This is the script for train and validate xgboost model

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```
In [1]: # Define some useful functions
        # Load some libs
        import os
        import numpy as np
        import h5py as h5
        from matplotlib import pyplot as plt
        from scipy.interpolate import interp1d
        import xgboost
        import pickle
        from sklearn.model selection import train test split, KFold, cross val score
        from sklearn import metrics
        # The function to load hdf5 files into tensors for traiining data
        def load_helix_training_data(datadir):
                f = open(os.path.join(datadir, 'data train.txt'));
                training_datalist = f.read();
                training_datalist = training_datalist.split('\n')[0:-1];
                f.close();
                # Getting the x train, y train
                x train = np.zeros((1, 191));
                y train = np.zeros(1);
                for filename in training datalist:
                         h5file = h5.File(os.path.join(datadir, filename));
                         x_train = np.vstack((x_train, np.array(h5file['data'][:,:])));
                         y train = np.hstack((y train, np.array(h5file['label'][:])));
                         print('INFO: %s Processing Done.. ' %filename);
                return (x_train[1:, :], y_train[1:]);
        # The function to load hdf file as tensors for validate data
        def load helix test data(datadir):
                f = open(os.path.join(datadir, 'data_test.txt'));
                validate datalist = f.read();
                validate datalist = validate datalist.split('\n')[0:-1];
                f.close();
                x validate = np.zeros((1, 191));
                y validate = np.zeros(1);
                for filename in validate datalist:
                         h5file = h5.File(os.path.join(datadir, filename));
                         x_validate = np.vstack((x_validate, np.array(h5file['data']
        [:,:])));
                         y validate = np.hstack((y validate, np.array(h5file['label']
        [:])));
                         print('INFO: %s Processing Done..' %filename);
                 return (x_validate[1:, :], y_validate[1:]);
        # interpolation
```

```
def q interp(x, npoints):
    n = x.shape[0];
    qmin = 0.0;
    qmax = 0.95;
    qold = np.linspace(qmin, qmax, num=x.shape[1])
    qnew = np.linspace(qmin, qmax, num=npoints)
    x_interp = np.zeros((n, npoints))
    for i in range(n):
        f = interp1d(qold, x[i, :], kind='cubic')
        x interp[i, :] = f(qnew)
    return x interp
# get importance
def get importance(model, n, im='weight'):
    imp = model.get booster().get score(importance type=im)
    keys = imp.keys()
    nkeys = len(keys)
    v = np.zeros(n)
    for i in range(nkeys):
        k = 'f%d' \%(i+1)
        try:
            v[i] = imp[k]
        except:
            continue
    return v / v.max()
```

```
In [2]: # Set up data directory
    datadir = 'G:/My Drive/14. CNNWAXS/data/helical_radius'
    prefix = 'radius'
    #os.chdir(datadir)

# open a file for logs, we're not using it rn ...
    #file = open(prefix + '_stats.dat', 'w');

# load the training data (training/validation)
    (x, y) = load_helix_training_data(datadir)

# split the data
    x_train, x_validate, y_train, y_validate = train_test_split(x, y, test_size=0.2, random_state=42)
```

INFO: helical_radius_train.hdf5 Processing Done..

In [4]: # Do a 10-fold cross validation

kfold = KFold(n_splits=10, shuffle=True, random_state=7)
cvrlt = cross_val_score(regressmodel, x_train, y_train, scoring='neg_mean_squa
red_error', cv=kfold, verbose=True)
print('Noise-free 10-fold Cross Validation Result: MSE= %f (%f) \n' % (-cvrlt.
mean(), cvrlt.std()))
#file.write('Noise-free 10-fold Cross Validation Result: MSE= %f (%f) \n' % (-cvrlt.mean(), cvrlt.std()))

[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.

[15:08:28] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:09:24] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:10:10] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:10:57] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:11:44] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:12:34] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:13:21] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:14:14] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror

[15:15:02] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:15:52] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

Noise-free 10-fold Cross Validation Result: MSE= 0.001013 (0.000097)

[Parallel(n jobs=1)]: Done 10 out of 10 | elapsed: 8.2min finished

