

DarwinTM Python SDK Manual

A SparkCognitionTM Education Document

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About this guide

This manual describes using the DarwinTM SDK to access and use the Darwin API in automated model building. It is intended for data scientists, software engineers and analysts who want to use the Darwin API to interact with Darwin to create and train models, monitor jobs and perform analysis. The SDK also provides some convenience functions. Note that throughout this document, long key and token values are truncated, indicated by ellipses (...).

Expectation

This document assumes experience of the data scientist or software engineer that is commensurate with data science techniques and associated programming tasks.

Darwin overview

Darwin is a SparkCognition® tool that automates model building processes to solve specific problems. This tool enhances data scientist potential because it automates various tasks that are often manually performed. These tasks include data cleaning, latent relationship extraction, and optimal model determination. Darwin promotes rapid and accurate feature generation through both automated windowing and risk generation. Darwin quickly creates highly-accurate, dynamic models using both supervised and unsupervised learning methods.

The general workflow for simple modeling includes:

- Upload training data
- Create model
- · Upload test data
- Test the model

· Download result artifact

Note: Darwin expects all uploaded ingestion files to be in a *rectangular* format. This means a flat file with features that span columns and data samples that span rows. Plan your data file so it fits this expectation to help prevent errors.

See the **SDK** example for a modeling example.

For additional information on Darwin, contact your local SparkCognition partner for access to the white paper titled: Darwin - A Neurogenesis Platform.

Accessing the API

This document describes the SDK and explains how to access the Darwin API and its functionality. Additional methods to access the Darwin API include:

- through the https://darwin-api.sparkcognition.com/v1 end point
- optionally, through user created curl commands

For additional information on the Darwin API, contact your local SparkCognition partner for access to see the SparkCognition Darwin API Guide.

Notes:

- An API key is necessary to use the Darwin SDK.
 Contact SparkCognition or your IT manager for an appropriate key.
- All methods return a 2-tuple, for example:

```
(True, <context-dependent-return-object>)
(False, <some-helpful-message>)
```

Darwin SDK interface

Setup Darwin SDK

Perform the following to download and setup the Darwin SDK:

- 1. Install Python 3.5 or greater. Alternatively, install Miniconda, from https://conda.io/miniconda.html.
- 2. Create a directory to receive the git repository clone.
- 3. Change (cd) into the new directory.
- 4. Clone the *darwin-sdk* repository:

```
git clone https://github.com/sparkcognition/darwin-sdk
```

5. Change into the new root directory of the darwin-sdk cloned darwin-sdk project:

```
cd <NewCloneRootDirectory>
```

Note: By default this is the master trunk .

6. Download the code:

git pull

7. Setup the SDK:

python setup.py install

The SDK defaults to using the production URL: https://darwin-api.sparkcognition.com/v1/

Connect to the Darwin interface

1. Obtain an api key.

To use the Darwin SDK, an API Key is required. A key can be obtained from SparkCognition support or your IT manager. An api_key is a long string, for example:

```
"RsJ74ZS5AvwznbHh0AfVSgrchhS9KxACDy3jefaQnxb9f6QTSDBFmhnGa0cOIWtNAIFRAG9ToOTpi0mnEo3zFA" \\
```

2. Register the api key.

The purpose of this method is to set a password for an api_key. Each api_key is synonymous with a service. This method must be invoked once for each api_key to establish a password for that key.

Notes:

- After successful registration, the service uses auth_login() to login as a service.
- In version 1, the password cannot be changed.

```
>>> from amb_sdk.sdk import DarwinSdk
>>> s = DarwinSdk()
>>> s.auth_register('asdf', 'iRgwut4kGs0ymULiuKtMd0WFvBYLMWSj16q2uysQeteqH9ssc+\
EETUvcysnPojRpfycLVHa2IlN1IlrfEk1YMA')
(True,'Bearer eyJ0eXAiOiJKV1QiLCJhbGciOiJIUzI1NiJ9.eyJleHAiOjE1MTU1MzM4NjEsImlh\
dCI6MTUxNTUzMDI2MS ... F56xZQiBT-89nrRz1nIXD5LfawHIj_MlUHQqM36vU')
```

3. Login.

Login as a service or create a user under the service and login as a user.

The following explains how to log in as a service.

Notes:

- o Although Bearer <auth-token>, returned by auth_login(), is used in subsequent calls to validate authenticity, it is not required for each method.
- The SDK remembers the auth token for the DarwinSdk object. Although an auth token is currently valid for 1 hour, if the DarwinSdk object session life time exceeds 1 hour, the SDK will request another auth token until the session ends.

Example

```
>>> s.auth_login('asdf', 'iRgwut4kGs0ymULiuKtMd0WFvBYLMWSj16q2uysQeteqH9ssc+EET\
UvcysnPojRpfycLVHaZIlNIIlrfEk1YMA')
(True,'Bearer eyJ0eXAi0iJKV1QiLCJhbGci0iJIUzI1NiJ9.eyJleHAi0jE1MTU1MzQxNzIsImlh\
dCI6MTUxNTUzMD ... UQQfoXqYFkJSoRXXDNPE985-a08cE6_o')
```

When login (as a service) successfully completes, the SDK can be used to create and model a workflow.

