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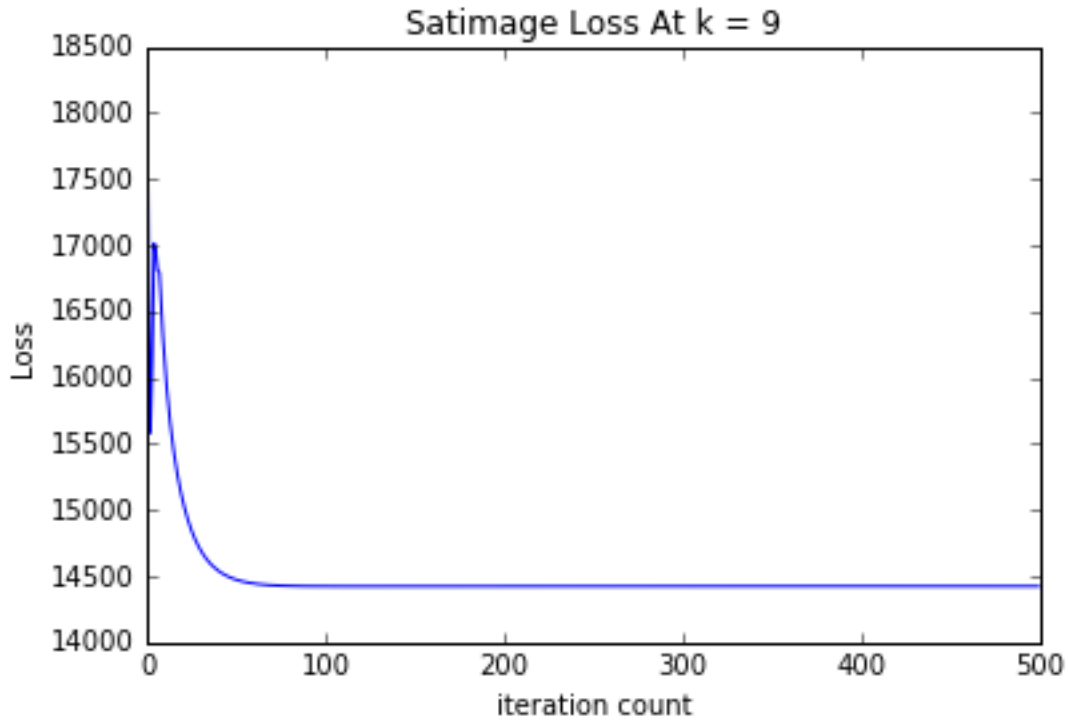
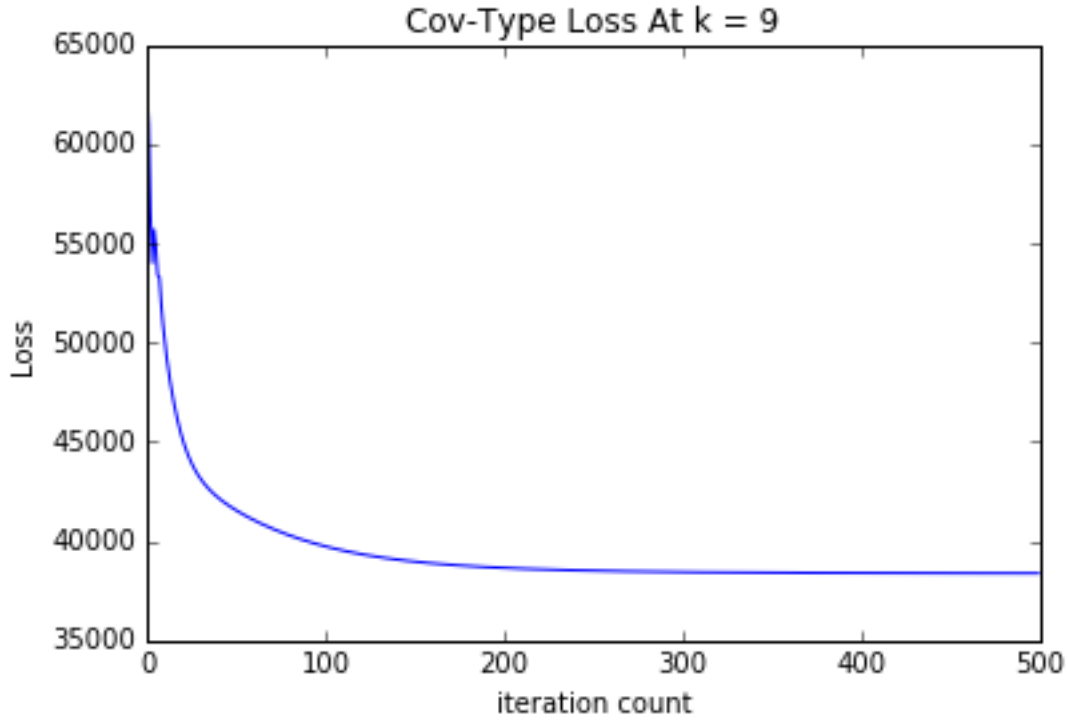
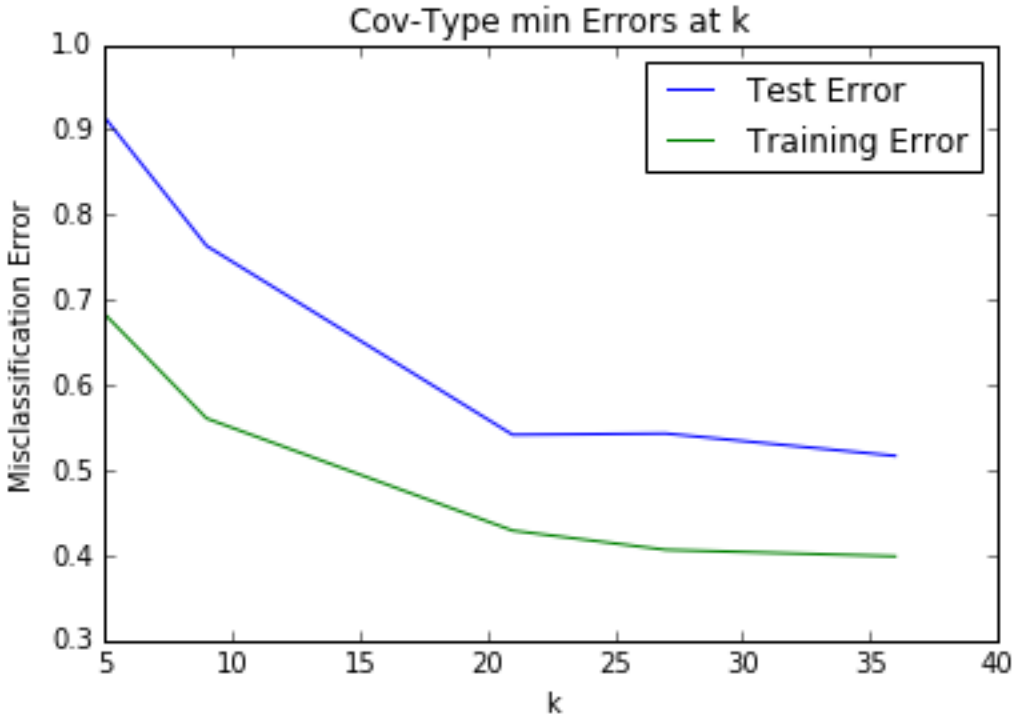
STA 5635