```
In [5]: # training and using early stopping to get our model
        # Typically stopping iteration is 7000-10000
        # verbose = False for demo
        regressmodel.n estimators = 7500
        regressmodel.fit(x train, y train,
                         eval_metric='rmse',
                         eval_set=[(x_validate, y_validate)],
                         early stopping rounds=int(0.01 * regressmodel.n estimators),
                         verbose=False)
        # See the performance on the training set
        train preds = regressmodel.predict(x train)
        train_mse = metrics.mean_squared_error(y_train, train_preds)
        print('Noise-free Training Result: MSE = %f \n' % train mse)
        #file.write('Noise-free Training Result: MSE = %f \n' % train mse)
        # See the performance on the validation set
        validate preds = regressmodel.predict(x validate)
        validate mse = metrics.mean squared error(y validate, validate preds)
        print('Noise-free Validation Result: MSE = %f \n' % validate mse)
        #file.write('Noise-free Validation Result: MSE = %f \n' % validate mse)
        # See the performance on the testing set
        (x test, y test) = load helix test data(datadir)
        test_preds = regressmodel.predict(x_test)
        test mse = metrics.mean_squared_error(y_test, test_preds)
        print('Noise-free Testing Result: MSE = %f \n' % test mse)
        #file.write('Noise-free Testing Result: MSE = %f \n\n' % test mse)
        [15:16:59] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec
        tive/regression obj.cu:152: reg:linear is now deprecated in favor of reg:squa
        rederror.
        Noise-free Training Result: MSE = 0.000028
        Noise-free Validation Result: MSE = 0.000044
        INFO: helical radius test.hdf5 Processing Done..
        Noise-free Testing Result: MSE = 0.000045
```

```
In [6]: ## The noisy regression model
        noise = 0.01 # Let's just try this and see ...
        # file.write('Noisy regression model with noise = %f : \n' % noise)
        x train noise = np.log10(np.multiply(10**x train, 1 + noise*np.random.rand(x t
        rain.shape[0], x train.shape[1])))
        x_validate_noise = np.log10(np.multiply(10**x_validate, 1 + noise*np.random.ra
        nd(x validate.shape[0], x train.shape[1])))
        # exactly the same xqb model, different name
        noiseregressmodel = xgboost.XGBRegressor(colsample_bytree=0.4,
                                             gamma=0,
                                             learning_rate=0.07,
                                             max_depth=3,
                                             min child weight=1.5,
                                             n estimators=750,
                                             reg_alpha=0.75,
                                             reg lambda=0.45,
                                             subsample=0.8,
                                             seed=42)
        # 10 fold cross validation
        kfold = KFold(n_splits=10, shuffle=True, random_state=7)
        cvrlt = cross_val_score(noiseregressmodel, x_train_noise, y_train, scoring='ne
        g_mean_squared_error', cv=kfold, verbose=True)
        print('Noisy 10-fold Cross Validation Result: MSE= %f (%f) \n' % (-cvrlt.mean
        (), cvrlt.std()))
        #file.write('Noise-free 10-fold Cross Validation Result: MSE= %f (%f) \n' % (-
        cvrlt.mean(), cvrlt.std()))
```

[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.

[15:27:28] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:28:41] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:29:54] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:31:08] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:32:21] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:33:36] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:34:45] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:35:57] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:37:08] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:38:21] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

Noisy 10-fold Cross Validation Result: MSE= 0.003049 (0.000280)

[Parallel(n jobs=1)]: Done 10 out of 10 | elapsed: 12.1min finished

