

»

In [2]:

```
1 import numpy as np
2 import matplotlib.pyplot as plt
```

»

In [3]:

```
1 #here i genereate 1,000,000 data points using the generation model
2 n_sample = 1000000
3 gn = [np.random.normal(),np.random.normal(),np.random.normal()]
4 data = []
5 for idx in range(3,n_sample+3):
6     g = np.random.normal()
7     gn.append(g)
8     data.append(.1*g+.5*gn[idx-1]-.5*gn[idx-2]+.1*gn[idx-3]) # x[n] = u[n] + 0.8*x[n-1]
```

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In [5]:

```
1 # the following code obtains the optimal weights and MSE
2
3 X = []
4 y = []
5 COV_X = []
6 for three_sample_before, two_sample_before, one_sample_before, curr in zip(data[0:-3:],
7     vector1 = np.array([one_sample_before, two_sample_before, three_sample_before])
8     vector2 = np.array([curr, one_sample_before, two_sample_before]) # for R
9     X.append(vector1)
10    y.append(curr)
11    COV_X.append(vector2)
12 X = np.vstack(X)
13 y = np.array(y)
14 COV_X = np.vstack(COV_X)
15
16
17 R = np.cov(COV_X.T)
18 p = y.dot(X)/len(y)
19 w = np.linalg.inv(R).dot(p)
20 prediction = X.dot(w)
21 print("Optimal Weights:",w)
22 error = y.dot(y)/n_sample - p.dot(w)
23 print("Optimal MSE:",error)
```

Optimal Weights: [-0.68014248 -0.41424331 -0.17878953]

Optimal MSE: 0.3513533155728955



In [44]:

```
1  # here are some helper functions
2  def quantize(values,BW):
3      quant = np.minimum(np.round(np.array(values)*np.power(2.0,BW-1.0))*np.power(2.0,1.0/BW),
4      return quant
5
6  def calc_MSE(x,x_hat):
7      #here x is true values, x_hat is prediction
8      return np.sum(np.square(np.array(x) - np.array(x_hat)))/n_sample
9
10 def evaluate_sqnr(data, dq,mserr):
11     SNR = 10*np.log10(np.true_divide(np.sum(np.var(data)),np.sum(np.var(np.array(data)
12     return SNR
13
14 def predict(wq,xq):
15     x_hats = np.matmul(xq,np.transpose(wq))
16     mse = calc_MSE(y,x_hats)
17     return mse
18
19
20 #first without quantizing:
21 print("Optimal MSE:",error)
22 optimal_SNR = 10*np.log10(np.var(y)/error)
23 print("Optimal SNR (dB):", optimal_SNR)
24
25 #now quantize for various bit precisions
26 precisions = np.arange(1,17,1)
27
28 print("Bits \t| MSE \t| SQNR\t|SQNR-SNR|\tWithin .5dB of SNR?")
29 MSEs = []
30 sqnrs = []
31 for BW in precisions:
32     wq = quantize(w,BW)
33     dq = quantize(X,BW)
34     MSE = predict(wq,dq)
35     MSEs.append(MSE)
36     sqnr = evaluate_sqnr(X,dq,MSE)
37     sqnrs.append(sqnr)
38     print("{:.4f}\t| {:.4f}\t| {:.3f}\t| {:.4f}|\t{}".format(BW,MSE,sqnr,abs(sqnr-optimal
39
```

Optimal MSE: 0.3513533155728955

Optimal SNR (dB): 1.698789959975315

Bits	MSE	SQNR	SQNR-SNR	Within .5dB of SNR?
1.0000	0.5718	-1.896	3.5949	NO
2.0000	0.4453	-0.053	1.7515	NO
3.0000	0.3851	0.900	0.7987	NO
4.0000	0.3698	1.208	0.4906	YES
5.0000	0.3654	1.312	0.3870	YES
6.0000	0.3634	1.357	0.3417	YES
7.0000	0.3627	1.375	0.3239	YES
8.0000	0.3625	1.383	0.3162	YES
9.0000	0.3623	1.387	0.3121	YES
10.0000	0.3623	1.389	0.3101	YES
11.0000	0.3622	1.390	0.3091	YES
12.0000	0.3622	1.390	0.3085	YES

13.0000	0.3622	1.391	0.3082	YES
14.0000	0.3622	1.391	0.3081	YES
15.0000	0.3622	1.391	0.3081	YES
16.0000	0.3622	1.391	0.3080	YES

»

In [28]:

