



2. The variance expands

^{3. &#}x27;squared_exponential', 'generalized_exponential', 'absolute_exponential', 'linear', 'cubic'

2. Bayesian Optimazation

- 1. from 'x=np.linspace(-2,2,200)'\
- 2. It is in /black_box/objectives.py
- 3. GP is trained in optimize()
- 4. (1) Generate training data
 - (2) Define objective function
 - (3) Training Gaussian process in interations: finding a valuable training point based on Expected Improvement

```
3.
X = self._ensure_shape(X)
y_pred, y_variance = self.gp.predict(X, eval_MSE=True)
if predict_variance:
 return y_pred, y_variance
   return y_pred
4.
X = self. ensure shape(X)
Y_pred, Y_variance = self.gp.predict(X, eval_MSE=True)
y_plus = np.max(Y_pred)
result = np.zeros(Y_pred.shape)
Xi = 0.1
for i in range(0,X.shape[0]):
  Z = (Y_pred[i]-y_plus-Xi)/Y_variance[i]
  result[i]= (Y_pred[i]-y_plus-Xi)*stats.norm.cdf(Z) + Y_variance[i]*stats.norm.pdf(Z)
return result
3. Random Forests Classifier for MNIST
2. (a) 0.876428790843
  (b) prameters = np.array([5,10,10,5,200])
3. (a)
              'min_samples_split': int(params[1]),
              'max_depth': int(params[2]),
              'min_samples_leaf': int(params[3]),
              'max_features': int(params[4])}
(b) Sometimes
  0.896714220283
  prameters: 5,18,6,68,166
(c)
  objective = black_box.RandomForestObjective(
     X_train=np.load("data/mnistFeatures.npy"),
     y_train=np.load("data/mnistlabels.npy"))
  xv = np.random.randint(1,200,(200,4))
  xu = np.random.randint(1,10,(200,1))
  x = np.vstack((xu.T,xv.T))
  bo = BO(objective, noise=1e-1)
  for _ in xrange(50):
     bo.optimize(num_iters=1)
```