Progress thus far

- Delay & Sum is done, with decent results. The sound localization is good, albeit a bit fuzzy.
- After playing around with β_{PM} for a little bit, I got very good results using Pressure Matching. The dark bands are bigger and there are 4 of them now, and the bright bands are more directed. However, I needed to multiply the entire $\mathbf{q}$ vector by scalar g to get the same magnitudes I got with D&S.
- Neither Amplitude Contrast Maximization (ACM) nor Energy Difference Maximization (EDM) have been developed yet. I have been having trouble calculation the regularization parameters (β_{ACM} and α) for both methods. This is likely due to an issue with the eigenvalue routine, which should hopefully get fixed soon. Talking with Dr. Osborne on this would also really help.
- The overall plan for the next few days is to get all four methods working here, in a Python Jupyter notebook, a purely synthetic environment. After that, I will start building the actual beamformer and make my final decision regarding what should be doing the DSP (an FPGA, processor, or my laptop). That choice will likely come down to time, I think. Or how much lab service I end up doing.

In [426]:
```
import numpy as np
import matplotlib.pyplot as plt
from math import exp, pi, e
```

Formulas for Delay & Sum

$$Z(\mathbf{x}_m, \mathbf{y}_\ell, \omega) = \frac{e^{-j\frac{\omega}{c}||\mathbf{x}_m - \mathbf{y}_\ell||}}{4\pi \ || \ \mathbf{x}_m - \mathbf{y}_\ell \ ||}$$

$$\mathbf{p}(\omega) = \mathbf{Z}(\omega)\mathbf{q}(\omega)s(\omega)$$

$$\mathbf{Z}_{M\times L}(\omega) = \begin{bmatrix} Z(\mathbf{x}_1, \mathbf{y}_1, \omega) & \dots & Z(\mathbf{x}_1, \mathbf{y}_L, \omega) \\ \vdots & \ddots & \vdots \\ Z(\mathbf{x}_M, \mathbf{y}_1, \omega) & \dots & Z(\mathbf{x}_M, \mathbf{y}_L, \omega) \end{bmatrix}$$

$$\mathbf{q}_{DAS} = \mathbf{\Gamma}\mathbf{z}_B^*$$

$$\mathbf{\Gamma} = \begin{bmatrix} \gamma_1 & 0 & \dots & 0 \\ 0 & \gamma_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \gamma_L \end{bmatrix}$$

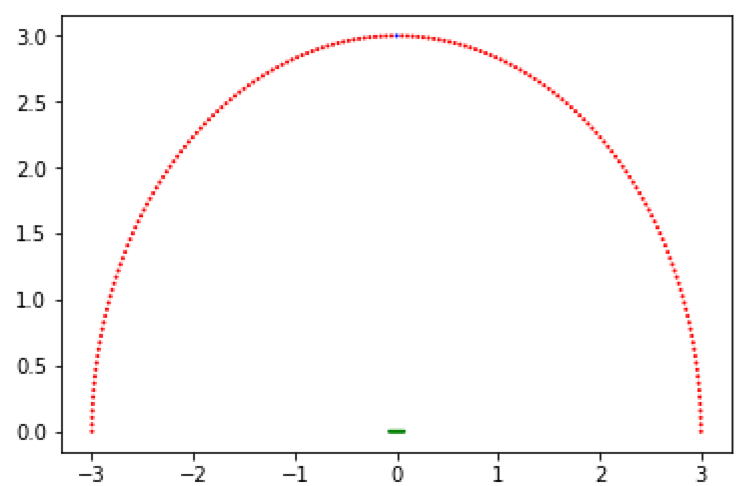
$$\gamma_\ell = \frac{16\pi^2 \ || \ \mathbf{x}_B - \mathbf{y}_\ell \ ||^2}{L}$$

In [427]:
```
c = 343 #speed of sound
l2 = lambda arr: np.sqrt(np.sum(arr**2))

Z = lambda x_m, y_l, omega: e**(-1j * omega / c * l2(x_m - y_l)) / (4 * pi * l2(x_
```

In [428]:
```
delta, delta_theta = 0.02, 1 * pi / 180 #sources 20 mm apart, validation points 1
M, L = 181, 8 #181 validation points, 8 sources
R = 3 # 3 meters
x = np.zeros((M, 3))
for i in range(M):
    x[i] = np.array([R*np.cos(delta_theta*(i)), R*np.sin(delta_theta*(i)), 0])
y = np.zeros((L, 3))
for l in range(L):
    y[l] = np.array([(l-3.5)*delta, 0, 0])
print(x, y)
```

```
[[  3.00000000e+00   0.00000000e+00   0.00000000e+00]
 [  2.99954309e+00   5.23572193e-02   0.00000000e+00]
 [  2.99817248e+00   1.04698490e-01   0.00000000e+00]
 [  2.99588860e+00   1.57007869e-01   0.00000000e+00]
 [  2.99269215e+00   2.09269421e-01   0.00000000e+00]
 [  2.98858409e+00   2.61467228e-01   0.00000000e+00]
 [  2.98356569e+00   3.13585390e-01   0.00000000e+00]
 [  2.97763845e+00   3.65608030e-01   0.00000000e+00]
 [  2.97080421e+00   4.17519303e-01   0.00000000e+00]
 [  2.96306502e+00   4.69303395e-01   0.00000000e+00]
 [  2.95442326e+00   5.20944533e-01   0.00000000e+00]
 [  2.94488155e+00   5.72426986e-01   0.00000000e+00]
 [  2.93444280e+00   6.23735072e-01   0.00000000e+00]
 [  2.92311019e+00   6.74853163e-01   0.00000000e+00]
 [  2.91088718e+00   7.25765687e-01   0.00000000e+00]
 [  2.89777748e+00   7.76457135e-01   0.00000000e+00]
 [  2.88378509e+00   8.26912067e-01   0.00000000e+00]
 [  2.86891427e+00   8.77115114e-01   0.00000000e+00]
 [  2.85316955e+00   9.27050983e-01   0.00000000e+00]
```

In [358]:
```
plt.scatter(np.concatenate((x, y))[:, 0], np.concatenate((x, y))[:, 1], c=['r']*(l
plt.show()
```



In [429]:
```
f = 3000 #3 kHz signal
omega = 2 * pi * f #angular frequency
b = M//2 + 1

print(omega, b)

z_b = np.vectorize(complex)(np.zeros((L,)))
Z_d = np.vectorize(complex)(np.zeros((M-1, L)))
for l in range(L):
    print(x[b] - y[l], l2(x[b] - y[l]), Z(x[b], y[l], omega))
    z_b[l] = Z(x[b], y[l], omega)
n = 0
for m in range(0, M):
    if m == M//2+1: continue
    for l in range(L):
        print(n, m, l)
        Z_d[n, l] = Z(x[n], y[l], omega)
    n += 1
gamma = lambda x_b, y, l: 16 * (pi ** 2) * l2(x_b - y[l])**2 / L
Gamma = np.asmatrix(np.zeros((L, L)))
for l in range(L):
    Gamma[l, l] = gamma(x[b], y, l)
print(z_b, Gamma)
```

```
8 8 6
8 8 7
9 9 0
9 9 1
9 9 2
9 9 3
9 9 4
9 9 5
9 9 6
9 9 7
10 10 0
10 10 1
10 10 2
10 10 3
10 10 4
10 10 5
10 10 6
10 10 7
11 11 0
```