```
In [7]:
        noiseregressmodel.n estimators = 7500
        noiseregressmodel.fit(x train noise, y train,
                          eval metric='rmse',
                          eval set=[(x validate noise, y validate)],
                          early stopping rounds=int(0.01 * noiseregressmodel.n estimato
        rs),
                          verbose=False)
        # See the performance on the training set
        train_preds = noiseregressmodel.predict(x_train_noise)
        train mse = metrics.mean squared error(y train, train preds)
        print('Noisy Training Result: MSE = %f \n' % train mse)
        #file.write('Noisy Training Result: MSE = %f \n' % train mse)
        # See the performance on the validation set
        validate_preds = noiseregressmodel.predict(x_validate_noise)
        validate mse = metrics.mean squared error(y validate, validate preds)
        print('Noisy Validation Result: MSE = %f \n' % validate mse)
        #file.write('Noisy Validation Result: MSE = %f \n' % validate mse)
        # See the performance on the testing set
        # trained on noisy data, test on noisy data
        x test noise = np.log10(np.multiply(10**x test, 1 + noise*np.random.rand(x test))
        t.shape[0], x_test.shape[1])))
        test preds = noiseregressmodel.predict(x test noise)
        test mse = metrics.mean squared error(y test, test preds)
        print('Noisy Testing Result: MSE = %f \n' % test mse)
        #file.write('Noisy Testing Result: MSE = %f \n\n' % test mse)
        [15:40:54] WARNING: C:/Jenkins/workspace/xgboost-win64 release 0.90/src/objec
        tive/regression obj.cu:152: reg:linear is now deprecated in favor of reg:squa
        rederror.
        Noisy Training Result: MSE = 0.000119
        Noisy Validation Result: MSE = 0.000549
```

Noisy Testing Result: MSE = 0.000553

```
In [8]:
        ## The interpolation model: 100 (down sampling)
        nsample = 100
        #file.write('Down sampling regression model with nsample = %d : \n' % nsample)
        x train down = q interp(x train, nsample)
        x_validate_down = q_interp(x_validate, nsample)
        downregressmodel = xgboost.XGBRegressor(colsample bytree=0.4,
                                             gamma=0,
                                             learning rate=0.07,
                                             max_depth=3,
                                             min child weight=1.5,
                                             n_estimators=750,
                                             reg_alpha=0.75,
                                             reg lambda=0.45,
                                             subsample=0.8,
                                             seed=42)
        # 10 fold cross validation too
        kfold = KFold(n_splits=10, shuffle=True, random_state=7)
        cvrlt = cross_val_score(downregressmodel, x_train_down, y_train, scoring='neg_
        mean_squared_error', cv=kfold, verbose=True)
        print('Down Sampling 10-fold Cross Validation Result: MSE= %f (%f) \n' % (-cvr
        lt.mean(), cvrlt.std()))
        #file.write('Down Sampling 10-fold Cross Validation Result: MSE= %f (%f) \n' %
         (-cvrlt.mean(), cvrlt.std()))
```

[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.

[15:55:51] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror

[15:56:19] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:56:45] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:57:10] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:57:36] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:58:02] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:58:30] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:58:58] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:59:27] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:59:53] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

Down Sampling 10-fold Cross Validation Result: MSE= 0.001146 (0.000110)

[Parallel(n jobs=1)]: Done 10 out of 10 | elapsed: 4.5min finished

```
In [9]:
        downregressmodel.n estimators = 7500
        downregressmodel.fit(x_train_down, y_train,
                         eval metric='rmse',
                         eval set=[(x validate down, y validate)],
                         early stopping rounds=int(0.01 * downregressmodel.n estimator
        s),
                         verbose=False)
        # See the performance on the training set
        train_preds = downregressmodel.predict(x_train_down)
        train mse = metrics.mean squared error(y train, train preds)
        print('Down Sampling Training Result: MSE = %f \n' % train mse)
        #file.write('Down Sampling Training Result: MSE = %f \n' % train mse)
        # See the performance on the validation set
        validate_preds = downregressmodel.predict(x_validate_down)
        validate mse = metrics.mean squared error(y validate, validate preds)
        print('Down Sampling Validation Result: MSE = %f \n' % validate mse)
        #file.write('Down Sampling Validation Result: MSE = %f \n' % validate_mse)
        # See the performance on the testing set
        x_test_down = q_interp(x_test, nsample)
        test preds = downregressmodel.predict(x test down)
        test_mse = metrics.mean_squared_error(y_test, test_preds)
        print('Down Sampling Testing Result: MSE = %f \n' % test mse)
        #file.write('Down Sampling Testing Result: MSE = %f \n\n' % test mse)
        [17:24:31] WARNING: C:/Jenkins/workspace/xgboost-win64 release 0.90/src/objec
        tive/regression obj.cu:152: reg:linear is now deprecated in favor of reg:squa
        rederror.
        Down Sampling Training Result: MSE = 0.000032
```

Down Sampling Validation Result: MSE = 0.000048

Down Sampling Testing Result: MSE = 0.000051

