unit types used by the SuperTask. The implementations above assume a DatasetField (inherits from pex.config.Field) is used in the SuperTask's config class to define each of its input and output datasets. A sketch of DatasetField itself is attached (but it may require a familiarity with pex.config and Python metaprogramming with descriptors to understand).

These interfaces have a few more advantages over the baseline interface in terms of what they demand from Butler:

- We do not require a Butler at all in defineQuanta, which means it is not strictly necessary for each activator to be able to construct a "pre-
- flight Butler" at all when the pre-flight environment is significantly different from the job execution environment.
- The only methods we need on a job execution Butler are get(dataset) and put(value, dataset).

Given a SuperTaskComposite, a PreFlightActivator will be able to inspect the constituent SuperTasks to build a list of all Dataset types used by the composite, and from that, it can build a list of all Units used by the composite. This provides one kind of bound on the DatasetGraph: the set of U nit types and and Dataset types that will be included (and by extension, which Unit link attributes will be permitted to be None). The much harder bound is the one values of the actual Unit instances present, which must be formed by intersecting the input and output data ID expressions. That was in many respects the hardest problem in the old design, and it remains the hardest problem in this one; all we have accomplished thus far is redefining it as a handshake between the Unit hierarchy and the Butler, rather than a handshake between the SuperTaskComposite and the Butler. But this is not an insignificant improvement, because the Unit hierarchy contains ~8 explicit relevant classes with well-defined relationships and the set of possible Supe rTasks is (like the set of Datasets) explicitly dynamic.