In [430]:
```
q_DAS = np.dot(Gamma, np.conjugate(z_b))
```

In [432]:
```
print(q_DAS)
```

```
[[ 0.42795313+4.69227772j  0.44108448+4.69098116j  0.41985128+4.69305916j
   0.36421000+4.6980447j   0.27406334+4.70471356j  0.14930517+4.71108159j
  -0.01010674+4.71440292j -0.20405013+4.71117287j]]
```

In [433]:
```
ZZ = np.vectorize(complex)(np.zeros((M, L)))
print(ZZ)
for m in range(M):
    for l in range(L):
        ZZ[m, l] = Z(x[m], y[l], omega)
print(ZZ)
```

```
[[ 0.+0.j  0.+0.j  0.+0.j ...,  0.+0.j  0.+0.j  0.+0.j]
 [ 0.+0.j  0.+0.j  0.+0.j ...,  0.+0.j  0.+0.j  0.+0.j]
 [ 0.+0.j  0.+0.j  0.+0.j ...,  0.+0.j  0.+0.j  0.+0.j]
 ...,
 [ 0.+0.j  0.+0.j  0.+0.j ...,  0.+0.j  0.+0.j  0.+0.j]
 [ 0.+0.j  0.+0.j  0.+0.j ...,  0.+0.j  0.+0.j  0.+0.j]
 [ 0.+0.j  0.+0.j  0.+0.j ...,  0.+0.j  0.+0.j  0.+0.j]]
[[ 0.01540833+0.02084422j -0.01164233+0.02334941j -0.02626209+0.00024055j
  ...,  0.02650657+0.0039125j   0.00861733+0.02556198j
  -0.01898357+0.01942331j]
 [ 0.01539644+0.02085311j -0.01165197+0.02334468j -0.02626219+0.00023402j
  ...,  0.02650752+0.00390577j  0.00862818+0.02555824j
  -0.01897185+0.01943462j]
 [ 0.01536076+0.02087974j -0.01168087+0.02333044j -0.02626247+0.00021443j
  ...,  0.02651036+0.00388558j  0.00866074+0.025547j
  -0.01893663+0.01946851j]
 ...,
 [-0.01893663+0.01946851j  0.00866074+0.025547j    0.02651036+0.00388558j
  ..., -0.02626247+0.00021443j -0.01168087+0.02333044j
   0.01536076+0.02087974j]
 [-0.01897185+0.01943462j  0.00862818+0.02555824j  0.02650752+0.00390577j
  ..., -0.02626219+0.00023402j -0.01165197+0.02334468j
   0.01539644+0.02085311j]
 [-0.01898357+0.01942331j  0.00861733+0.02556198j  0.02650657+0.0039125j
  ..., -0.02626209+0.00024055j -0.01164233+0.02334941j
   0.01540833+0.02084422j]]
```

In [434]:
```
p = np.dot(ZZ, q_DAS.T)
print(p)
```

```
 [ -1.79685708e-01 -8.21324226e-03j]
 [ -1.84721700e-01 -8.59396503e-03j]
 [ -1.89312462e-01 -8.95750762e-03j]
 [ -1.93485799e-01 -9.30365511e-03j]
 [ -1.97268909e-01 -9.63229262e-03j]
 [ -2.00688276e-01 -9.94339641e-03j]
 [ -2.03769585e-01 -1.02370248e-02j]
 [ -2.06537638e-01 -1.05133091e-02j]
 [ -2.09016281e-01 -1.07724436e-02j]
 [ -2.11228347e-01 -1.10146772e-02j]
 [ -2.13195602e-01 -1.12403035e-02j]
 [ -2.14938698e-01 -1.14496521e-02j]
 [ -2.16477139e-01 -1.16430796e-02j]

 [ -2.17829245e-01 -1.18209615e-02j]
 [ -2.19012131e-01 -1.19836842e-02j]
 [ -2.20041686e-01 -1.21316369e-02j]
 [ -2.20932556e-01 -1.22652049e-02j]
 [ -2.21698130e-01 -1.23847629e-02j]
 [ -2.22350537e-01 -1.24906686e-02j]
```

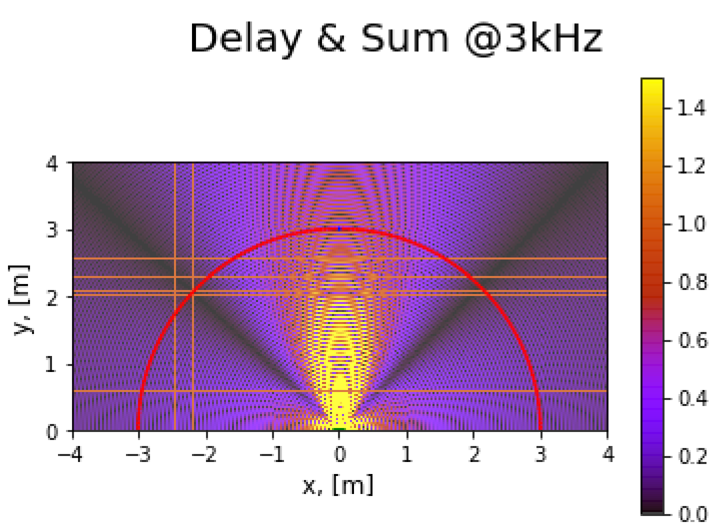
In [435]:
```
M_test = 800 * 400
x_test = np.ones((M_test, 3))
n = 0
for i in range(-400, 400):
    for j in range(0, 400):
        x_test[n] = np.array([i / 100, j / 100, 0])
        n += 1
print(x_test)
ZZ_test = np.vectorize(complex)(np.zeros((M_test, L)))
for m in range(M_test):
    for l in range(L):
        try:
            ZZ_test[m, l] = Z(x_test[m], y[l], omega)
        except:
            ZZ_test[m, l] = 0
print(ZZ_test)
p_test = np.dot(ZZ_test, q_DAS.T)
print('>>', p_test)
print('\n\n>>', x_test, y)
```

```
[[-4.    0.    0.  ]
 [-4.    0.01  0.  ]
 [-4.    0.02  0.  ]
 ...,
 [ 3.99  3.97  0.  ]
 [ 3.99  3.98  0.  ]
 [ 3.99  3.99  0.  ]]
[[-0.01415315-0.01448099j -0.01923294+0.00599694j -0.00338024+0.01975763j
  ...,  0.00027128-0.01974441j -0.01737877-0.00916799j
  -0.01598478+0.01125946j]
 [-0.01416322-0.01447105j -0.01922870+0.0060103j  -0.00336655+0.01975991j
  ...,  0.00025782-0.01974453j -0.01738494-0.00915617j
  -0.01597713+0.01127022j]
 [-0.01419341-0.01444117j -0.01921593+0.00605036j -0.00332548+0.01976667j
  ...,  0.00021743-0.01974483j -0.01740338-0.00912068j
  -0.01595412+0.01130245j]
 ...,
 [-0.00709451+0.01208554j -0.01359476+0.00354495j -0.01216731-0.00709489j
  ...,  0.01365331-0.00387155j  0.01249317+0.00680733j
   0.00419675+0.01363189j]
 [-0.00204135+0.0138471j  -0.01125075+0.00838551j -0.01392665-0.00198404j
  ...,  0.01114799-0.00875306j  0.01412564+0.00154013j
   0.00906690+0.010987j  ]
 [ 0.00330375+0.01358355j -0.00725792+0.01198883j -0.01362893+0.00341299j
  ...,  0.00697658-0.01231724j  0.01363005-0.00395169j
   0.01255904+0.00668404j]]
>> [[ 0.01374098 -1.68089574e-01j]
 [ 0.01362540 -1.68097977e-01j]
 [ 0.01327863 -1.68122706e-01j]
 ...,
 [-0.00865227 -3.41849132e-03j]
 [-0.00866375 -9.18241382e-05j]
 [-0.00753846 +2.76953159e-03j]]
```

```
>> [[-4.    0.    0.  ]
 [-4.    0.01  0.  ]
 [-4.    0.02  0.  ]
 ...,
 [ 3.99  3.97  0.  ]
 [ 3.99  3.98  0.  ]
 [ 3.99  3.99  0.  ]] [[-0.07  0.    0.  ]
 [-0.05  0.    0.  ]
 [-0.03  0.    0.  ]
 [-0.01  0.    0.  ]
 [ 0.01  0.    0.  ]
 [ 0.03  0.    0.  ]
 [ 0.05  0.    0.  ]
 [ 0.07  0.    0.  ]]
```