Note, there are also auth_register_user() and auth_login_user() methods that allow you to create users and login as a specific user. You can choose to use the SDK as a service or create users underneath the service to partition datasets/models to be owned by specific users. It is more convenient to employ user accounts because the api_key is not necessary for logging in as a user.

Example

4. Verify the connection.

The default url in the SDK is https://darwin-api.sparkcognition.com/v1/. Use get_url() and set_url() to verify connection to the right Darwin service.

Darwin SDK methods

URL Get/Set methods

 $DarwinSdk. \boldsymbol{get_url}()$

Get Darwin service url.

Parameters: None

Returns:

(True, <url-string>)

Example

```
In [10]: s.get_url()
Out[10]: (True, 'https://darwin-api.sparkcognition.com/v1/')
```

DarwinSdk.set_url(url, version='v1')

Set Darwin service url and version.

Parameters:

- url URL to the Darwin service
- version (optional) defaults to 'v1'

Returns:

```
(True, \url>) or (False, 'invalid url')
```

Example

```
In [9]: s.set_url('https://darwin-api.sparkcognition.com/v1/')
Out[9]: (True, 'https://darwin-api.sparkcognition.com/v1/')
```

Authentication methods

DarwinSdk.auth_register(password, api key)

Register as a service. The purpose of this method is to set a password for an api_key. Each api_key is synonymous with a service. This method is invoked only once for each api_key to establish a password for that key. After registration, the service can use *auth_login()* to login as a service.

Note: In version 1, the password cannot be changed.

Parameters:

- password The service level password
- api_key The api key for the service

Returns:

```
(True, 'Bearer <auth-token>') or (False, <error-message>)
```

Bearer <auth-token> is used in subsequent calls to validate authenticity.

The SDK remembers the auth token for the AmkSdk object.

Note: Although an auth token is currently valid for 1 hour, if the AmkSdk object session life time exceeds 1 hour, the SDK will request another auth token until the session ends.

Example

```
In [4]: s.auth_register('asdf', 'RsJ74ZS5AvwznbHh0AfVSgrchhS9KxACDy\
3jefaQnxb9f6QTSDBFmhnGa0cOIWtNAIFRAG9ToOTpi0mnEo3zFA')
Out[4]:
(True,
    'Bearer eyJhbGci0iJIUzI1NiIsInR5cCI6IkpXVCJ9.eyJleHAi0...iSdU8xlF4yJk')
```

DarwinSdk.auth_login(password, api key)

Login as a service.

Note: A service must have a password set using auth register() to login successfully.

Parameters:

- password The service level password
- api_key The api key for the service

Returns:

```
(True, 'Bearer <auth-token>') or (False, <error-message>)
```

Bearer \auth-token> is used in subsequent calls to validate authenticity. The SDK remembers the auth token for the AmkSdk object.

Note: Although an auth token is currently valid for 1 hour, if the AmkSdk object session life time exceeds the 1 hour limit, the SDK will acquire another auth token until the session ends.

```
In [5]: s.auth_login('asdf',
   'iRgwut4kGs0ymULiuKtMd0WFvBYLMWSj16q2uysQeteqH9ssc+EETUvcysnPojRpfyc\
LVHa2IINIIlrfEk1YMA')
Out[5]:
(True,
   'Bearer
```

```
eyJ0eXAiOiJKV1QiLCJhbGciOiJIUzI1NiJ9.eyJleHAiOjE1MTU1MzQxN....')
```

DarwinSdk.auth_register_user(username, password)

Register a user. This method registers a new user.

Note: You must be logged in as a service to create a user.

Parameters:

- username The new end user's username
- password new end user's password

Returns:

```
(True, 'Bearer <auth-token>') or (False, <error-message>)
```

Bearer (auth-token) is used in subsequent calls to validate authenticity. The SDK remembers the auth token for the AmkSdk object.

Note: Although an auth token is currently valid for 1 hour, if the AmkSdk object session life time exceeds the 1 hour limit, the SDK will acquire another auth token until the session ends.

Example

```
In [8]: s.auth_register_user('user1', 'user1-password',
'iRgwut4kGs0ymULiuKtMd0WFvBYLMWSj16q2uysQeteqH9ssc+EETUvcysnPojRpfycLV\
Ha2IlN1IlrfEk1YMA')
Out[8]:
(True,
'Bearer
eyJ0eXAiOiJKV1QiLCJhbGciOiJIUzI1NiJ9.eyJleHAiOjE1MTU1MzQyN....')
```

DarwinSdk.auth_login_user(username, password)

Login as a user.

Note: A user must have a username and password set using auth_register_user() to successfully login.

Parameters:

- username The end user's username
- password The service level password

Returns:

```
(True, 'Bearer <auth-token>') or (False, <error-message>)
```

Bearer <auth-token> is used in subsequent calls to validate authenticity. The SDK remembers the auth token for the AmkSdk object.

Note: Although an auth token is currently valid for 1 hour, if the AmkSdk object session life time exceeds the 1 hour limit, the SDK will acquire another auth token until the session ends.

Example

```
In [9]: s.auth_login_user('user1', 'user1-password',
'iRgwut4kGs0ymULiuKtMd0WFvBYLMWSj16q2uysQeteqH9ssc+EETUvcysnPojRpfycLV\
Ha2IlN1IlrfEk1YMA')
Out[9]:
(True,
'Bearer
eyJ0eXAiOiJKV1QiLCJhbGciOiJIUzI1NiJ9.eyJleHAiOjE1MTU1MzQzM....')
```

Job status methods

DarwinSdk.lookup job status(age=None, status=None)

Get status information for all jobs belonging to the current user or service.