```
In [10]: | ## The interpolation model: 400 (up sampling)
         nsample = 400
         #file.write('Up sampling regression model with nsample = %d : \n' % nsample)
         x train up = q interp(x train, nsample)
         x_validate_up = q_interp(x_validate, nsample)
         upregressmodel = xgboost.XGBRegressor(colsample bytree=0.4,
                                              gamma=0,
                                              learning rate=0.07,
                                              max_depth=3,
                                              min child weight=1.5,
                                              n estimators=750,
                                              reg alpha=0.75,
                                              reg lambda=0.45,
                                              subsample=0.8,
                                              seed=42)
         # 10 fold cross validation too
         kfold = KFold(n splits=10, shuffle=True, random state=7)
         cvrlt = cross val score(upregressmodel, x train up, y train, scoring='neg mean
          _squared_error', cv=kfold, verbose=True)
         print('Up Sampling 10-fold Cross Validation Result: MSE= %f (%f) \n' % (-cvrlt
         .mean(), cvrlt.std()))
         #file.write('Up Sampling 10-fold Cross Validation Result: MSE= %f (%f) \n' %
          (-cvrlt.mean(), cvrlt.std()))
         upregressmodel.n estimators = 7500
         upregressmodel.fit(x train up, y train,
                           eval metric='rmse',
                           eval_set=[(x_validate_up, y_validate)],
                           early_stopping_rounds=int(0.01 * upregressmodel.n_estimators
         ),
                           verbose=False)
         # See the performance on the training set
         train preds = upregressmodel.predict(x train up)
         train mse = metrics.mean squared error(y train, train preds)
         print('Up Sampling Training Result: MSE = %f \n' % train mse)
         #file.write('Up Sampling Training Result: MSE = %f \n' % train mse)
         # See the performance on the validation set
         validate preds = upregressmodel.predict(x validate up)
         validate mse = metrics.mean squared error(y validate, validate preds)
         print('Up Smpling Validation Result: MSE = %f \n' % validate mse)
         #file.write('Up Smpling Validation Result: MSE = %f \n' % validate mse)
         # See the performance on the testing set
         x_test_up = q_interp(x_test, nsample)
         test preds = upregressmodel.predict(x test up)
         test mse = metrics.mean squared error(y test, test preds)
         print('Up Sampling Testing Result: MSE = %f \n' % test mse)
         #file.write('Up Sampling Testing Result: MSE = %f \n\n' % test mse)
```

[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.

[17:30:27] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:32:02] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:33:37] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:35:22] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:37:06] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:38:47] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:40:30] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:42:09] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:43:46] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[17:45:20] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[Parallel(n jobs=1)]: Done 10 out of 10 | elapsed: 16.5min finished

Up Sampling 10-fold Cross Validation Result: MSE= 0.000849 (0.000098)

[17:47:00] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

Up Sampling Training Result: MSE = 0.000026

Up Smpling Validation Result: MSE = 0.000041

Up Sampling Testing Result: MSE = 0.000042

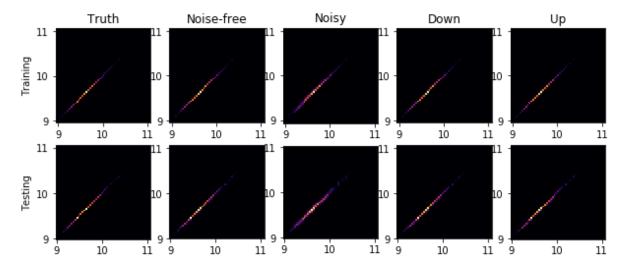
```
In [11]: ## try the random dataset and show that our model does not fit the random dat
         ## If it fits, then xqb might be just fitting anything
         #file.write('Random regression model : \n')
         randommodel = xgboost.XGBRegressor(colsample bytree=0.4,
                                              gamma=0,
                                              learning rate=0.07,