In [436]:
```
output = np.ones((400,800))
for n in range(x_test.shape[0]):
    output[399 - int(x_test[n, 1]*100), int(x_test[n, 0]*100+400)] = p_test[n, 0]
fig = plt.figure()
plt.scatter(np.concatenate((x, y))[:, 0], np.concatenate((x, y))[:, 1], c=['r']*(
implot = plt.imshow(np.abs(output), interpolation='nearest', cmap='gnuplot', exte
plt.colorbar()
fig.suptitle("Delay & Sum @3kHz", fontsize=20)
plt.xlabel("x, [m]", fontsize=12)
plt.ylabel("y, [m]", fontsize=12)
plt.show()
```

```
C:\Users\sarki\Anaconda3\lib\site-packages\ipykernel_launcher.py:3: ComplexWarn
ing: Casting complex values to real discards the imaginary part
  This is separate from the ipykernel package so we can avoid doing imports unt
il
```



Formulas for Pressure matching

$$\min_{\mathbf{q}} J_{PM} = \min_{\mathbf{q}} (\mathbf{e}_{PM}^H \mathbf{e}_{PM} + \beta_{PM} E_q)$$

$$\mathbf{e}_{PM} = \hat{\mathbf{p}} - \mathbf{Zq}$$

$$E_{\mathbf{q}} = \mathbf{q}^H \mathbf{q}$$

$$\mathbf{q}_{PM} = (\mathbf{Z}^H \mathbf{Z} + \beta_{PM} \mathbf{I})^{-1} \mathbf{Z}^H \hat{\mathbf{p}}$$

In [445]:
```
E_ref = (4 * pi * R)**2 / L
sigma = 25
epsilon_beta, beta_min, E_max = 10e-6, 10e-20, sigma * E_ref
p_b_hat = 1
p_hat = np.array([0]*(M//2) + [1] + [0]*(M//2))
g = 30 # weakness compensation

W = lambda q: p_b_hat / np.dot(z_b.T, q)
E = lambda q: np.dot(np.conjugate(q).T, q)
```

In [446]:
```
### Calculate Tikhonov Regularization Parameter ###

beta = beta_min
print('a')
q_temp = np.dot(np.dot(np.linalg.inv(np.dot(np.conjugate(ZZ).T, ZZ)
                                + beta * np.eye(L)), np.conjugate(ZZ).T), p_hat).
print('b')
q_hat = W(q_temp) * q_temp
n = 0
while E(q_hat) > E_max:
    if n % 1000 == 0: print(beta, E(q_hat), E_max, E(q_hat) > E_max)
    beta += epsilon_beta
    q_temp = np.dot(np.dot(np.linalg.inv(np.dot(np.conjugate(ZZ).T, ZZ)
                                + beta * np.eye(L)), np.conjugate(ZZ).T), p_hat).
    q_hat = W(q_temp) * q_temp
    n += 1

print(beta, E(q_hat), E_max, E(q_hat) > E_max)
beta_PM = beta
```

```
a
b
1e-19 (49355562.8913+0j) 4441.321980490211 True
0.00024000000000000001 (4345.68735005+0j) 4441.321980490211 False
```