Parameters:

• age - (optional) Filter jobs that are less than X units old, for example 3w, 2d, or 1h.

- Optional parameters:
 - o status If not specified, returns all jobs.
 - o running
 - requested
 - o complete
 - failed

Returns:

```
(True, <list-of-jobs>) or (False, <error-message>)
```

Example

```
In [6]: s.lookup_job_status(status='Complete')
Out[6]:
(True,
   [{'artifact_names': None,
        'dataset_names': ['cancer-train'],
        'endtime': '2018-02-01T10:53:50.451598',
        'generations': 0,
        'job_name': 'eeef500d629e4a2185eb8af6e18a83b4',
        'job_type': 'TrainModel',
        'loss': 2.0,
        'model_name': 'cancer-model',
        'percent_complete': 100,
        'starttime': '2018-02-01T10:52:42.280929',
        'status': 'Complete'}])
```

 $DarwinSdk. \textbf{lookup_job_status_name}(job_name)$

Get job status information for a job by name.

Parameters:

• *job_name* - The job name.

Returns:

```
(True, <job-info>) or (False, <error-message>)
```

Example

Lookup methods

DarwinSdk.lookup_artifact(type=None)

Get a list of artifacts belonging to the current user or service.

Parameters:

 $type\ -\ (optional)\ specifies\ the\ type\ of\ artifact.\ Values\ can\ be\ 'Model',\ 'Test',\ 'Dataset',\ 'Risk',\ 'Run',\ '$

Returns:

```
(True, <job-info>) or (False, <error-message>)
```

```
In [30]: s.lookup_artifact('Run')
http://localhost:5000/v1/lookup/artifact
Out[30]:
(True,
   [{'created_at': '2018-02-01T11:09:55.731040',
```

```
'id': 'b9a9205a-0772-11e8-a003-3b1c8766dad0',

'mbytes': 0.0,

'name': '8a63e21030d1483abb0f892963c1728f',

'type': 'Run'},

{'created_at': '2018-02-01T11:11:17.560360',

'id': 'ea6f3f80-0772-11e8-9abe-77bc32e350c5',

'mbytes': 0.0,

'name': 'artifact-1',

'type': 'Run'}]
```

DarwinSdk.lookup_artifact_name(artifact_name)

Get information for an artifact specified by its name.

Parameters:

· artifact - specifies an artifact by its name

Returns:

```
(True, <job-info>) or (False, <error-message>)
```

Example:

DarwinSdk.lookup_client()

Get a client's metadata. A client is the current user or service in context.

Parameters: None

Returns:

```
(True, <client-info>) or (False, <error-message>)
```

Example

 $DarwinSdk. \textbf{lookup_dataset}()$

Get the dataset(s) metadata for a user. The user is the current user or service in the current context. This is useful for listing all created datasets.

Parameters: None

Returns:

```
(True, t-of-dataset-info>) or (False, <error-message>)
```

```
In [4]: s.lookup_dataset()
Out[4]:
(True,
   [{'categorical': None,
        'imbalanced': None,
        'mbytes': 0.02019977569580078,
        'name': 'unittest-cancer-dataset2',
        'sequential': None,
        'updated_at': '2018-01-31T15:37:28.310994'},
        {'categorical': None,
        'imbalanced': None,
```

```
'mbytes': 0.02019977569580078,
'name': 'cancer-train',
'sequential': None,
'updated_at': '2018-02-01T10:52:06.076279'}])
```

 $DarwinSdk. \textbf{lookup_dataset_name}(dataset_name)$

Get a specific dataset's metadata.

Parameters:

• dataset_name - The name of the dataset. The name of a dataset is established in the upload_dataset() method.

Returns

```
(True, <dataset-info>) or (False, <error-message>)
```

Example

$DarwinSdk. \textbf{lookup_model}()$

Get the model(s) metadata for a user. The user is the current user or service in the current context. This is useful for listing all models.

Parameters: None

Returns:

```
(True, t-of-model-info>) or (False, <error-message>)
```

Example

```
In [37]: s.lookup_model()
Out[37]:
(True,
  [{'generations': 0,
    'loss': 2.0,
    'name': 'cancer-model',
    'parameters': {'target': 'Diagnosis'},
    'trained_on': ['cancer-train'],
    'updated_at': '2018-02-01T10:53:50.443166'}])
```

DarwinSdk.lookup_model_name(model_name)

Get a specific model's metadata. The name of a model is established in the create_model() method.

Parameters:

• model_name - The name of the model

Returns:

```
(True, <model-info>) or (False, <error-message>)
```

Example

DarwinSdk.lookup_tier()

Get metadata for all tiers. A tier specifies certain usage limits such as number of models and datasets.

Parameters: None

Returns:

```
(True, t-of-tier-info>) or (False, <error-message>)
```

Example

DarwinSdk.lookup_tier_num(tier_num)

Get a specific tier's metadata. A tier specifies certain usage limits such as the number of models or datasets.

Parameters:

• tier_num - The name of the model

Returns:

```
(True, <tier-info>) or (False, <error-message>)
```

Example

Datasets and artifact methods

DarwinSdk.upload_dataset(dataset, dataset name=None)

Upload a dataset, model, or a figure.