                                              max depth=3,
                                              min child weight=1.5,
                                              n estimators=7500,
                                              reg alpha=0.75,
                                              reg lambda=0.45,
                                              subsample=0.8,
                                              seed=42)
         y_random = y.min() + (y.max() - y.min()) * np.random.rand(x.shape[0]);
         randommodel.fit(x, y_random, verbose=False)
         # See the performance on the random training data
         random preds = randommodel.predict(x)
         random_mse = metrics.mean_squared_error(y_random, random_preds)
         print('Random Training Result: MSE = %f' % random_mse)
         #file.write('Random Training Result: MSE = %f' % random mse)
         # See the performance on the validation set
         validate preds = randommodel.predict(x validate)
         validate mse = metrics.mean squared error(y validate, validate preds)
         print('Random Validation Result: MSE = %f \n' % validate mse)
         #file.write('Random Validation Result: MSE = %f \n' % validate mse)
         # See the performance on the testing set
         test preds = randommodel.predict(x test)
         test mse = metrics.mean squared error(y test, test preds)
         print('Random Testing Result: MSE = %f \n' % test_mse)
         #file.write('Random Testing Result: MSE = %f \n\n' % test_mse)
```

[18:04:30] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

Random Training Result: MSE = 0.345567 Random Validation Result: MSE = 0.218672

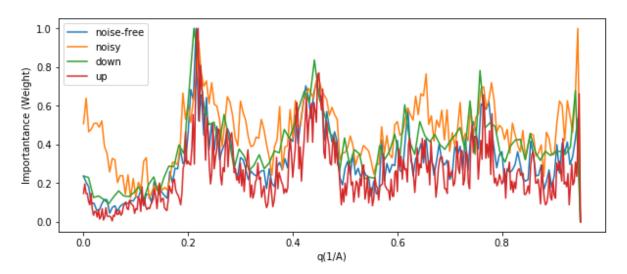
Random Testing Result: MSE = 0.215838

In [12]: | ## Plot the confusion matrix import matplotlib.gridspec as gridspec p = plt.figure("Confusion Matrices", figsize=[10, 4]); plt.subplot(2,5,1, aspect='equal'); plt.hist2d(y_train, y_train, bins=(50, 50), cmap=plt.cm.inferno); plt.title('Truth'); plt.ylabel('Training'); plt.subplot(2,5,2, aspect='equal'); plt.hist2d(y train, regressmodel.predict(x train), bins=(50, 50), cmap=plt.cm. inferno); plt.title('Noise-free'); plt.subplot(2,5,3, aspect='equal'); plt.hist2d(y train, noiseregressmodel.predict(x train noise), bins=(50, 50), c map=plt.cm.inferno); plt.title('Noisy'); plt.subplot(2,5,4, aspect='equal'); plt.hist2d(y_train, downregressmodel.predict(x_train_down), bins=(50, 50), cma p=plt.cm.inferno); plt.title('Down'); plt.subplot(2,5,5, aspect='equal'); plt.hist2d(y train, upregressmodel.predict(x train up), bins=(50, 50), cmap=pl t.cm.inferno); plt.title('Up'); plt.subplot(2,5,6, aspect='equal'); plt.hist2d(y_test, y_test, bins=(50, 50), cmap=plt.cm.inferno); plt.ylabel('Testing'); plt.subplot(2,5,7, aspect='equal'); plt.hist2d(y_test, regressmodel.predict(x_test), bins=(50, 50), cmap=plt.cm.in ferno); plt.subplot(2,5,8, aspect='equal'); plt.hist2d(y_test, noiseregressmodel.predict(x_test_noise), bins=(50, 50), cma p=plt.cm.inferno); plt.subplot(2,5,9, aspect='equal'); plt.hist2d(y test, downregressmodel.predict(x test down), bins=(50, 50), cmap= plt.cm.inferno); plt.subplot(2,5,10, aspect='equal'); plt.hist2d(y test, upregressmodel.predict(x test up), bins=(50, 50), cmap=plt. cm.inferno);



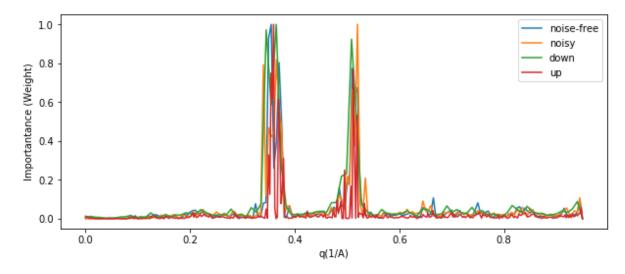
```
In [13]:
         # Plot the importances
         down = 100
         up = 400
         imp type = 'weight'
         q = np.linspace(0.0, 0.95, 191)
         q down = np.linspace(0.0, 0.95, down)
         q_up = np.linspace(0.0, 0.95, up)
         v = get importance(regressmodel, 191, im=imp type)
         v noise = get importance(noiseregressmodel, 191, im=imp type)
         v_down = get_importance(downregressmodel, down, im=imp_type)
         v up = get importance(upregressmodel, up, im=imp type)
         p = plt.figure('Information Flow', figsize=[10, 4]);
         plt.plot(q, v)
         plt.plot(q, v_noise)
         plt.plot(q_down, v_down)
         plt.plot(q up, v up)
         plt.legend(['noise-free', 'noisy', 'down', 'up'])
         plt.xlabel('q(1/A)')
         plt.ylabel('Importantance (Weight)')
```

Out[13]: Text(0, 0.5, 'Importantance (Weight)')