In [447]:
```
q_PM = g * np.dot(np.dot(np.linalg.inv(np.dot(np.conjugate(ZZ).T, ZZ) + beta_PM *

p = np.dot(ZZ, q_PM.T)
print(p)
```

```
[ 0.14848238 +7.69761176e-03j  0.14810132 +7.67514038e-03j
  0.14695907 +7.60785406e-03j  0.14505836 +7.49613600e-03j
  0.14240381 +7.34062376e-03j  0.13900194 +7.14220758e-03j
  0.13486119 +6.90202805e-03j  0.12999204 +6.62147287e-03j
  0.12440706 +6.30217293e-03j  0.11812104 +5.94599731e-03j
  0.11115101 +5.55504741e-03j  0.10351645 +5.13164985e-03j
  0.09523935 +4.67834830e-03j  0.08634435 +4.19789388e-03j
  0.07685889 +3.69323421e-03j  0.06681332 +3.16750096e-03j
  0.05624101 +2.62399578e-03j  0.04517855 +2.06617452e-03j
  0.03366581 +1.49762982e-03j  0.02174607 +9.22071732e-04j
  0.00946612 +3.43306710e-04j -0.00312361 -2.34785364e-04j
 -0.01596906 -8.08275859e-04j -0.02901242 -1.37321384e-03j
 -0.04219213 -1.92565411e-03j -0.05544287 -2.46168685e-03j
 -0.06869564 -2.97746844e-03j -0.08187777 -3.46925357e-03j
 -0.09491302 -3.93342813e-03j -0.10772178 -4.36654268e-03j
 -0.12022121 -4.76534628e-03j -0.13232548 -5.12682011e-03j
 -0.14394605 -5.44821063e-03j -0.15499198 -5.72706184e-03j
 -0.16537032 -5.96124608e-03j -0.17498654 -6.14899304e-03j
 -0.18374499 -6.28891640e-03j -0.19154940 -6.38003759e-03j
 -0.19830352 -6.42180629e-03j -0.20391169 -6.41411704e-03j
 -0.20827950 -6.35732164e-03j -0.21131454 -6.25223681e-03j
 -0.21292711 -6.10014682e-03j -0.21303101 -5.90280069e-03j
 -0.21154435 -5.66240364e-03j -0.20839036 -5.38160280e-03j
 -0.20349827 -5.06346681e-03j -0.19680411 -4.71145943e-03j
 -0.18825159 -4.32940727e-03j -0.17779294 -3.92146159e-03j
 -0.16538970 -3.49205478e-03j -0.15101354 -3.04585157e-03j
 -0.13464696 -2.58769571e-03j -0.11628406 -2.12255259e-03j
 -0.09593108 -1.65544854e-03j -0.07360706 -1.19140758e-03j
 -0.04934425 -7.35386532e-04j -0.02318854 -2.92209389e-04j
  0.00480026 +1.33497988e-04j  0.03454826 +5.37371863e-04j
  0.06596736 +9.15372202e-04j  0.09895542 +1.26384077e-03j
  0.13339633 +1.57955467e-03j  0.16916044 +1.85977438e-03j
  0.20610493 +2.10228518e-03j  0.24407447 +2.30543128e-03j
  0.28290195 +2.46814170e-03j  0.32240932 +2.58994742e-03j
  0.36240862 +2.67098903e-03j  0.40270312 +2.71201492e-03j
  0.44308855 +2.71436932e-03j  0.48335447 +2.67997067e-03j
  0.52328572 +2.61128012e-03j  0.56266400 +2.51126072e-03j
  0.60126946 +2.38332777e-03j  0.63888238 +2.23129109e-03j
  0.67528491 +2.05929003e-03j  0.71026280 +1.87172242e-03j
  0.74360716 +1.67316840e-03j  0.77511620 +1.46831073e-03j
  0.80459690 +1.26185276e-03j  0.83186674 +1.05843556e-03j
  0.85675522 +8.62555824e-04j  0.87910538 +6.78485908e-04j
  0.89877520 +5.10197591e-04j  0.91563889 +3.61290940e-04j
  0.92958798 +2.34929634e-04j  0.94053235 +1.33783983e-04j
  0.94840102 +5.99827260e-05j  0.95314282 +1.50745394e-05j
  0.95472686 -1.22818422e-15j  0.95314282 +1.50745394e-05j
  0.94840102 +5.99827260e-05j  0.94053235 +1.33783983e-04j
  0.92958798 +2.34929634e-04j  0.91563889 +3.61290940e-04j
  0.89877520 +5.10197591e-04j  0.87910538 +6.78485908e-04j
  0.85675522 +8.62555824e-04j  0.83186674 +1.05843556e-03j
  0.80459690 +1.26185276e-03j  0.77511620 +1.46831073e-03j
```

```
  0.74360716 +1.67316840e-03j  0.71026280 +1.87172242e-03j
  0.67528491 +2.05929003e-03j  0.63888238 +2.23129109e-03j
  0.60126946 +2.38332777e-03j  0.56266400 +2.51126072e-03j
  0.52328572 +2.61128012e-03j  0.48335447 +2.67997067e-03j
  0.44308855 +2.71436932e-03j  0.40270312 +2.71201492e-03j
  0.36240862 +2.67098903e-03j  0.32240932 +2.58994742e-03j
  0.28290195 +2.46814170e-03j  0.24407447 +2.30543128e-03j
  0.20610493 +2.10228518e-03j  0.16916044 +1.85977438e-03j
  0.13339633 +1.57955467e-03j  0.09895542 +1.26384077e-03j
  0.06596736 +9.15372202e-04j  0.03454826 +5.37371863e-04j
  0.00480026 +1.33497988e-04j -0.02318854 -2.92209389e-04j
 -0.04934425 -7.35386532e-04j -0.07360706 -1.19140758e-03j
 -0.09593108 -1.65544854e-03j -0.11628406 -2.12255259e-03j
 -0.13464696 -2.58769571e-03j -0.15101354 -3.04585157e-03j
 -0.16538970 -3.49205478e-03j -0.17779294 -3.92146159e-03j
 -0.18825159 -4.32940727e-03j -0.19680411 -4.71145943e-03j
 -0.20349827 -5.06346681e-03j -0.20839036 -5.38160280e-03j
 -0.21154435 -5.66240364e-03j -0.21303101 -5.90280069e-03j
 -0.21292711 -6.10014682e-03j -0.21131454 -6.25223681e-03j
 -0.20827950 -6.35732164e-03j -0.20391169 -6.41411704e-03j
 -0.19830352 -6.42180629e-03j -0.19154940 -6.38003759e-03j
 -0.18374499 -6.28891640e-03j -0.17498654 -6.14899304e-03j
 -0.16537032 -5.96124608e-03j -0.15499198 -5.72706184e-03j
 -0.14394605 -5.44821063e-03j -0.13232548 -5.12682011e-03j
 -0.12022121 -4.76534628e-03j -0.10772178 -4.36654268e-03j
 -0.09491302 -3.93342813e-03j -0.08187777 -3.46925357e-03j
 -0.06869564 -2.97746844e-03j -0.05544287 -2.46168685e-03j
 -0.04219213 -1.92565411e-03j -0.02901242 -1.37321384e-03j
 -0.01596906 -8.08275859e-04j -0.00312361 -2.34785364e-04j
  0.00946612 +3.43306710e-04j  0.02174607 +9.22071732e-04j
  0.03366581 +1.49762982e-03j  0.04517855 +2.06617452e-03j
  0.05624101 +2.62399578e-03j  0.06681332 +3.16750096e-03j
  0.07685889 +3.69323421e-03j  0.08634435 +4.19789388e-03j
  0.09523935 +4.67834830e-03j  0.10351645 +5.13164985e-03j
  0.11115101 +5.55504741e-03j  0.11812104 +5.94599731e-03j
  0.12440706 +6.30217293e-03j  0.12999204 +6.62147287e-03j
  0.13486119 +6.90202805e-03j  0.13900194 +7.14220758e-03j
  0.14240381 +7.34062376e-03j  0.14505836 +7.49613600e-03j
  0.14695907 +7.60785406e-03j  0.14810132 +7.67514038e-03j
  0.14848238 +7.69761176e-03j]
```

In [448]:
```
M_test = 800 * 400
x_test = np.ones((M_test, 3))
n = 0
for i in range(-400, 400):
    for j in range(0, 400):
        x_test[n] = np.array([i / 100, j / 100, 0])
        n += 1
print(x_test)
ZZ_test = np.vectorize(complex)(np.zeros((M_test, L)))
for m in range(M_test):
    for l in range(L):
        try:
            ZZ_test[m, l] = Z(x_test[m], y[l], omega)
        except:
            ZZ_test[m, l] = 0
print(ZZ_test)
p_test2 = np.dot(ZZ_test, q_PM.T)
print('>>', p_test2, p_test2.shape)
print('\n\n>>', x_test, y)
```