Parameters:

- dataset- Path to dataset
- dataset_name Name to be given to dataset, or defaults to filename

Returns:

```
(True, {dataset_name:<name-given-to-dataset>}) or (False, <error-message>)
```

Example

DarwinSdk.delete_dataset(dataset_name)

Delete the named dataset.

Parameters:

• dataset name - The name of the dataset to be deleted.

Returns:

```
(True, None) or (False, <error-message>)
```

Example

```
In [6]: s.delete_dataset('unittest-cancer-dataset')
Out[6]:
(True, None)
```

DarwinSdk.download_artifact(artifact name)

Download artifact given its name. The methods that return artifacts are:

- analyze_data()
- analyze model()
- run_model()
- create risk info()

Note: The artifact for *analyze_model()* is a png file.

Parameters:

• artifact name - The name of the artifact to download.

Returns:

```
(True, <path-to-file>) or (True, <python-list>) or (False, <error-message>)
```

Example run_model() or prediction artifact

Example analyze_data() artifact

```
In [17]: c, r = s.download_artifact('03676cb98d5b48fc89b2d058022781d8')
In [18]: print(r)
                  col_name
  categories
                              col_type
                                             max
                                                     mean
                                                                min \
                              numerical 180.219 95.3869
                                                             5.8125
        None mean_profile
               std_profile
                                         91.8086
                                                  44.3984
                                                             24.772
                              numerical
        None
              kurt_profile
                                                  1.24083
                                                           -1.60483
        None
                              numerical
                                         8.06952
             skew_profile
                              numerical
                                         68.1016
                                                  5.72565
                                                           -1.78189
        None
4
        None
               mean_dmsnr
                              numerical
                                         222.421
                                                  23.786 0.273411
        None
                std_dmsnr
                              numerical
                                         110.642
                                                   35,272
                                                           7.56568
                                         32.1986
                                                  6.66098
               kurt dmsnr
6
        None
                              numerical
                                                           -3.13927
                                                  79.4015
                                                           -1.97698
                skew_dmsnr
                              numerical
                                         1072.96
        None
                    _
class categorical
                                            None
                                                     None
         0,1
                                                               None
   num_categories
           3939.0 36.4588
1
           4635.0
                   8.03585
           4639.0
                   1.80623
2
3
4
           4639.0
                  11.0723
           4012.0
                   39.2822
           4639.0 23.9892
6
           4639.0
                   4.87737
7
8
           4639.0
                   103.01
              2.0
                      None
```

Example analyze model() or prediction artifact

Example create risk info() artifact

DarwinSdk.delete artifact(artifact name)

Delete the artifact given its name.

Parameters:

• artifact name - The name of the artifact to be deleted.

Returns:

```
(True, None) or (False, <error-message>)
```

Example

```
In [8]: s.delete_artifact('6c482eac9f894cdb9b0e1e487e41730a')
Out[8]:
(True, None)
```

Modeling and analysis methods

DarwinSdk.create_model(dataset_names, **kwargs)

Create a model trained on the dataset identified by dataset_name. The name of a model is specified in a parameter in kwargs. **Note**: If no name is specified, the model is named with a *uuid-like* name.

Parameters:

- · dataset_names A single dataset name as a string or a list of dataset string names to be used for training
- **kwargs variable number of keyword arguments, described in parameters.
- parameters
 - o model_name: The string identifier of the model to be trained. If no name is specified, the model is named with a uuid-like name.
 - o job_name: If no name is specified, the job is named with a uuid-like name.
 - o target: String denoting target prediction column in input data.
 - o Model_name: The string identifier of the model to be trained.
 - max_train_time: Sets the training time for the model in 'HH:MM' format.
 Note: This overrides any values set for max_generation.
 - o max_generation: Expected input/type: numeric. Sets the training time for the model in generations. If max_train_time is set, this parameter is ignored.
 - recurrent: Expected input/type: true/false. Enables recurrent connections to be evolved in the model. This option can be useful for timeseries or sequential data.
 - Note: This option is automatically enabled if a datetime column is detected in the input data. This can result in slower model evolution.
 - o impute: String alias that indicates how to fill in missing values in input data.

ALIAS	DESCRIPTION	COMPLEXITY
'genetic'	Genetic Fill: Automatically determines the most appropriate fast imputation method using evolutionary methods.	Linear
'ffill'		Linear Fast
'bfill'	Backward Fill: Propagate values backward from one example into the missing cell of the previous example. Might be useful for timeseries data, but also applicable for both numerical and categorical data.	Linear Fast

