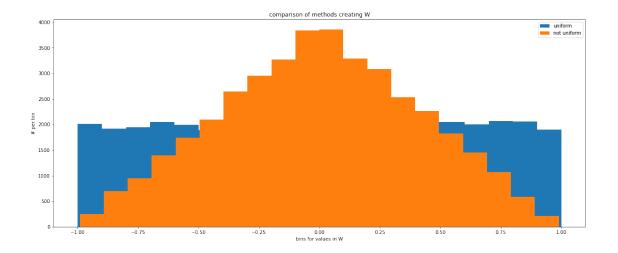
solution07

November 10, 2020

Exercise Sheet 7 Stochastic Optimization

plt.show()

```
[2]: import numpy as np
     import matplotlib.pyplot as plt
     np.random.seed(43222)
     plt.rc('figure', figsize=(20.0, 8.0))
[3]: def make W(N=6):
         W = np.random.uniform(high=1,low=-1,size=(N,N))
         i_lower = np.tril_indices(N,-1)
         W[i_lower] = W.T[i_lower]
         np.fill_diagonal(W,0)
         return W
     def make_W_not_quite_uniform(N=6):
         W = np.random.uniform(high=1,low=-1,size=(N,N))
         W = (W+W.T)/2
         np.fill_diagonal(W,0)
         return W
     W_uni = make_W(200).flatten()
     W_not_uni = make_W_not_quite_uniform(200).flatten()
     plt.hist(W_uni,bins=20,label="uniform")
     plt.hist(W_not_uni,bins=20,label="not uniform")
     plt.title("comparison of methods creating W")
     plt.xlabel("bins for values in W")
     plt.ylabel("# per bin")
     plt.legend()
```

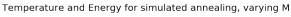


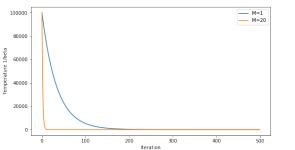
7.1 Simulated Annealing

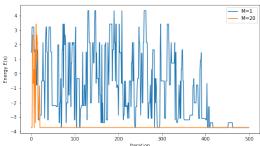
```
[4]: def make_s(N=6):
         return np.random.choice([-1,1],size=N)
     def E(s,W):
         return -s.T@W@s/2
     def E_si(i,s,W):
         return -s[i]*W[i]@s
     def P_s(s,Z,beta,W):
         return np.exp(-beta*E(s,W))/Z
     def sim_annealing(W,s,tmax=500,M=1,N=6,beta=0.00001,tau=1.03):
         T_t = []
         E_s = []
         for t in range(tmax):
             T_t.append(1/beta)
             E_s.append(E(s,W))
             for m1 in range(M):
                 for m2 in range(M):
                     i = np.random.randint(0,high=6)
                     delta_E = -2*E_si(i,s,W)
                     P = 1/(1+np.exp(beta*delta_E))
                     s[i] *= np.random.choice([-1,1],p=[P,1-P])
                 beta *= tau
         return (T_t, E_s)
```

```
[5]: s = make_s()
W = make_W()
```

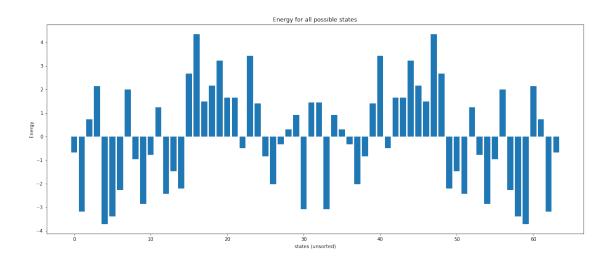
/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:23: RuntimeWarning: overflow encountered in exp







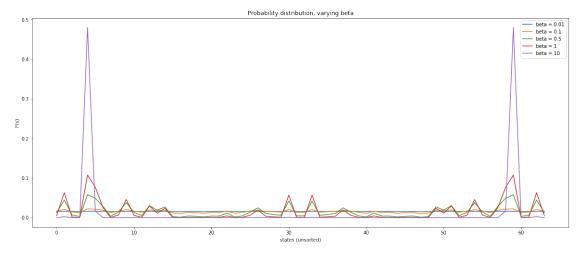
[6]: []



```
betas = [0.01,0.1,0.5,1,10]

for i,beta in enumerate(betas):
    Z = np.sum(np.exp(-beta*np.array(E_s_full)))
    P = [P_s(s,Z,beta,W) for s in s_permutations]
    plt.plot(range(64),P,label=f"beta = {beta}")

plt.legend()
plt.xlabel("states (unsorted)")
plt.ylabel("P(s)")
plt.title("Probability distribution, varying beta")
plt.show()
```



7.2 Mean-Field Annealing

```
[11]: def Q_s(beta,e,s,Z_Q):
          return np.exp(-beta*e@s)/Z_Q
      def e_func(s,W):
          return -W@s
      def mean_field_annealing(W,tmax=50,epsilon=0.00000001,N=6,beta=1,tau=1.1):
          s = np.random.rand(N)
          T t = []
          E_s = []
          e old = np.zeros((N))
          e_new = e_func(s,W)
          for t in range(tmax):
              T_t.append(1/beta)
              E_s.append(E(s,W))
              if np.linalg.norm(e_new-e_old)<epsilon:</pre>
                  break
              s = np.tanh(-beta*e_new)
              e_old = e_new
              e_new = e_func(s,W)
              beta *= tau
          return (T_t, E_s)
      fig,ax = plt.subplots(1,2, figsize=(20,5))
      T_t, E_s = mean_field_annealing(W)
      ax[0].plot(T_t)
      ax[1].plot(E_s)
      ax[0].set_xlabel("Iteration")
      ax[0].set_ylabel("Temperature 1/beta")
      ax[1].set_xlabel("Iteration")
      ax[1].set_ylabel("Energy E(s)")
      fig.suptitle("mean field annealing",fontsize=15)
      plt.show()
```