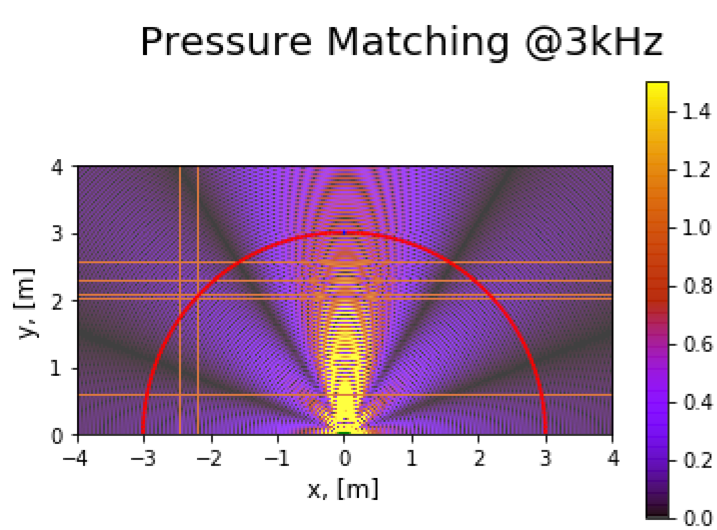
```
[[-4.    0.    0.  ]
 [-4.    0.01  0.  ]
 [-4.    0.02  0.  ]
 ...,
 [ 3.99  3.97  0.  ]
 [ 3.99  3.98  0.  ]
 [ 3.99  3.99  0.  ]]
[[-0.01415315-0.01448099j -0.01923294+0.00599694j -0.00338024+0.01975763j
  ...,  0.00027128-0.01974441j -0.01737877-0.00916799j
  -0.01598478+0.01125946j]
 [-0.01416322-0.01447105j -0.01922870+0.0060103j  -0.00336655+0.01975991j
  ...,  0.00025782-0.01974453j -0.01738494-0.00915617j
  -0.01597713+0.01127022j]
 [-0.01419341-0.01444117j -0.01921593+0.00605036j -0.00332548+0.01976667j
  ...,  0.00021743-0.01974483j -0.01740338-0.00912068j
  -0.01595412+0.01130245j]
 ...,
 [-0.00709451+0.01208554j -0.01359476+0.00354495j -0.01216731-0.00709489j
  ...,  0.01365331-0.00387155j  0.01249317+0.00680733j
   0.00419675+0.01363189j]
 [-0.00204135+0.0138471j  -0.01125075+0.00838551j -0.01392665-0.00198404j
  ...,  0.01114799-0.00875306j  0.01412564+0.00154013j
   0.00906690+0.010987j  ]
 [ 0.00330375+0.01358355j -0.00725792+0.01198883j -0.01362893+0.00341299j
  ...,  0.00697658-0.01231724j  0.01363005-0.00395169j
   0.01255904+0.00668404j]]
>> [-0.01115970+0.11094035j -0.01108295+0.11094179j -0.01085274+0.1109458j
 ..., -0.11057798-0.01391259j -0.10735617+0.02884685j
 -0.08822202+0.06715735j] (320000,)


>> [[-4.    0.    0.  ]
 [-4.    0.01  0.  ]
 [-4.    0.02  0.  ]
```

```
 ...,
 [ 3.99  3.97  0.  ]
 [ 3.99  3.98  0.  ]
 [ 3.99  3.99  0.  ]] [[-0.07  0.    0.  ]
 [-0.05  0.    0.  ]
 [-0.03  0.    0.  ]
 [-0.01  0.    0.  ]
 [ 0.01  0.    0.  ]
 [ 0.03  0.    0.  ]
 [ 0.05  0.    0.  ]
 [ 0.07  0.    0.  ]]
```

In [449]:

```
output = np.ones((400,800))
for n in range(x_test.shape[0]):
    output[399 - int(x_test[n, 1]*100), int(x_test[n, 0]*100+400)] = p_test2[n]
fig = plt.figure()
plt.scatter(np.concatenate((x, y))[:, 0], np.concatenate((x, y))[:, 1], c=['r']*(l
implot = plt.imshow(np.abs(output), interpolation='nearest', cmap='gnuplot', exte
plt.colorbar()
fig.suptitle("Pressure Matching @3kHz", fontsize=20)
plt.xlabel("x, [m]", fontsize=12)
plt.ylabel("y, [m]", fontsize=12)
plt.show()
```

```
C:\Users\sarki\Anaconda3\lib\site-packages\ipykernel_launcher.py:3: ComplexWarn
ing: Casting complex values to real discards the imaginary part
  This is separate from the ipykernel package so we can avoid doing imports unt
il
```



In [372]:

```
print(q_DAS, q_PM, np.average(q_DAS / q_PM))
```

```
[[-4.42774546+1.61111273j -4.42310489+1.62357719j -4.43059043+1.60341753j
  -4.44975458+1.55046996j -4.47941986+1.46432173j -4.51766439+1.34435495j
  -4.56180116+1.18981816j -4.60835469+0.99992665j]] [-5.46750534+1.52310267j -
3.24539183+1.09455475j -3.74183017+1.28210136j
 -5.03568418+1.6294137j  -5.03568418+1.6294137j  -3.74183017+1.28210136j
 -3.24539183+1.09455475j -5.46750534+1.52310267j] (1.07095055903+0.006224611498
36j)
```

Acoustic Contract Maximization Formulas

$$E_B = \frac{1}{\rho_0 c^2} \mid p_B \mid^2 = \frac{1}{\rho_0 c^2} \mathbf{q}^H \mathbf{R}_B \mathbf{q}$$

$$E_D = \frac{1}{\rho_0 c^2} \frac{1}{M_D} \mid\mid \mathbf{p}_D \mid\mid^2 = \frac{1}{\rho_0 c^2} \mathbf{q}^H \mathbf{R}_D \mathbf{q}$$

$$\mathbf{R}_B = \mathbf{z}_B^H \mathbf{z}_B$$

$$\mathbf{R}_D = \mathbf{Z}_D^H \mathbf{Z}_D / M_D$$

$$AC = \frac{E_B}{E_D} = \frac{\mathbf{q}^H \mathbf{R}_B \mathbf{q}}{\mathbf{q}^H \mathbf{R}_D \mathbf{q}}$$

$$\max_{\mathbf{q}} J_{ACM} = \max_{\mathbf{q}} \frac{\mathbf{q}^H \mathbf{R}_B \mathbf{q}}{\mathbf{q}^H \mathbf{R}_D \mathbf{q}}$$

$$\mathbf{q}_{ACM} = \Phi(\mathbf{R}_D^{-1} \mathbf{R}_B)$$

However, $\mathbf{R}_D^{-1}$ is rather unsavory and often ill-conditioned to inversion. Rather, we would use

$$\mathbf{q}_{ACM} = \Phi\Big((\mathbf{R}_D + \beta_{ACM} \mathbf{I})^{-1} \mathbf{R}_B\Big)$$

where $\Phi(\mathbf{A})$ is the eigenvector corresponding to the maximum eigenvalue of matrix $\mathbf{A}$.

In [450]:
```
E_ref = (4 * pi * R)**2 / L
sigma = 25
epsilon_beta, beta_min, E_max = 10e-9, 10e-20, sigma * E_ref
p_b_hat = 1
p_hat = np.array([0]*(M//2) + [1] + [0]*(M//2))

W = lambda q: p_b_hat / np.dot(z_b.T, q)
E = lambda q: np.dot(np.conjugate(q).T, q)
def PHI(A):
    w, v = np.linalg.eig(A)
    i = np.argmax(w)
    return v[:, i]

M_D = M - 1
R_D = np.dot(np.conjugate(Z_d).T, Z_d) / (M-1)
R_B = np.dot(np.conjugate(z_b).T, z_b)

print(np.conjugate(z_b).T, z_b)

print(R_B)

print(type(R_D), type(R_B))

print(R_D.shape, R_B.shape)
```

```
[  2.40957934e-03+0.02641975j   2.48359946e-03+0.02641335j
   2.36391255e-03+0.0264236j    2.05033734e-03+0.02644786j
   1.54249351e-03+0.02647924j   8.40054803e-04+0.02650656j
  -5.68415159e-05+0.02651437j  -1.14703023e-03+0.02648299j] [  2.40957934e-03-
0.02641975j   2.48359946e-03-0.02641335j
   2.36391255e-03-0.0264236j    2.05033734e-03-0.02644786j
   1.54249351e-03-0.02647924j   8.40054803e-04-0.02650656j
  -5.68415159e-05-0.02651437j  -1.14703023e-03-0.02648299j]
(0.00562764336198+0j)
<class 'numpy.ndarray'> <class 'numpy.complex128'>
(8, 8) ()
```