ALIAS	DESCRIPTION	COMPLEXITY
'mean'	Mean Fill: Computes the mean value of all non-missing examples in a column to fill in missing examples. The result may or might not be interpretable in terms of the input space for categorical variables.	Linear Fast
'median'	Median Fill: Computes the median value of all non-missing examples in a column to fill in missing examples. While the result is interpretable in terms of the input space for categorical variables, the approach might not be appropriate for non-ordinal data.	Linear Fast
'mode'	Mode Fill: Uses the most common value on a column-by-column basis to fill in missing examples. The result is interpretable for both numerical and categorical variables.	Linear Fast
'spline'	Spline Fill: Interpolation using a spline (piecewise function). Might be useful for timeseries or sequential data.	Linear Fast
'Linear'	Linear Interpolation Fill: Interpolation using a Linear function. Might be useful for timeseries or sequential data.	Linear Fast
'knn'	K-Nearest Neighbors Fill: Fills in missing values by averaging the cell values of the k nearest neighbors in the reduced feature space defined by all non-missing columns.	Polynomial Slow
'rmf'	Robust Matrix Factorization Fill: Computes low-rank matrices L (observations x rank), R (features x rank), and E where X is input data, and $X = LR^T + E$.	Polynomial Slow
'mice'	Multiple Imputation by Chained Equations: First imputes missing values using <i>Forward Fill</i> . Then, column-by-column, missing values are reintroduced and regressed upon using the other (non-missing) columns. Continues iteratively.	Polynomial, Iterative Very Slow

drop: Expected input/type: true/false. Enables automatic pruning of input columns based on different criteria such as amount of missing data, number of unique values, and standard deviation.

Note: This automatically drops identifier columns (unique value for each sample) and columns that do not contain sufficient data to aid prediction.

- max_int_uniques: Expected input/type: integer. Threshold for automatic encoding of categorical variables. If a column contains at least max_int_uniques unique values, it is treated as categorical and one hot encoded during preprocessing.
- max_unique_values: Expected input/type: integer. Threshold for automatic pruning of categorical columns prior to one hot encoding based on the number of unique values.

Note: If a categorical column contains at least max unique values, it is dropped during preprocessing prior to one hot encoding.

 feature_eng: Enables automatic feature generation. Identifies an appropriate time window and augments input with new features derived in the frequency and time domains.

Note: Can only be applied to timeseries data. String aliases specify methods for window computation.

ALIAS	DESCRIPTION
None	No feature generation will be applied.
ʻmi'	Uses mutual information to estimate the window length.
'auc'	(Default) Uses autocorrelation to estimate the window length.
'user'	User specified window length: see* window_len*.

window len: Expected input/type: integer. User specified window length for feature generation.
 Note: This parameter is used only in the case that user is provided for the feature eng parameter.

- feature_select: A number in [0,1] specifying the percentage of numerical features to maintain based on their dependency to the target. Ranks all features using mutual information and drops (1 feature_select)% of the lowest-ranking features. Default is 1 (keep all features).
- outlier: A string alias that indicates the outlier detection to apply during preprocessing.
 Note: Outliers are removed and later filled using imputation.

ALIAS	DESCRIPTION
None	(Default) No outlier detection will be applied.
'mad'	Uses Median Absolute Deviation to detect outliers.
'perc'	Uses Percentile-based outlier detection.
'isol'	Uses an Isolation Forest to detect outliers.

 auto_save_per (supervised only): Expected input/type: integer. Sets the checkpoint frequency. The model creation progress is recorded after every auto_save_per generations.

Note: If the model is retrained, the model begins from the last recorded checkpoint. The model is automatically saved at the end of evolution.

- imbalance (supervised only): Expected input/type: true/false. Enables automatic imbalance correction that selectively applies random oversampling, random undersampling, synthetic minority oversampling (SMOTE), or adaptive synthetic sampling (ADASYN) to the input data depending on problem characteristics.
- o clustering: Expected input/type: true/false. Enables clustering for unsupervised problems. If false, detects outliers.
- n_clusters: Expected input/type: integer. Specifies the number of clusters to be used if clustering is enabled.
 Note: If this value is not provided, the number of clusters will be heuristically determined.
- anomaly_prior: Expected input/type: *between [0,1]. *Significance level at which a point is defined as anomalous.
 Note: This parameter is used only for unsupervised problems if clustering is disabled.

Returns:

```
(True, {'job_id': <uuid1>, model_name: <model_name>}) or (False, <error-message>)
```

DarwinSdk.delete model(model name)

Delete a model named by model name.

Parameters:

o model name - Name of the model to be deleted.

Returns:

```
(True, None) or (False, <error-message>)
```

Example

```
In [5]: s.delete_model('unittest-cancer-model')
Out[5]: (True, None)
```

DarwinSdk.resume_training_model(model_name, dataset_names, **kwargs)

Resume training for a model on the dataset(s) identified by dataset_names.

Parameters:

- o dataset name- Name of dataset(s) used to train.
- o model name Name of the model to train.
- o **kwargs variable number of keyword arguments, described below:.
 - job name If not specified, a uuid is created as the job_name.
 - max_train_time If not specified, the default is used.

Returns:

```
(True, { "job_id""<uuid>", "model_name": "<model_name>" }) or (False, <error-message>)
```

Example

```
In [8]: s.resume_training_model('unittest-cancer-model', 'unittest-cancer-\
dataset', target="Diagnosis", max_train_time="00:01")
Out[8]:
(True, {"job_id": "4e59ffc425e047e1a3b872f1e7396976", "model_name": "unittest-\
cancer-model"})
```

DarwinSdk.analyze_data(dataset name, **kwargs)

Analyze the dataset given its name.

Parameters:

- $\circ \ \ \textit{dataset_name}$ The name of the dataset to be analyzed.
- $\circ \ \ **kwargs$ variable number of keyword arguments, described below:
 - job_name (optional) If not specified, a unid will be created as the job_name.
 artifact name: (optional) If not specified, a unid will be created as the artifact name.
 - target: String denoting target prediction column in input data.
 - impute: String alias that indicates how to fill in missing values in input data.

