File - performance_test			
	C:\Users\usuario\Anaconda3\envs\tfq\python.exe		quantumGAN/performance_test.py
	[0.83809885 0.59662554 0.08574274 0.23931475 0 Epoch 0: Loss: [-0.51708544] [0.45231102 0.	.13270372 0.67136706] 0.4109069 0.] [0.40039062 0.37158203 0.17626953 0.05175781]
	[0.10325458] [0.10480057]	ט 4טערטדע.ט.] [0.1/5/81] C.5/130203 0.1/020733 0.031/5/81]
	Epoch 1: Loss: [-0.41802769] [0.44482384 0.	0.41277293 0.] [0.40478516 0.33886719 0.19580078 0.06054688]
	[0.17808944] [0.18095745] Epoch 2: Loss: [-0.3584017] [0.45722264 0.	0.43477556 0.] [0.38671875 0.35205078 0.21240234 0.04882812]
8	[0.26187903] [0.26701365]		
	Epoch 3: Loss: [-0.32810519] [0.4421636 0. [0.33543319] [0.34206411]	0.44236231 0.] [0.39794922 0.35693359 0.19091797 0.05419922]
11	Epoch 4: Loss: [-0.3142884] [0.44340551 0.	0.47991339 0.] [0.39550781 0.36083984 0.19726562 0.04638672]
	[0.39006103] [0.39703704] Epoch 5: Loss: [-0.30841528] [0.45722264 0.	0.43477556 0.] [0.38867188 0.36132812 0.18994141 0.06005859]
	[0.42679137] [0.43382093]	0.43477330 0.	1 [0.3880/100 0.30132812 0.18//4141 0.000383/]
	Epoch 6: Loss: [-0.30592869] [0.4858715 0.	0.47115998 0.] [0.40722656 0.33642578 0.20166016 0.0546875]
	[0.4508546] [0.45786666] Epoch 7: Loss: [-0.30485919] [0.45231102 0.	0.4109069 0.] [0.38818359 0.34423828 0.21875 0.04882812]
	[0.46629032] [0.47322492]	0 /4/04/4/ 0	1 [0 (400(574 0 7054(0)) 0 40050507 0 05(40000]
	Epoch 8: Loss: [-0.30391934] [0.42334983 0. [0.47623022] [0.48198257]	0.41691616 0.] [0.41894531 0.32714844 0.19970703 0.05419922]
21	Epoch 9: Loss: [-0.30349532] [0.46876017 0.	0.4111282 0.] [0.40380859 0.34960938 0.20068359 0.04589844]
	[0.48281253] [0.48804613] Epoch 10: Loss: [-0.30314494] [0.42334983 0.	0.41691616 0.] [0.42285156 0.32666016 0.20117188 0.04931641
]		
	[0.48677289] [0.49139136] Epoch 11: Loss: [-0.30280091] [0.4421636 0.	0.44236231 0.] [0.39501953 0.33251953 0.21582031 0.05664062
		J.77200201 U.	. [0.07.001.00 0.00201.00 0.21002001 0.00004002
1	[0.48947248] [0.49339451] Epoch 12: Loss: [-0.30229636] [0.4456309 0.	0.40199548 0.] [0.39306641 0.35449219 0.20019531 0.05224609
		0.40177040 0.] [0.07300041 U.33447217 U.200173331 U.03224009
	[0.49181477] [0.49463434]	0 (5004770 0	1 [0 (04055/2 0 750700// 0 400/2/5/ 0 0540/057
1	Epoch 13: Loss: [-0.30212915] [0.44340551 0.]	0.47991339 0.] [0.40185547 0.35839844 0.18847656 0.05126953
	[0.49273141] [0.49518592]		1.5
	Epoch 14: Loss: [-0.30190912] [0.4456309 0.]	0.40199548 0.] [0.390625 0.34375 0.21484375 0.05078125
32	[0.49362745] [0.49559144]		
	Epoch 15: Loss: [-0.3015627] [0.4421636 0. [0.49440403] [0.49557966]	0.44236231 0.] [0.39941406 0.35302734 0.19921875 0.04833984]
	Epoch 16: Loss: [-0.30127617] [0.48331707 0.	0.49868168 0.] [0.40332031 0.34228516 0.19921875 0.05517578
7.4] [0.49496051] [0.49548149]		
	Epoch 17: Loss: [-0.30094585] [0.46876017 0.	0.4111282 0.] [0.40087891 0.34521484 0.20849609 0.04541016
7.0			
	[0.49573965] [0.49550759] Epoch 18: Loss: [-0.3002969] [0.48331707 0.	0.49868168 0.] [0.40917969 0.34521484 0.19287109 0.05273438]
	[0.49695639] [0.49523653]		1.5
	Epoch 19: Loss: [-0.30048304] [0.4471763 0.]	0.41560002 0.] [0.40722656 0.34033203 0.20361328 0.04882812
	[0.49619388] [0.49489402]		
1	Epoch 20: Loss: [-0.30017981] [0.4471763 0.]	0.41560002 0.] [0.40234375 0.35595703 0.19677734 0.04492188
44	[0.49674341] [0.49474773]		
	Epoch 21: Loss: [-0.29940173] [0.48331707 0.]	0.49868168 0.] [0.41357422 0.34814453 0.18701172 0.05126953
46	[0.49836712] [0.49458612]		
47	Epoch 22: Loss: [-0.29943368] [0.45722264 0.	0.43477556 0.] [0.39306641 0.33740234 0.21386719 0.05566406
48] [0.49751183] [0.49379174]		
	Epoch 23: Loss: [-0.29857057] [0.4421636 0.	0.44236231 0.] [0.43798828 0.33544922 0.18652344 0.04003906