Satimage



Cov-Type


Cov-Type Loss At k = 9
Loss
iteration count
65000
60000
55000
50000
45000
40000
35000
0
100
200
300
400
500


Cov-Type min Errors at k
Misclassification Error
k
Test Error
Training Error
1.0
0.9
0.8
0.7
0.6
0.5
0.4
0.3
5
10
15
20
25
30
35
40

Misclassification Error

Dataset	k	Test Error	Train Error
Satimage	5	0.655500	0.656595
Satimage	9	0.651000	0.651409
Satimage	21	0.264500	0.238331
Satimage	27	0.261500	0.235400
Satimage	36	0.245000	0.218715
Cov-Type	5	0.914708	0.683532
Cov-Type	9	0.763602	0.560847
Cov-Type	21	0.541600	0.428968
Cov-Type	27	0.543063	0.406415
Cov-Type	36	0.516842	0.398810

Code

Included are functions for training and updating the weights, and setting the schedule for eliminating variables. For full code see `https://github.com/johansent/ML_Fall2017/tree/master/hw7`.

```
def updateWeights1(X,Y,w, Ncol,Nrow, learningRate, s):
    L = len(w)
    derivative = np.array([[0.0]*(Ncol)]*L)
    lorenz = 0.0
    X = np.array(X)
    U = np.dot(X, np.transpose(w))
    Uy = [U[i][Y[i]] for i in range(Nrow)]
    for l in range(L):
        diff = (Uy - U[:,l]) - 1
        diff = np.array([-((2 * d) / (1 + d*d)) if d <= 0 else 0 for d in diff])
```

```
            logical = [l != Y[i] for i in range(Nrow)]
            diff = diff * logical
            lorenz += sum(np.log(1 + diff**2))
            dmat = diff * X.transpose()
            derivative[l] = np.sum(dmat, axis = 1)
        total = [X[i][0] if logical[i] else 0 for i in range(Nrow)]
        w = w - (learningRate * (derivative + (s * w)))
        loss = (lorenz + s * np.linalg.norm(w, 'fro'))
        return w, loss

def TrainWeights(X,Y,Xtest,Ytest,niter,k, learnRate = .01):
    Nrow = len(X)
    Ncol = len(X.columns)
    L = 7
    X1 = X.copy()
    Xtest1 = Xtest.copy()
    s = .001
    w = np.array([[0]*(Ncol)]*L)
    loss = []
    testErrors = []
    trainingErrors = []
    for i in range(niter):
        w, newloss = updateWeights1(X1,Y,w, Ncol,Nrow, learnRate, s)
        loss.append(newloss)
        if Ncol > k:
            Mi = getMi(Ncol, k, niter, i, u = 100)
            w,X,Xtest,Ncol = getMBest(w, X1, Xtest1, Mi, Ncol)
        testErrors.append(Test(w,Xtest1, Ytest))
        trainingErrors.append(Test(w,X1,Y))
        if(i % 50 == 0):
            print('i', i)
    return testErrors,trainingErrors, loss

def getMi(M, k, N, i, u = 100):
    return round(k + (M - k) * max([0,(N - 2 * i)/(2 * i * u + N)]))
```

```
def getMBest(w, X, Xtest, M, Ncol):
    summation = sum(w)
    best = sorted(range(len(summation)), key=lambda i: summation[i])[-M:]
    worst = sorted(range(len(summation)), key = lambda i: summation[i])[0:(Ncol - M)]
    w = np.array([[x[i] for i in sorted(best)] for x in w])
    X.drop(X.columns[worst], axis=1, inplace=True)
    Xtest.drop(Xtest.columns[worst], axis=1, inplace=True)
    return w, X, Xtest, len(X.columns)
```

Bibliography

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