ALIAS	DESCRIPTION	COMPLEXITY
'genetic'	Genetic Fill: Automatically determines the most appropriate fast imputation method using evolutionary methods.	Linear
'ffill'	(Default) Forward Fill: Propagate values forward from one example into the missing cell of the next example. Might be useful for timeseries data, but also applicable for both numerical and categorical data.	Linear Fast
ʻbfill'	Backward Fill: Propagate values backward from one example into the missing cell of the previous example. Might be useful for timeseries data, but also applicable for both numerical and categorical data.	Linear Fast

ALIAS	DESCRIPTION	COMPLEXITY
'mean'	Mean Fill: Computes the mean value of all non-missing examples in a column to fill in missing examples. The result may or might not be interpretable in terms of the input space for categorical variables.	Linear Fast
'median'	levamples. While the result is interpretable in terms of the input space for categorical variables, the	Linear Fast
'mode'	Mode Fill: Uses the most common value on a column-by-column basis to fill in missing examples. The result is interpretable for both numerical and categorical variables.	Linear Fast
'spline'	Spline Fill: Interpolation using a spline (piecewise function). Might be useful for timeseries or sequential data.	Linear Fast
'Linear'	Linear Interpolation Fill: Interpolation using a Linear function. Might be useful for timeseries or sequential data.	Linear Fast
	K-Nearest Neighbors Fill: Fills in missing values by averaging the cell values of the k nearest neighbors in the reduced feature space defined by all non-missing columns.	Polynomial Slow
'rmf"	Robust Matrix Factorization Fill: Computes low-rank matrices L (observations x rank), R (features x rank), and E where X is input data, and $X = LR^T + E$.	Polynomial Slow
'mice'	Multiple Imputation by Chained Equations: First imputes missing values using Forward Fill. Then, column-by-column, missing values are reintroduced and regressed upon using the other (non-missing) columns. Continues iteratively.	Polynomial, Iterative Very Slow

- drop: Expected input/type: true/false. Enables automatic pruning of input columns based on different criteria such as amount of missing data, number of unique values, and standard deviation.
 - Note: This automatically drops identifier columns (unique value for each sample) and columns that do not contain sufficient data to aid prediction.
- max_int_uniques: Expected input/type: integer. Threshold for automatic encoding of categorical variables. If a column contains at least max_int_uniques unique values, it is treated as categorical and one hot encoded during preprocessing.
- max_unique_values: Expected input/type: integer. Threshold for automatic pruning of categorical columns prior to one hot encoding based on the number of unique values.
 - Note: If a categorical column contains at least max_unique_values, it is dropped during preprocessing prior to one hot encoding.
- feature_eng: Enables automatic feature generation. Identifies an appropriate time window and augments input with new features derived in the frequency and time domains.

Note: Can only be applied to timeseries data. String aliases specify methods for window computation.

ALIAS	DESCRIPTION
None	No feature generation will be applied.
ʻmi'	Uses mutual information to estimate the window length.
'auc'	(Default) Uses autocorrelation to estimate the window length.
'user'	User specified window length: see* window_len*.

- window_len: Expected input/type: integer. User specified window length for feature generation.
 Note: This parameter is used only in the case that user is provided for the feature_eng parameter.
- feature_select: A number in [0,1] specifying the percentage of numerical features to maintain based on their dependency to the target. Ranks all features using mutual information and drops (1 feature_select)% of the lowest-ranking features. Default is 1 (keep all features).
- outlier: A string alias that indicates the outlier detection to apply during preprocessing.
 Note: Outliers are removed and later filled using imputation.

Returns:

```
({\tt True, \{"job\_name": <string>, "artifact\_name": <string>}) \quad {\tt or (False, <error-message>)}
```

Example

```
In [10]: s.analyze_data('pulsars', feature_select=1, impute="mean", drop=True, \
target="string", max_int_uniques=15, feature_en="mi", outlier="mad", \
max_unique_values=50)

Out [10]:
(True, {'artifact_id': '2c90c1dc402a4a6da37a139dfc2f7871', \
'job_id': '7871ebb62ad1458da64d800bc73019de'})
```

DarwinSdk.analyze_model(model name, job name=None, artifact name=None)

Analyze the model given its model name.

Parameters:

- model_name The name of the model to be analyzed.
- job_name (optional) If not specified, a unid is created as the job_name.
- artifact_name (optional) If not specified, a unid is created as the artifact_name.

Returns:

```
(True, {"job_name": <string>, "artifact_name": <string>}) or (False, <error-message>)
```

Example

```
In [5]: s.analyze_model('unittest-cancer-model')
Out [5]:
(True, {'artifact_id': '71a8ae55f2934014b45c13a3975f419c', 'job_id': \
    '4e59ffc425e047e1a3b872f1e7396976'})
```

DarwinSdk.run_model(dataset name, model name, supervised=True, job name=None, artifact name=None)

Run the model given its name and a dataset to use. Use upload_dataset() to upload a data set.

Parameters:

- dataset_nameThe name of a dataset to use for running the model.
- model name The name of the model to run.
- supervised (optional) If not specified, assumes a supervised model.
- job name (optional) If not specified, a unid is created as the job name.
- artifact_name (optional) If not specified, a unid is created as the artifact_name.

