Был выполнен 2 вариант.

Функция для перемешивания данных:

```
def shuffle_data(data, label):
    tmp_data = []
    for i in range(len(data)):
        tmp_data.append([data[i], label[i]])
    random.shuffle(tmp_data)
    res_data = []
    res_label = []
    for i in range(len(data)):
        res_data.append(tmp_data[i][0])
        res_label.append(tmp_data[i][1])
    return np.asarray(res_data), np.asarray(res_label)
```

Функция для деления данных на обучающую, валидационную и тестовую выборки:

Генерация, перемешивание и разбиение данных:

```
x, y = generator.gen_data()
print(x.shape)
y = np.asarray([[0.] if item == 'Empty' else [1.] for item in y])
x, y = shuffle_data(x, y)
(train_x, val_x, test_x), (train_y, val_y, test_y) = split_data(x, y)
```

Параметры для построения и обучения нейросети:

```
batch_size = 32
num_epochs = 10

kernel_size = 3
pool_size = 2
conv_depth_1 = 32
conv_depth_2 = 64
drop_prob_1 = 0.25
hidden size = 512
```

Архитектура нейросети:

```
inp = Input(shape=(50, 50, 1))

conv_1 = Convolution2D(conv_depth_1, (kernel_size, kernel_size),
    padding='same', activation='relu', data_format='channels_last')(inp)
    conv_2 = Convolution2D(conv_depth_1, (kernel_size, kernel_size),
    padding='same', activation='relu', data_format='channels_last')(conv_1)
    pool_1 = MaxPooling2D(pool_size=(pool_size, pool_size))(conv_2)
    drop_1 = Dropout(drop_prob_1)(pool_1)

conv_3 = Convolution2D(conv_depth_2, (kernel_size, kernel_size),
    padding='same', activation='relu', data_format='channels_last')(drop_1)
    conv_4 = Convolution2D(conv_depth_2, (kernel_size, kernel_size),
    padding='same', activation='relu', data_format='channels_last')(conv_3)
    pool_2 = MaxPooling2D(pool_size=(pool_size, pool_size))(conv_4)
    drop_2 = Dropout(drop_prob_1)(pool_2)

flat = Flatten()(drop_2)
    hidden = Dense(hidden_size, activation='relu')(flat)
    drop_3 = Dropout(drop_prob_1)(hidden)
    out = Dense(1, activation='sigmoid')(drop_3)
```

Обучение и тестирование модели

Результат:

```
Epoch 6/10
10/10 [============= ] - 1s 144ms/step - loss: 0.2376
- accuracy: 0.8915 - val_loss: 0.7385 - val_accuracy: 0.8000
Epoch 7/10
10/10 [============= ] - 1s 144ms/step - loss: 0.1378
- accuracy: 0.9552 - val loss: 0.5580 - val accuracy: 0.8500
Epoch 8/10
10/10 [============= ] - 2s 155ms/step - loss: 0.1035
- accuracy: 0.9501 - val loss: 0.4214 - val accuracy: 0.8750
Epoch 9/10
- accuracy: 0.9704 - val loss: 0.3704 - val accuracy: 0.9250
Epoch 10/10
10/10 [============ ] - 1s 142ms/step - loss: 0.0306
- accuracy: 0.9876 - val loss: 0.4882 - val accuracy: 0.9500
accuracy: 0.9200
```

В результате точность на тренировочных данных 98.76%, на валидационных – 95%, на тестовых – 92%.