] [0.49976548] [0.49406746]		
51	Epoch 24: Loss: [-0.29847003] [0.4421636 0.	0.44236231 0.] [0.43847656 0.31933594 0.19775391 0.04443359
] [0.49851762] [0.49256616]		
53	Epoch 25: Loss: [-0.29793837] [0.48548539 0.	0.43194116 0.] [0.41357422 0.33837891 0.19726562 0.05078125
1] [0.50014561] [0.49297802]		
	Epoch 26: Loss: [-0.29835464] [0.42334983 0.	0.41691616 0.] [0.41650391 0.33935547 0.20068359 0.04345703
	[0.49856246] [0.49234211] Epoch 27: Loss: [-0.2977076] [0.48548539 0.	0.43194116 0.] [0.39599609 0.36181641 0.19482422 0.04736328]
58	[0.49926399] [0.49154261]		
1	Epoch 28: Loss: [-0.29682537] [0.4456309 0.]	0.40199548 0.] [0.40966797 0.34814453 0.20068359 0.04150391
60	[0.50149353] [0.49174235]		1 to descend a service of
	Epoch 29: Loss: [-0.29606766] [0.45722264 0.]	0.43477556 0.] [0.41552734 0.33447266 0.20849609 0.04150391
62	[0.50271769] [0.49120771]		
	Epoch 30: Loss: [-0.29590153] [0.47960695 0.]	0.49233286 0.] [0.42480469 0.35595703 0.18408203 0.03515625
64	[0.50245838] [0.49055553]		
65	Epoch 31: Loss: [-0.29642237] [0.4421636 0.	0.44236231 0.] [0.43017578 0.33496094 0.18945312 0.04541016
] [0.50078302] [0.49007574]		
67	Epoch 32: Loss: [-0.29506462] [0.46876017 0.	0.4111282 0.] [0.41943359 0.34863281 0.19384766 0.03808594
] [0.50374254] [0.489892]		
	Epoch 33: Loss: [-0.29525821] [0.45231102 0.	0.4109069 0.] [0.42285156 0.33740234 0.19580078 0.04394531
_		Dona 1 of 2	

```
70 [0.50239094] [0.48897544]
71 Epoch 34: Loss: [-0.2949515] [0.4695876 0.
                                                       0.40827971 0.
                                                                           [0.41845703 0.34667969 0.19628906 0.03857422
 72 [0.5021059] [0.48796261]
73 Epoch 35: Loss: [-0.29319614] [0.4858715 0.
                                                       0.47115998 0.
                                                                            ] [0.43847656 0.34130859 0.17919922 0.
   041015621
 74 [0.50615083] [0.48793186]
75 Epoch 36: Loss: [-0.29441378] [0.45722264 0.
                                                       0.43477556 0.
                                                                            ] [0.42480469 0.35595703 0.17578125 0.
   04345703]
 76 [0.50220361] [0.48679296]
77 Epoch 37: Loss: [-0.29376369] [0.4471763 0. 03955078]
                                                       0.41560002 0.
                                                                            ] [0.41357422 0.36914062 0.17773438 0.
 78 [0.50315597] [0.48622852]
 79 Epoch 38: Loss: [-0.2932563] [0.44482384 0.
                                                       0.41277293 0.
                                                                           ] [0.42382812 0.36376953 0.16943359 0.04296875
 80 [0.50397811] [0.48586672]
81 Epoch 39: Loss: [-0.29278594] [0.45722264 0.
                                                       0 43477556 0
                                                                            1 [0.42529297 0.36865234 0.16455078 0.
   041503911
82 [0.5042344] [0.48501374]
83 Epoch 40: Loss: [-0.2906624] [0.4791745 0.
                                                     0.4725308 0.
                                                                       [0.42529297 0.37548828 0.16113281 0.03808594]
 84 [0.50867478] [0.48449249]
85 Epoch 41: Loss: [-0.29001426] [0.48548539 0.
                                                       0.43194116 0.
                                                                            ] [0.43701172 0.37109375 0.16015625 0.
   031738281
86 [0.50883978] [0.48311917]
87 Epoch 42: Loss: [-0.28925511] [0.4471763 0.
                                                       0.41560002 0.