Returns:

```
(True, {"job_name": <string>, "artifact_name": <string>}) or (False, <error-message>)
```

Example

```
[In [9]: s.run_model('unittest-cancer-testdataset', 'unittest-cancer-model')
Out [9]:
(True, {'artifact_id': '6c482eac9f894cdb9b0e1e487e41730a', 'job_id': \
'1696e03c8165404c8e05685ea68baa3c'})
```

DarwinSdk.create risk info(failure dataset name, timeseries dataset name, **kwargs)

Create risk information given failure and timeseries data. Use upload dataset() to upload datasets.

Notes concerning risk - Risk is a value used in calculating future events. Risk is calculated using algorithms based on sliding time frames and associated historical data that projects forward in time to predict the likelihood of the event. The outcome of the calculations is that the likelihood of an event occurring within a particular time frame becomes available for use. Note that the risk values are dependent on the quality and extent of the historical data as well as the scope of the timeframe used for evaluation.

Parameters:

- failure_dataset_name The name of a failure dataset.
- timeseries_dataset_name The name of a timeseries dataset.
- **kwargs variable number of keyword arguments, described below:
 - job_name (optional) If not specified, a uuid is created as the job_name.
 - artifact name (optional) If not specified, a unid is created as the artifact name.
 - risk_columns: A list of column names in the index.
 - shutdown_column: Name of the column in the risk data that denotes the beginning of the predicted event of interest.
 - return_column: Name of the column in the risk data that denotes the end of the predicted event and when all data can again be considered "normal".
 - asset_column: Name of the asset column in the risk data. This parameter is used when the datasets consist of multiple different assets.
 - lead_time: Lead time in seconds. This value is half width of the risk function that means. the risk index is 0 prior to 2* lead_time and increases to 1 at a failure time.
 - Functional_form: Shape of a risk function, includes:
 - step: Step function
 - linear: Linear function
 - sigmoid: Sigmoid function
 - exponential: Exponential function

Returns:

```
(True, {"job_name": <string>, "artifact_name": <string>}) or (False, <error-message>)
```

```
s.create_risk_info('failure-data', 'timeseries-data', return_column=\
"Date Returned to Service", shutdown_column="Shutdown Date", lead_time=1.0, \
functional_form="linear")
```

Convenience methods

DarwinSdk.delete_all_datasets()

Deletes user datasets. This method deletes all datasets in the current user or service context. **Note**: Use *lookup dataset()* to view/verify the datasets for deletion.

Parameters: None

Returns:

```
(True, None) or (False, <error-message>)
```

DarwinSdk.delete_all_models()

Delete all models for a user. This method will delete all models in the

current user's or service's context.

Note: Use lookup_model() to review and verify that you want to delete all listed models.

Parameters: None

Returns:

```
(True, None) or (False, <error-message>)
```

DarwinSdk.wait_for_job(job name, time limit=600)

Synchronously wait for a job to complete, limited by time limit that defaults to 600 seconds.

Parameters:

- job name The id for the job
- time limit (optional) defaults to 600 seconds

Returns::

```
(True, None) or (False, <error-message>)
```

Reference

- SDK modeling example
- SDK analyze data workflow example
- Revision table

SDK modeling example

The following example shows the Darwin SDK performing a modeling process:

```
'generations': 17, 'status': 'Running', 'starttime': '2018-02-01T14:16:51.4\
'generations': 21, 'status': 'Running', 'starttime': '2018-02-01714:16:51.4\
64827', 'artifact_names': None}
{'percent_complete': 100, 'job_type': 'TrainModel', 'model_name': 'my-model', \
'dataset_names': ['mydata'], 'endtime': '2018-02-01714:18:02.072976', 'loss': \
0.39636287093162537, 'generations': 23, 'status': 'Complete', 'starttime': \
'2018-02-01714:16:51.464827', 'artifact_names': None}
Out[15]: (True, 'Job completed')
In [16]: s.upload_dataset('sets/cancer_test.csv', 'mytestdata')
Out[16]: (True, {'dataset_name': 'mytestdata'})
In [19]: s.run_model('mytestdata', 'my-model')
Out[19]:
 (True,
 {'artifact_name': '9a6d41532cec47618beee6236b02c129',
   'job_name': '91c7813334ee4c37a733761dce71c0b3'})
In [21]: s.wait_for_job('91c7813334ee4c37a733761dce71c0b3')
Out[21]: (True, 'Job completed')
In [22]: s.download_artifact('9a6d41532cec47618beee6236b02c129')
              Diagnosis
           BENIGN
           BENIGN
           BENIGN
 3
           BENTGN
 4
           BENTGN
       MALIGNANT
 6
       MALIGNANT
       MALIGNANT
       MALIGNANT
 [100 rows x 1 columns])
```