In [452]:
```python
### Calculate ACM Regularization Parameter ###

beta = beta_min
print('a')
q_temp = PHI(np.dot(np.linalg.inv(R_D + beta * np.eye(L)), R_B))
print('b')
q_hat = W(q_temp) * q_temp
n = 0
while E(q_hat) > E_max:
    if n % 1000 == 0:
        print(beta, E(q_hat), E_max, E(q_hat) > E_max)
    beta += epsilon_beta
    q_temp = PHI(np.dot(np.linalg.inv(R_D + beta * np.eye(L)), R_B))
    q_hat = W(q_temp) * q_temp
    n += 1

print(beta, E(q_hat), E_max, E(q_hat) > E_max)
beta_ACM = beta
```

```
---------------------------------------------------------------------------
KeyboardInterrupt                         Traceback (most recent call last)
<ipython-input-452-7d27047b8722> in <module>()
     11         print(beta, E(q_hat), E_max, E(q_hat) > E_max)
     12     beta += epsilon_beta
---> 13     q_temp = PHI(np.dot(np.linalg.inv(R_D + beta * np.eye(L)), R_B))
     14     q_hat = W(q_temp) * q_temp
     15     n += 1

<ipython-input-450-61d4077aecaa> in PHI(A)
      8 E = lambda q: np.dot(np.conjugate(q).T, q)
      9 def PHI(A):
---> 10     w, v = np.linalg.eig(A)
     11     i = np.argmax(w)
     12     return v[:, i]
```

In [410]:
```python
q_ACM = PHI(np.dot(np.linalg.inv(R_D), R_B))
print(q_ACM, q_DAS, q_DAS.real / q_ACM.real)
```

```
[-0.03056138 +3.46882534e-05j  0.15525338 +6.72881479e-06j
 -0.38045526 -1.33860403e-04j  0.57467685 +0.00000000e+00j
 -0.57465195 +5.41708986e-04j  0.38040639 -9.41969248e-04j
 -0.15522097 +7.24904088e-04j  0.03055282 -2.36279763e-04j] [[-4.42774546+1.611
11273j -4.42310489+1.62357719j -4.43059043+1.60341753j
  -4.44975458+1.55046996j -4.47941986+1.46432173j -4.51766439+1.34435495j
  -4.56180116+1.18981816j -4.60835469+0.99992665j]] [[ 144.88042254  -28.489587
71   11.6454965    -7.74305515    7.79501369
   -11.87588991   29.3890781  -150.8323745 ]]
```

In [413]:
```
M_test = 800 * 400
x_test = np.ones((M_test, 3))
n = 0
for i in range(-400, 400):
    for j in range(0, 400):
        x_test[n] = np.array([i / 100, j / 100, 0])
        n += 1
print(x_test)
ZZ_test = np.vectorize(complex)(np.zeros((M_test, L)))
for m in range(M_test):
    for l in range(L):
        try:
            ZZ_test[m, l] = Z(x_test[m], y[l], omega)
        except:
            ZZ_test[m, l] = 0
print(ZZ_test)
p_test = np.dot(ZZ_test, q_ACM.T)
print('>>', p_test, p_test.shape)
print('\n\n>>', x_test, y)
```

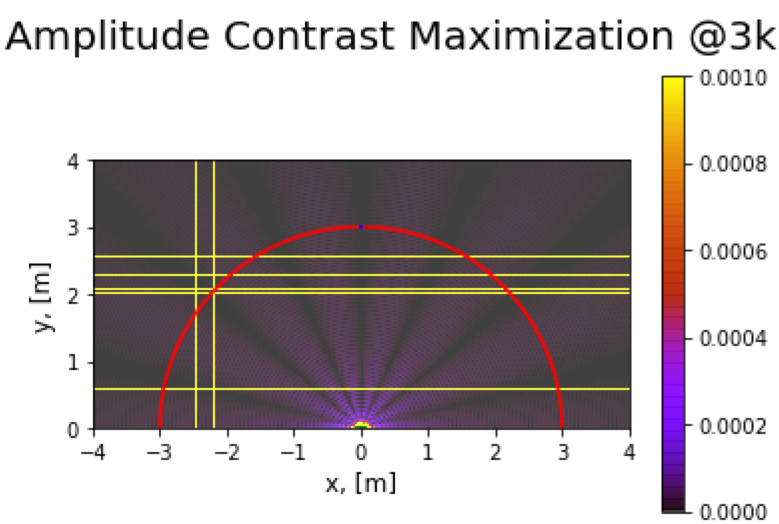
```
[[-4.    0.    0.  ]
 [-4.    0.01  0.  ]
 [-4.    0.02  0.  ]
 ...,
 [ 3.99  3.97  0.  ]
 [ 3.99  3.98  0.  ]
 [ 3.99  3.99  0.  ]]
[[ -1.22167916e-02 +1.61480861e-02j   8.94701020e-03 +1.80504902e-02j
    2.00446263e-02 +5.55163640e-05j ...,  -1.94300831e-02 +3.51952905e-03j
   -5.50079608e-03 +1.88630578e-02j   1.43501319e-02 +1.32801509e-02j]
 [ -1.22053688e-02 +1.61566395e-02j   8.95963577e-03 +1.80441545e-02j
    2.00445965e-02 +4.15323070e-05j ...,  -1.94275999e-02 +3.53287071e-03j
   -5.48787841e-03 +1.88667576e-02j   1.43591223e-02 +1.32703426e-02j]
 [ -1.21710649e-02 +1.61822508e-02j   8.99748531e-03 +1.80250947e-02j
    2.00444488e-02 -4.19159442e-07j ...,  -1.94200953e-02 +3.57288492e-03j
   -5.44911071e-03 +1.88778040e-02j   1.43860533e-02 +1.32408811e-02j]
 ...,
 [  1.29958148e-02 -5.24415624e-03j   1.28983411e-02 +5.56927750e-03j
    5.14539576e-03 +1.31112826e-02j ...,  -1.30084970e-02 -5.67280682e-03j
   -5.25325409e-03 -1.32220353e-02j   5.57023837e-03 -1.31306361e-02j]
 [  1.00380666e-02 -9.75431055e-03j   1.40294079e-02 +2.68074806e-04j
    9.72632718e-03 +1.01629990e-02j ...,  -1.41711966e-02 -2.66411240e-04j
   -9.90793098e-03 -1.01852206e-02j   9.69116900e-05 -1.42447708e-02j]
 [  5.59438307e-03 -1.28113377e-02j   1.30667884e-02 -5.06640627e-03j
    1.28436598e-02 +5.69532832e-03j ...,  -1.31766618e-02 +5.17326929e-03j
   -1.30416426e-02 -5.59549850e-03j  -5.38376418e-03 -1.31689423e-02j]]
>> [  1.47576456e-05 -2.61536238e-06j   1.47538823e-05 -2.62518670e-06j
   1.47425539e-05 -2.65463586e-06j ...,   3.76992311e-06 +3.07837368e-06j
   4.72108447e-06 +1.43089564e-06j   4.97513263e-06 -4.82560764e-07j] (320000,)


>> [[-4.    0.    0.  ]
 [-4.    0.01  0.  ]
 [-4.    0.02  0.  ]
```

```
 ...,
 [ 3.99  3.97  0.  ]
 [ 3.99  3.98  0.  ]
 [ 3.99  3.99  0.  ]] [[-0.07  0.    0.  ]
 [-0.05  0.    0.  ]
 [-0.03  0.    0.  ]
 [-0.01  0.    0.  ]
 [ 0.01  0.    0.  ]
 [ 0.03  0.    0.  ]
 [ 0.05  0.    0.  ]
 [ 0.07  0.    0.  ]]
```

