# Hidden Markov Models

April 14, 2015

### 1 Imports utiles

```
In [49]: import numpy as np
    import scipy, scipy.linalg, scipy.signal
    import matplotlib.pyplot as plt
    import yahmm as yh
    import pickle
    import pylab

%matplotlib inline
    pylab.rcParams['figure.figsize'] = (14.0, 8.0)
```

## 2 Tool functions

```
In [50]: def modelFromLambda(p0,A,B):
             nStates=A.shape[0]
             nObs=B.shape[1]
             model=yh.Model()
             states=[]
             for i in range(nStates):
                 distparam=dict(zip(range(nObs),B[i]))
                 dist=yh.DiscreteDistribution(distparam)
                 states.append(yh.State(dist,name='{}-state'.format(i)))
                 model.add_transition(model.start,states[i],p0[i])
             for i1,s1 in enumerate(states):
                 for i2,s2 in enumerate(states):
                     if A[i1,i2] != 0.:
                         model.add_transition(s1,s2,A[i1,i2])
             return model
         def modelToLambda(model):
             states=filter(lambda s:s.name.endswith('-state'), model.states)
             states=sorted(states,key=lambda s:int(s.name.split('-')[0]))
             nStates=len(states)
             nObs=len(states[0].distribution.parameters[0])
```

```
A=np.zeros((nStates,nStates))
B=np.zeros((nStates,nObs))
p0=np.zeros((nStates,))
for s1,s2,d in model.graph.edges(data=True):
    if (s2 in states):
        i2=states.index(s2)
        if s1==model.start:
            p0[i2]=np.exp(d['weight'])
        if (s1 in states):
            i1=states.index(s1)
            A[i1,i2]=np.exp(d['weight'])
for s in states:
    if (s in states):
        i=states.index(s)
        for o in range(nObs):
            B[i,o]=s.distribution.parameters[0][o]
return (p0,A,B)
```

# 3 Apprentissage des HMM

```
In [15]: # Load data
         data = pickle.load(open('TD4b-x1.pick', 'rb'))
         n0bs = data['n0bs']
         nStates = data['nStates']
         sequences = data['sequences']
In [68]: # Init model
         A = np.abs(np.random.normal(0.5, 0.25, (nStates,nStates)))
         B = np.abs(np.random.normal(0.5, 0.25, (nStates,nObs)))
         p0 = np.abs(np.random.normal(0.5, 0.25, (nStates,)))
         A /= np.sum(A, axis=1, keepdims=True)
         B /= np.sum(B, axis=1, keepdims=True)
         p0 /= np.sum(p0)
         model = modelFromLambda(p0, A, B)
In [78]: # Train model
         model.bake()
         model.train(sequences, min_iterations=10, verbose=False)
         (p0, A, B) = modelToLambda(model)
```

#### 4 HMM à observations continues

```
x = data['x']
         y = data['y']
In [88]: # Init model
        model = yh.Model()
         s1 = yh.State(yh.NormalDistribution(mean=-1, std=1))
         s2 = yh.State(yh.NormalDistribution(mean=1, std=2))
         model.add_state(s1)
         model.add_state(s2)
         model.add_transition(model.start, s1, 1)
         model.add_transition(model.start, s2, 0)
         model.add_transition(s1, s1, 0.868)
         model.add_transition(s1, s2, 1 - 0.868)
         model.add_transition(s2, s2, 0.819)
         model.add_transition(s2, s1, 1 - 0.819)
         model.bake()
In [124]: # Predict
          (_, xhat) = model.viterbi(y)
          xhat = xhat[1:-1]
          xhat = np.array(zip(*xhat)[0])
In [132]: # Plot result
         plt.subplot(2,1,1)
         plt.plot(y, 'b')
          plt.subplot(2,1,2)
         plt.plot(x, 'r.')
         plt.plot(xhat, 'b--')
         plt.ylim([-0.5,1.5])
Out[132]: (-0.5, 1.5)
```

