TheCannon Documentation

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CHAPTER

ONE

INTRODUCTION

This is the software package used for *The Cannon*, a data-driven approach to determining stellar parameters and abundances (hereafter referred to as *labels*) corresponding to a vast set of stellar spectra.

A brief overview of *The Cannon* is below. For more details, see our publication! :D

For *The Cannon* to work, this vast set must have a subset of *reference objects*: spectra that already have high-fidelity labels. Labels for the remaining spectra are determined via a process called *label transfer*.

There are two fundamental steps in label transfer.

- 1. The *training step*: the reference objects are assembled into a *training set* that is used to solve for a flexible generative model (with ~80,000 parameters). The model describes the flux in every pixel of the continuum-normalized spectrum as a function of stellar labels.
- 2. The *test step*: we assume that this generative model holds for all of the other objects in the survey (dubbed *survey objects*). Then, the spectra of the survey objects and the generative model from the reference objects allow us to solve for or infer the labels of the survey objects.

A few notable features of this package are:

- Helpful (optional) methods for preparing spectra and labels to be fed into *The Cannon*
- Generate a training set and a test set by supplying spectra and training labels
- Merge multiple training sets into one
- Generate a model by running the training set through Step 1 of *The Cannon*
- Determine labels for the test set by running the model and training set through Step 2 of *The Cannon*.

CHAPTER

TWO

GETTING STARTED

The basic workflow for creating a training set and test set, solving for the model (Step 1) and finally determining labels (Step 2) is shown below.

2.1 Step 1: Prepare Data

The Cannon expects all spectra - for reference and survey objects - to be continuum-normalized in a consistent way, and sampled on a consistent rest-frame wavelength grid, with the same line-spread function. It also assumes that the flux variance, from photon noise and other sources, is known at each spectral pixel of each spectrum.

The user is left to prepare the data him or herself, but we include some optional methods that may be helpful.

2.2 Step 2: Construct Training Set

The training set is a set of stars from the survey under consideration for which the user has spectra and also high-fidelity labels (that is, stellar parameters and element abundances that are deemed both accurate and precise.) The set of reference objects is critical, as the label transfer to the survey objects can only be as good as the quality of the training set.

Labels for a training set will not necessarily come from the same data, and therefore spectra will not necessarily be in the same data format. For flexibility, *The Cannon* software package allows the user to construct training subsets, one data type each, and then merge the subsets.

```
>>> dataset import Dataset
>>> fts_trainingset = Dataset(objectIDs = [], spectra = [], labelnames = [], labelvals = [])
>>> vesta_trainingset = Dataset(objectIDs = [], spectra = [], labelnames = [], labelvals = [])
>>> cluster_trainingset = Dataset(objectIDs = [], spectra = [], labelnames = [], labelvals = [])
>>> trainingset = mergesets(fts_trainingset, vesta_trainingset, cluster_trainingset)
```

There are a few ways to examine the dataset. You can retrieve the spectra as follows:

```
>>> pixels = trainingset.spectra[:,:,0]
>>> fluxes = trainingset.spectra[:,:,1]
>>> fluxerrs = trainingset.spectra[:,:,2]
```

2.3 Step 3: Construct Test Set

```
>>> testset = Dataset(objectIDs = [], spectra = [], labelnames = [], labelvals = None)
```

2.4 Step 4: The Cannon Step 1 - Generate Model

```
>>> from spectral_model import SpectralModel
>>> model = SpectralModel(label_names, modeltype)
>>> model.train(trainingset)
```

2.5 Step 5: The Cannon Step 2 - Infer Labels

```
>>> from cannon_labels import CannonLabels
>>> labels = CannonLabels(label_names)
>>> labels.solve(model, testset)
```

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