In [418]:

```
output = np.ones((400,800))
for n in range(x_test.shape[0]):
    output[399 - int(x_test[n, 1]*100), int(x_test[n, 0]*100+400)] = p_test[n]
print(np.max(output))
fig = plt.figure()
plt.scatter(np.concatenate((x, y))[:, 0], np.concatenate((x, y))[:, 1], c=['r']*(
implot = plt.imshow(np.abs(output), interpolation='nearest', cmap='gnuplot', vmin
plt.colorbar()
fig.suptitle("Amplitude Contrast Maximization @3kHz", fontsize=20)
plt.xlabel("x, [m]", fontsize=12)
plt.ylabel("y, [m]", fontsize=12)
plt.show()
```

```
C:\Users\sarki\Anaconda3\lib\site-packages\ipykernel_launcher.py:3: ComplexWarn
ing: Casting complex values to real discards the imaginary part
  This is separate from the ipykernel package so we can avoid doing imports unt
il

2.33394760509
```



The above plot makes no sense, likely because it omits the regularization parameter β_{ACM}. This cements the importance of regularization

Equations for Energy Difference

Maximization

$$\max_{\mathbf{q}} \frac{E_B - \alpha E_D}{E_{\mathbf{q}}} = \max_{\mathbf{q}} \frac{\mathbf{q}^H(\mathbf{R}_B - \alpha\mathbf{R}_D)\mathbf{q}}{\mathbf{q}^H\mathbf{q}}$$

$$\mathbf{q}_{EDM} = \Phi(\mathbf{R}_B - \alpha\mathbf{R}_D)$$

In [384]:

```
E_ref = (4 * pi * R)**2 / L
sigma = 1.05
epsilon_alpha, alpha_max, E_max = 1e2, 1e6, sigma * E_ref
p_b_hat = 1
p_hat = np.array([0]*(M//2) + [1] + [0]*(M//2))

W = lambda q: p_b_hat / np.dot(z_b.T, q)
E = lambda q: np.dot(np.conjugate(q).T, q)
def PHI(A):
    w, v = np.linalg.eig(A)
    i = np.argmax(w)
    return v[:, i]

M_D = M - 1
R_D = np.dot(np.conjugate(Z_d).T, Z_d).real / (M_D)
R_B = np.dot(np.conjugate(z_b).T, z_b).real

print(M_D, R_D, R_B)
```

```
180 [[  7.03820003e-04   5.03863000e-04   7.16140042e-05  -2.45788624e-04
   -2.36252144e-04   4.55922445e-06   1.96757577e-04   1.57841092e-04]
 [  5.03863000e-04   7.03684463e-04   5.03774310e-04   7.15699207e-05
   -2.45807129e-04  -2.36245596e-04   4.59739389e-06   1.96806324e-04]
 [  7.16140042e-05   5.03774310e-04   7.03611504e-04   5.03730421e-04
    7.15322220e-05  -2.45847472e-04  -2.36260054e-04   4.63597547e-06]
 [ -2.45788624e-04   7.15699207e-05   5.03730421e-04   7.03601088e-04
    5.03731310e-04   7.15008879e-05  -2.45909677e-04  -2.36295526e-04]
 [ -2.36252144e-04  -2.45807129e-04   7.15322220e-05   5.03731310e-04
    7.03653209e-04   5.03776978e-04   7.14759017e-05  -2.45993776e-04]
 [  4.55922445e-06  -2.36245596e-04  -2.45847472e-04   7.15008879e-05
    5.03776978e-04   7.03767895e-04   5.03867449e-04   7.14572501e-05]
 [  1.96757577e-04   4.59739389e-06  -2.36260054e-04  -2.45909677e-04
    7.14759017e-05   5.03867449e-04   7.03945207e-04   5.04002770e-04]
 [  1.57841092e-04   1.96806324e-04   4.63597547e-06  -2.36295526e-04
   -2.45993776e-04   7.14572501e-05   5.04002770e-04   7.04185240e-04]] 0.00562
764336198
```

In [385]:
```
### Calculate alpha ###

alpha = alpha_max
q_temp = PHI(R_B - alpha * R_D)
q_hat = W(q_temp) * q_temp
n = 0
while E(q_hat) > E_max and alpha >= 0:
    if n % 1000:
        print(alpha, E(q_hat), E_max, E(q_hat) > E_max)
    alpha = alpha - epsilon_alpha
    q_temp = PHI(R_B - alpha * R_D)
    q_hat = W(q_temp) * q_temp
    n += 1

print(alpha, E(q_hat), E_max, E(q_hat) > E_max)
alpha_EDM = alpha
```

```
80300.0 (202233112841+0j) 186.53552318058888 True
80200.0 (202233112834+0j) 186.53552318058888 True
80100.0 (202233112813+0j) 186.53552318058888 True
80000.0 (202233112797+0j) 186.53552318058888 True
79900.0 (202233112779+0j) 186.53552318058888 True
79800.0 (202233112773+0j) 186.53552318058888 True
79700.0 (202233112747+0j) 186.53552318058888 True
79600.0 (202233112737+0j) 186.53552318058888 True
79500.0 (202233112719+0j) 186.53552318058888 True
79400.0 (202233112698+0j) 186.53552318058888 True
79300.0 (202233112692+0j) 186.53552318058888 True
79200.0 (202233112673+0j) 186.53552318058888 True
79100.0 (202233112651+0j) 186.53552318058888 True
79000.0 (202233112635+0j) 186.53552318058888 True
78900.0 (202233112631+0j) 186.53552318058888 True
78800.0 (202233112603+0j) 186.53552318058888 True
78700.0 (202233112596+0j) 186.53552318058888 True
78600.0 (202233112566+0j) 186.53552318058888 True
78500.0 (202233112561+0j) 186.53552318058888 True
78400.0 (202233112542+0j) 186.53552318058888 True
```