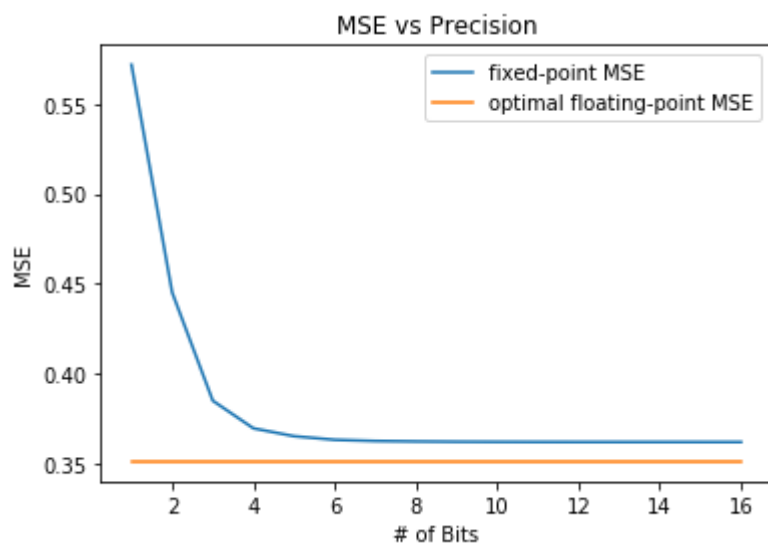
```

1 # we can also plot the MSE and SQNRs across different bit widths
2 plt.figure()
3 plt.plot(precisions, MSEs, label = "fixed-point MSE")
4 plt.title("MSE vs Precision")
5 plt.xlabel("# of Bits")
6 plt.ylabel("MSE")
7 plt.plot(precisions,[error]*len(precisions),label = "optimal floating-point MSE")
8 plt.legend()

```

Out[28]:

<matplotlib.legend.Legend at 0x17a0a5b5198>



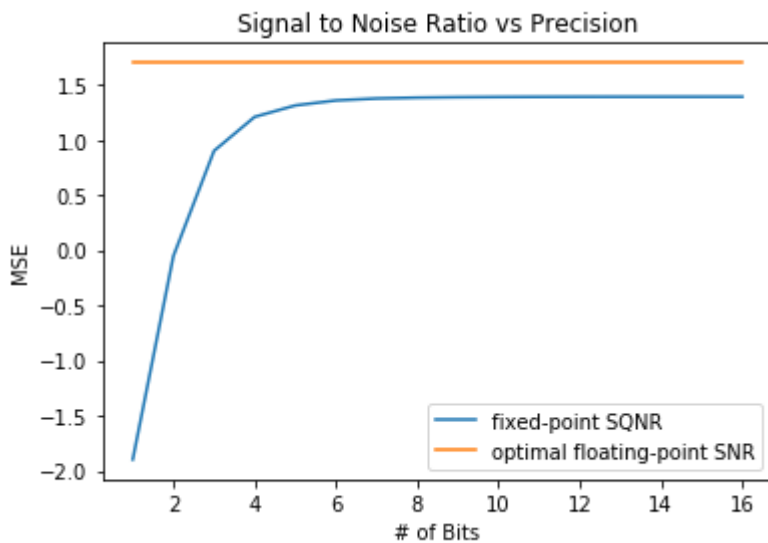
►

In [47]:

```
1 plt.figure()
2 plt.plot(precisions, sqnrs, label = "fixed-point SQNR")
3 plt.title("Signal to Noise Ratio vs Precision")
4 plt.xlabel("# of Bits")
5 plt.ylabel("MSE")
6 plt.plot(precisions,[optimal_SNR]*len(precisions),label = "optimal floating-point SNR")
7 plt.legend()
```

Out[47]:

<matplotlib.legend.Legend at 0x17a0a48c9e8>



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In [81]:

```
1 #verilog confirmation:
2 xs = np.array([np.array([0.73500254, 0.55186864, 0.19778589]),[-0.69549231, 0.73500254, -0.66702645]])
3 y_dec= [-0.69549231, -0.66702645]
4 ws_quant = quantize(w,4)*16
5 for x in xs[:-1]:
6     xs_quant = quantize(x,4)
7     print(xs_quant*16,ws_quant)
8     xs_quant = xs_quant*16
9     print(xs_quant.dot(ws_quant)/256,x.dot(w))
```

```
[12.  8.  4.] [-10.  -6.  -2.]
-0.6875 -0.7638763927922381
```

►

In [86]:

```
1 x_dpctest = np.array([0.73500254, 0.55186864, 0.19778589])
2 w_dpctest = np.array([-0.68014248, -0.41424331, -0.17878953])
3 print("Floating point dot-product:", x_dpctest.dot(w_dpctest))
```

Floating point dot-product: -0.7638763887944293