                                                                            ] [0.41845703 0.37988281 0.15673828 0.
   04492188]
 88 [0.50953163] [0.48201328]
89 Epoch 43: Loss: [-0.28857524] [0.4858715 0.
                                                       0.47115998 0.
                                                                            ] [0.44482422 0.37109375 0.14501953 0.0390625
 90 [0.51065367] [0.48153068]
 91 Epoch 44: Loss: [-0.28878623] [0.4791745 0.
                                                     0.4725308 0.
                                                                        ] [0.42675781 0.38525391 0.14208984 0.04589844]
 92 [0.50906462] [0.48041737]
93 Epoch 45: Loss: [-0.28696625] [0.45231102 0.
                                                       0.4109069 0.
                                                                            ] [0.4296875 0.39355469 0.14404297 0.
   032714841
 94 [0.511743] [0.47878657]
95 Epoch 46: Loss: [-0.28653143] [0.46876017 0.
                                                       0.4111282 0.
                                                                            ] [0.44628906 0.37011719 0.14941406 0.
   03417969]
96 [0.51303215] [0.47905418]
97 Epoch 47: Loss: [-0.28757953] [0.4471763 0.
                                                       0.41560002 0.
                                                                            ] [0.44189453 0.38085938 0.14306641 0.
   034179691
98 [0.50918255] [0.47764298]
99 Epoch 48: Loss: [-0.28555609] [0.4695876 0.
                                                       0.40827971 0.
                                                                            ] [0.43847656 0.38525391 0.14355469 0.
   03271484]
100 [0.51441684] [0.4781176]
101 Epoch 49: Loss: [-0.2841952] [0.48548539 0.
                                                      0.43194116 0.
                                                                           ] [0.44775391 0.40136719 0.12109375 0.02978516
102 [0.51298653] [0.47337236]
103 Timer unit: 1e-07 s
104
105 Total time: 1408.91 s
106 File: C:\Users\usuario\qGAN\quantumGAN\quantum_qenerator.py
107 Function: train_mini_batch at line 129
108
109 Line #
                            Time Per Hit % Time Line Contents
111
      129
                                                      def train_mini_batch(self):
                          9456.0
                                               0.0
112
      130
                100
                                     94.6
                                                          nabla_theta = np.zeros(self.parameter_values.shape)
                                                          new_images = []
      131
               100
                         1534.0
113
                                    15.3
                                              0.0
      132
114
115
      133
                850
                         35391.0
                                     41.6
                                               0.0
                                                           for _, noise in self.mini_batch:
      134
               5250
                        122553.0
                                                              for index in range(len(self.parameter_values)):
116
                                     23.3
                                               0.0
117
      135
               4500
                        381119.0
                                     84.7
                                               0.0
                                                                  perturbation_vector = np.zeros(len(self.
   parameter_values))
118
                        107277.0
                                     23.8
                                                                  perturbation_vector[index] = 1
      136
               4500
                                               0.0
119
      137
               4500
                        434522.0
120
      138
                                     96.6
                                               0.0
                                                                   pos_params = self.parameter_values + (np.pi / 4) *
   perturbation_vector
121
      139
               4500
                       285990.0
                                     63.6
                                               0.0
                                                                  neg_params = self.parameter_values - (np.pi / 4) *
   perturbation_vector
122
      140
               4500 6513287643.0 1447397.3
123
      141
                                                                  pos_result = self.get_output(noise, params=pos_params)
124
      142
               4500 6481559230.0 1440346.5
                                                                   neg_result = self.get_output(noise, params=neg_params)
                                               46.0
125
      143
               4500
                       4769147.0
                                   1059.8
126
      144
                                               0.0
                                                                  pos_result = self.discriminator.predict(pos_result)
                                                                  neg_result = self.discriminator.predict(neg_result)
               4500
      145
                       2344015.0
127
                                    520.9
                                               0.0
                                                                  gradient = self.BCE(pos_result, np.array([1.])) - self.
               4500
                       2283285.0
                                    507.4
128
      146
                                               0.0
   BCE(neg_result, np.array([1.]))
               4500
      147
129
                       811930.0
                                   180.4
                                               0.0
                                                                  nabla_theta[index] += gradient
                750 1082594178.0 1443458.9
130
      148
                                               7.7
                                                              new_images.append(self.get_output(noise))
131
      149
                700
132
      150
                         11500.0
                                               0.0
                                                           for index in range(len(self.parameter_values)):
                                     16.4
133
      151
                600
                         22507.0
                                     37.5
                                                              self.parameter_values[index] -= (self.learning_rate / self.
                                               0.0
   mini_batch_size) * nabla_theta[index]
134
      152
                100
                         13418.0
135
      153
                                   134.2
                                               0.0
                                                           self.mini_batch = [(datapoint[0], fake_image) for datapoint,
   fake_image in zip(self.mini_batch, new_images)]
136
137
```

File - performance_test 138 Process finished with exit code 0 139