```
In [14]: | # Plot the importance - gain
         imp_type = 'total_gain'
         q = np.linspace(0.0, 0.95, 191)
         q down = np.linspace(0.0, 0.95, down)
         q up = np.linspace(0.0, 0.95, up)
         v = get importance(regressmodel, 191, im=imp type)
         v noise = get importance(noiseregressmodel, 191, im=imp type)
         v down = get importance(downregressmodel, down, im=imp type)
         v_up = get_importance(upregressmodel, up, im=imp_type)
         p = plt.figure('Information Flow', figsize=[10, 4]);
         plt.plot(q, v)
         plt.plot(q, v_noise)
         plt.plot(q down, v down)
         plt.plot(q_up, v_up)
         plt.legend(['noise-free', 'noisy', 'down', 'up'])
         plt.xlabel('q(1/A)')
         plt.ylabel('Importantance (Weight)')
```

Out[14]: Text(0, 0.5, 'Importantance (Weight)')



```
In [15]: from scipy.signal import savgol_filter
# Sampling from a n x 3 data
def curve_sampling(data, nsample, chaos=0.05):

    n = data.shape[0];
    q = data[:, 0];
    sampled_data = np.zeros((nsample, n));
    for i in range(n):
        sampled_data[:, i] = np.random.normal(data[i, 1], data[i, 2]*chaos, (n sample,));
    for i in range(nsample):
        sampled_data[i, :] = savgol_filter(sampled_data[i, :], 21, 3);

    return q, sampled_data;
```

```
In [16]: err500k = np.loadtxt('F:\\Yen\\DuplexData_Full\\rna12k500.dat');
   __, chaos1 = curve_sampling(err500k, 5000, chaos=1.0);
   chaos1 = np.log10(chaos1) - np.log10(chaos1[0,0]) + x_train[0,0];
   m500k = regressmodel.predict(chaos1).mean() # 9.81 - ish where the real value should be 9.76 +- 0.04
   print('The prediction of radius for 500mM KCl data is %f' % m500k)
```

C:\Users\Yen-Lin\AppData\Roaming\Python\Python37\site-packages\scipy\signal_ arraytools.py:45: FutureWarning: Using a non-tuple sequence for multidimensio nal indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result either in an error or a different result.

```
b = a[a slice]
```

C:\Users\Yen-Lin\WPy64-3720\python-3.7.2.amd64\lib\site-packages\ipykernel_la uncher.py:3: RuntimeWarning: invalid value encountered in log10

This is separate from the ipykernel package so we can avoid doing imports u ntil

The prediction of radius for 500mM KCl data is 9.811168

```
In [17]: err5mg = np.loadtxt('F:\\Yen\\DuplexData_Full\\rna12mg5.dat');
    __, chaos1 = curve_sampling(err5mg, 5000, chaos=0.25); # try a smaller chaos p
    arameter
    chaos1 = np.log10(chaos1) - np.log10(chaos1[0,0]) + x_train[0,0];
    m5mg = regressmodel.predict(chaos1).mean() # 9.52 - ish where the real value s
    hould be 9.54 +- 0.15
    print('The prediction of radius for 5mM MgCl2 data is %f' % m5mg)
```

The prediction of radius for 5mM MgCl2 data is 9.558842

And the rest of the numbers/figures in the paper were basically generated this way.

The training time reported here is about 3x more since I was running other routines in the background.

If you find this useful, please cite our paper.

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