SDK analyze data workflow example

The following example shows a Darwin SDK data analysis workflow example:

```
In [1]: from amb_sdk.sdk import DarwinSdk
 In [2]: s = DarwinSdk()
In [3]: s.auth_login_user('username', 'password')
Out[3]:
 (True,
   Bearer eyJhbGciOiJIUzI1NiIsInR5cCI6IkpXVCJ9.eyJqdGkiOiI2ZDZkMTI3Mi0wZDAxLTRh
 YmMtOWYwOCÓxYWEwZmYxNDY2NjAiLCJpYXQiOjE1MTYyMTY1MTMsImV4cCI6MTUxNjIyMDExMywidH\
 lwZSI6ImFjY2VzcyIsIm5iZiI6MTUxNjIxNjUxMywiaWRlbnRpdHkiOiJjNTc2NzFjNC1lNTAwLTEx\
ZTctOWY4ZS1iNzk2ODU2ZTcwMGYiLCJmcmVzaCI6ZmFsc2V9.slh1mYPy_M7DqAok-tV1NT0kU41A\
 gKQoQHk6nYtetg4')
In [24]: s.upload_dataset('sets/pulsars.csv', 'pulsars-data')
Out[24]: (True, {'dataset_name': 'pulsars-data'})
 In [14]: s.create_model('mydata', target="Diagnosis", model_name="my-model")
Out[14]:
 (True,
  { 'job_name': '1661fb302af149798c34ca9db9e1b0ae', 'model_name': 'my-model'})
In [15]: s.wait_for_job('1661fb302af149798c34ca9db9e1b0ae')
{'percent_complete': 39, 'job_type': 'TrainModel', 'model_name': 'my-model', \
'dataset_names': ['mydata'], 'endtime': None, 'loss': 0.4169575273990631, \
'generations': 11, 'status': 'Running', 'starttime': '2018-02-01T14:16:51.\
464827', 'artifact_names': None}
{'percent_complete': 62, 'job_type': 'TrainModel', 'model_name': 'my-model', \
'dataset_names': ['mydata'], 'endtime': None, 'loss': 0.39973780512809753, \
'generations': 17, 'status': 'Running', 'starttime': '2018-02-01T14:16:51.\
464827', 'artifact_names': None)
generations : 17, Status : None}
{/generations : 17, Status : None}
{/generat_complete': 84, 'job_type': 'TrainModel', 'model_name': 'my-model', 'dataset_names': ['mydata'], 'endtime': None, 'loss': 0.39636287093162537, \
'generations': 21, 'status': 'Running', 'starttime': '2018-02-01T14:16:51.4648\
 27', 'artifact_names': None}
27, artitatc_names: None;
{'percent_complete': 100, 'job_type': 'TrainModel', 'model_name': 'my-model', \
'dataset_names': ['mydata'], 'endtime': '2018-02-01T14:18:02.072976', 'loss': \
0.39636287093162537, 'generations': 23, 'status': 'Complete', 'starttime': \
'2018-02-01T14:16:51.464827', 'artifact_names': None}
 Out[15]: (True, 'Job completed')
 In [16]: s.upload_dataset('sets/cancer_test.csv', 'mytestdata')
Out[16]: (True, {'dataset_name': 'mytestdata'})
```

```
In [17]: s.analyze_data('pulsars-data', feature_select=None, impute="mean", \
drop=True, target=None, max_int_uniques=15, feature_eng="mi", outlier=None, \
 max_unique_values=50)
 Out[17]:
  (True.
   { 'artifact_name': '929bd117a07d411ba40d148ddd686d51',
        'job_name': '4df3e87a87224c1993120482b9b00843'})
In [19]: s.wait_for_job('4df3e87a87224c1993120482b9b00843')
{'starttime': '2018-02-01T15:13:05.624744', 'model_name': None, 'dataset_names'\
: ['yulsars-data'], 'artifact_names': ['929bd117a07d411ba40d148dd686d51'], \
'percent_complete': 100, 'job_type': 'AnalyzeData', 'endtime': '2018-02-01T15:\
13:09.0199', 'generations': None, 'loss': None, 'status': 'Complete'}
Out[19]: (True, 'Job completed')
 In [20]: s.download_artifact('929bd117a07d411ba40d148ddd686d51')
 Out[20]:
  (True,
                      categories
                                                            col_name
                                                                                           col type
                                                                                                                                                mean
                                                                                                                                                                          min \
                                                                                                                             max
                                                                                                                                               5.8125
                      None mean_profile numerical 180.219 95.3869
None std_profile numerical 91.8086 44.3984
   1
                                                                                                                                                    24.772

        None
        kurt_dmsnr
        numerical
        8.06952
        1.24083
        -1.60483

        None
        skew_profile
        numerical
        68.1016
        5.72565
        -1.78189

        None
        mean_dmsnr
        numerical
        222.421
        23.786
        0.273411

        None
        std_dmsnr
        numerical
        110.642
        35.272
        7.56568

        None
        kurt_dmsnr
        numerical
        32.1986
        6.66098
        -3.13927

        None
        skew_dmsnr
        numerical
        10.796
        79.4015
        -1.97698

        A 1
        class
        categorical
        None
        None
        None

   3
   4
   5
   6
   8
                        0,1
                                                   class categorical None
                                                                                                                           None
                                                                                                                                                          None
          num_categories
                                                          std
                             3939.0 36.4588
   0
                              4635.0 8.03585
   1
   2
                              4639.0 1.80623
   3
                             4639.0 11.0723
   4
                             4012.0 39.2822
   5
                             4639.0 23.9892
   6
7
                             4639.0 4.87737
                             4639.0
                                               103.01
                                                       None )
```

Revision Table

Version	Author	Date	Notes
v 1.0	SparkCognition, SCheng	02.05.2018	First Release