In [453]:
```
q_EDM = PHI(R_D - alpha * R_B)
print(q_EDM, q_DAS, np.average(q_DAS / q_EDM))
p = np.dot(ZZ, q_EDM.T)
print(p)
print(np.max(p))
```

```
[-0.41100165-0.01978789j -0.28569511+0.03771158j  0.11410313+0.08442244j
  0.46737839+0.07169087j  0.47335254+0.j          0.12701134-0.07695935j
 -0.27611268-0.09927769j -0.41072655-0.05482335j] [[ 0.42795313+4.69227772j  0.
44108448+4.69098116j  0.41985128+4.69305916j
   0.36421000+4.6980447j   0.27406334+4.70471356j  0.14930517+4.71108159j
  -0.01010674+4.71440292j -0.20405013+4.71117287j]] (0.239559987472+2.246323529
83j)
[  9.46335080e-03-0.05666284j   9.46404126e-03-0.05667323j
   9.46610277e-03-0.05670436j   9.46950572e-03-0.05675607j
   9.47420084e-03-0.05682807j   9.48011927e-03-0.05692001j
   9.48717269e-03-0.05703137j   9.49525349e-03-0.05716158j
   9.50423500e-03-0.05730991j   9.51397171e-03-0.05747554j
   9.52429964e-03-0.05765754j   9.53503664e-03-0.05785486j
   9.54598288e-03-0.05806633j   9.55692129e-03-0.05829066j
   9.56761816e-03-0.05852647j   9.57782377e-03-0.05877221j
   9.58727311e-03-0.05902627j   9.59568667e-03-0.05928687j
   9.60277138e-03-0.05955215j   9.60822158e-03-0.0598201j
   9.61172011e-03-0.06008861j   9.61293954e-03-0.06035546j
   9.61154345e-03-0.06061829j   9.60718786e-03-0.06087465j
   9.59952276e-03-0.06112197j   9.58819373e-03-0.06135761j
   9.57284371e-03-0.06157878j   9.55311479e-03-0.06178265j
   9.52865022e-03-0.06196627j   9.49909643e-03-0.06212664j
   9.46410512e-03-0.06226069j   9.42333552e-03-0.0623653j
   9.37645663e-03-0.0624373j    9.32314955e-03-0.0624735j
   9.26310984e-03-0.06247071j   9.19604994e-03-0.06242572j
   9.12170154e-03-0.06233538j   9.03981798e-03-0.06219655j
   8.95017661e-03-0.06200617j   8.85258114e-03-0.06176124j
   8.74686383e-03-0.0614589j    8.63288767e-03-0.06109638j
   8.51054842e-03-0.0606711j    8.37977654e-03-0.06018062j
   8.24053888e-03-0.05962273j   8.09284033e-03-0.05899541j
   7.93672514e-03-0.05829694j   7.77227812e-03-0.05752582j
   7.59962553e-03-0.05668088j   7.41893582e-03-0.05576128j
   7.23041999e-03-0.05476649j   7.03433177e-03-0.05369639j
   6.83096750e-03-0.05255121j   6.62066565e-03-0.0513316j
   6.40380618e-03-0.05003863j   6.18080948e-03-0.0486738j
   5.95213511e-03-0.04723906j   5.71828020e-03-0.04573681j
   5.47977762e-03-0.0441699j    5.23719386e-03-0.04254166j
   4.99112668e-03-0.04085586j   4.74220252e-03-0.03911672j
   4.49107370e-03-0.03732892j   4.23841542e-03-0.03549757j
   3.98492263e-03-0.03362816j   3.73130674e-03-0.03172662j
   3.47829216e-03-0.02979921j   3.22661288e-03-0.02785255j
   2.97700886e-03-0.02589355j   2.73022246e-03-0.02392939j
   2.48699482e-03-0.02196749j   2.24806231e-03-0.02001544j
   2.01415299e-03-0.018081j     1.78598310e-03-0.01617197j
   1.56425373e-03-0.01429626j   1.34964753e-03-0.0124617j
   1.14282554e-03-0.01067611j   9.44424216e-04-0.00894718j
   7.55052574e-04-0.00728241j   5.75289495e-04-0.00568909j
   4.05681238e-04-0.00417425j   2.46739113e-04-0.00274456j
   9.89373526e-05-0.00140634j  -3.72888181e-05-0.00016547j
  -1.61544923e-04+0.00097264j  -2.73478774e-04+0.00200309j
```

```
  -3.72781809e-04+0.0029215j   -4.59190239e-04+0.00372411j
  -5.32485994e-04+0.00440773j  -5.92497473e-04+0.00496983j
  -6.39100105e-04+0.00540851j  -6.72216711e-04+0.00572254j
  -6.91817689e-04+0.00591135j  -6.97920991e-04+0.00597505j
  -6.90591928e-04+0.00591442j  -6.69942780e-04+0.00573089j
  -6.36132220e-04+0.00542653j  -5.89364548e-04+0.00500405j
  -5.29888745e-04+0.00446674j  -4.57997330e-04+0.00381849j
  -3.74025038e-04+0.00306368j  -2.78347308e-04+0.00220725j
  -1.71378586e-04+0.00125454j  -5.35704455e-05+0.00021135j
   7.45904714e-05-0.00091617j   2.12584688e-04-0.00212155j
   3.59862297e-04-0.00339805j   5.15845525e-04-0.0047387j
   6.79931451e-04-0.00613638j   8.51494864e-04-0.00758385j
   1.02989125e-03-0.00907379j   1.21445990e-03-0.01059888j
   1.40452708e-03-0.01215183j   1.59940935e-03-0.01372545j
   1.79841684e-03-0.01531264j   2.00085667e-03-0.0169065j
   2.20603629e-03-0.01850033j   2.41326688e-03-0.02008767j
   2.62186666e-03-0.02166235j   2.83116419e-03-0.02321849j
   3.04050153e-03-0.02475056j   3.24923737e-03-0.02625339j
   3.45674993e-03-0.02772216j   3.66243978e-03-0.02915246j
   3.86573247e-03-0.03054028j   4.06608092e-03-0.03188201j
   4.26296770e-03-0.03317446j   4.45590691e-03-0.03441485j
   4.64444605e-03-0.03560082j   4.82816739e-03-0.03673039j
   5.00668928e-03-0.037802j     5.17966712e-03-0.03881446j
   5.34679399e-03-0.03976699j   5.50780119e-03-0.04065911j
   5.66245832e-03-0.04149073j   5.81057326e-03-0.04226207j
   5.95199183e-03-0.04297365j   6.08659717e-03-0.04362627j
   6.21430902e-03-0.04422099j   6.33508267e-03-0.04475914j
   6.44890776e-03-0.04524223j   6.55580692e-03-0.04567198j
   6.65583422e-03-0.04605031j   6.74907346e-03-0.04637926j
   6.83563643e-03-0.04666101j   6.91566096e-03-0.04689786j
   6.98930892e-03-0.04709219j   7.05676421e-03-0.04724647j
   7.11823065e-03-0.04736321j   7.17392981e-03-0.04744496j
   7.22409897e-03-0.04749429j   7.26898890e-03-0.04751377j
   7.30886180e-03-0.04750598j   7.34398924e-03-0.04747346j
   7.37465011e-03-0.04741873j   7.40112865e-03-0.04734425j
   7.42371258e-03-0.04725245j   7.44269127e-03-0.04714567j
   7.45835399e-03-0.04702622j   7.47098825e-03-0.04689629j
   7.48087826e-03-0.04675801j   7.48830347e-03-0.04661343j
   7.49353715e-03-0.0464645j    7.49684517e-03-0.04631308j
   7.49848479e-03-0.04616093j   7.49870358e-03-0.04600971j
   7.49773844e-03-0.04586098j   7.49581465e-03-0.04571621j
   7.49314512e-03-0.04557674j   7.48992958e-03-0.04544385j
   7.48635397e-03-0.04531868j   7.48258984e-03-0.04520228j
   7.47879382e-03-0.0450956j    7.47510716e-03-0.04499949j
   7.47165538e-03-0.0449147j    7.46854790e-03-0.04484186j
   7.46587775e-03-0.04478153j   7.46372136e-03-0.04473414j
   7.46213836e-03-0.04470005j   7.46117146e-03-0.04467949j
   7.46084631e-03-0.04467262j]
(0.0096129395363-0.0603554565737j)
```

I gave up on EDM after seeing this maximum value. I need to get the regularization parameters figured out.

In [ ]: