

```

import numpy as np
import matplotlib.pyplot as plt

#  $\hbar = 1$ , Energy levels:  $E_n = (n + 0.5) * \omega$ 
omegas = np.linspace(0.1, 5.0, 200) # frequencies
n_levels = 5 # energy levels  $n = 0$  to 4

# Build dataset: For each omega, calculate E0 to E4
X = []
y = []

for omega in omegas:
    for n in range(n_levels):
        X.append([omega, n]) # input: frequency & energy level index
        E_n = (n + 0.5) * omega
        noise = np.random.normal(0, 0.05) # small noise
        y.append(E_n + noise)

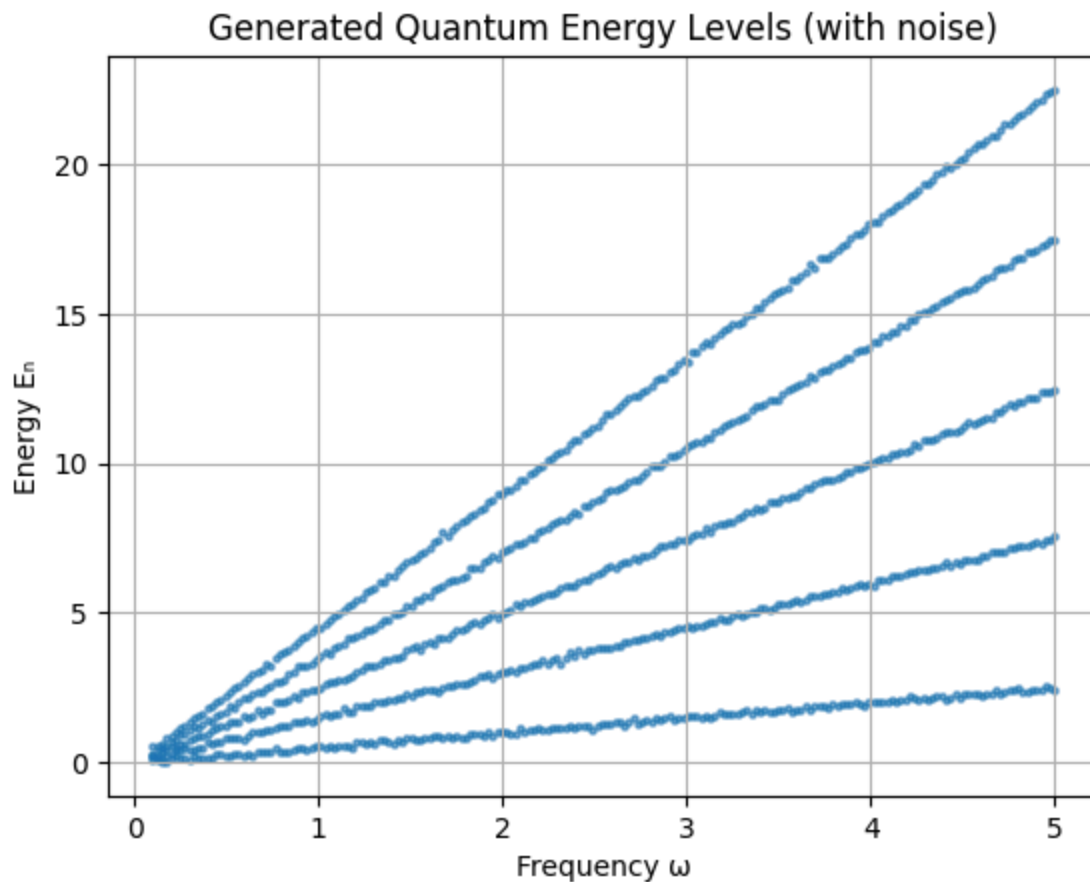
X = np.array(X)
y = np.array(y)

print(f"Input shape: {X.shape}, Output shape: {y.shape}")

# Optional: visualize
plt.scatter(X[:,0], y, s=5, alpha=0.6)
plt.xlabel("Frequency  $\omega$ ")
plt.ylabel("Energy  $E_n$ ")
plt.title("Generated Quantum Energy Levels (with noise)")
plt.grid(True)
plt.show()

```

⇒ Input shape: (1000, 2), Output shape: (1000,)



```
import tensorflow as tf
from tensorflow import keras
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
import matplotlib.pyplot as plt

# Scale inputs (omega and n)
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# Split data into training and test sets (80% train, 20% test)
X_train, X_test, y_train, y_test = train_test_split(
    X_scaled, y, test_size=0.2, random_state=42
)

# Build the neural network model
model = keras.Sequential([
    keras.layers.Dense(32, activation='relu', input_shape=(2,)),
    keras.layers.Dense(32, activation='relu'),
    keras.layers.Dense(1) # Output: predicted energy
])

# Compile the model with optimizer and loss function
model.compile(optimizer='adam', loss='mse', metrics=['mae'])
```

```
# Train the model for 100 epochs
history = model.fit(X_train, y_train, epochs=100, batch_size=16,
                    validation_data=(X_test, y_test), verbose=1)

# Plot training and validation loss curves
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss (MSE)')
plt.title('Training Performance')
plt.legend()
plt.grid(True)
plt.show()
```



```
/usr/local/lib/python3.11/dist-packages/keras/src/layers/core/dense.py:87: UserW  
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```
Epoch 1/100  
50/50 ————— 2s 7ms/step - loss: 56.5495 - mae: 5.6821 - val_lo  
Epoch 2/100  
50/50 ————— 0s 4ms/step - loss: 42.8536 - mae: 4.5953 - val_lo  
Epoch 3/100  
50/50 ————— 0s 4ms/step - loss: 19.3913 - mae: 3.3286 - val_lo  
Epoch 4/100  
50/50 ————— 0s 3ms/step - loss: 4.9550 - mae: 1.7785 - val_lo  
Epoch 5/100  
50/50 ————— 0s 4ms/step - loss: 2.5157 - mae: 1.2466 - val_lo  
Epoch 6/100  
50/50 ————— 0s 4ms/step - loss: 1.5066 - mae: 0.9306 - val_lo  
Epoch 7/100  
50/50 ————— 0s 3ms/step - loss: 1.2531 - mae: 0.8507 - val_lo  
Epoch 8/100  
50/50 ————— 0s 3ms/step - loss: 0.8853 - mae: 0.7041 - val_lo  
Epoch 9/100  
50/50 ————— 0s 4ms/step - loss: 0.7380 - mae: 0.6573 - val_lo  
Epoch 10/100  
50/50 ————— 0s 3ms/step - loss: 0.4777 - mae: 0.5091 - val_lo  
Epoch 11/100  
50/50 ————— 0s 4ms/step - loss: 0.3530 - mae: 0.4357 - val_lo  
Epoch 12/100  
50/50 ————— 0s 3ms/step - loss: 0.2570 - mae: 0.3694 - val_lo  
Epoch 13/100  
50/50 ————— 0s 3ms/step - loss: 0.2035 - mae: 0.3385 - val_lo  
Epoch 14/100  
50/50 ————— 0s 3ms/step - loss: 0.1866 - mae: 0.3167 - val_lo  
Epoch 15/100  
50/50 ————— 0s 3ms/step - loss: 0.1399 - mae: 0.2784 - val_lo  
Epoch 16/100  
50/50 ————— 0s 4ms/step - loss: 0.1395 - mae: 0.2733 - val_lo  
Epoch 17/100  
50/50 ————— 0s 3ms/step - loss: 0.1128 - mae: 0.2483 - val_lo  
Epoch 18/100  
50/50 ————— 0s 3ms/step - loss: 0.1109 - mae: 0.2378 - val_lo  
Epoch 19/100  
50/50 ————— 0s 3ms/step - loss: 0.1174 - mae: 0.2476 - val_lo  
Epoch 20/100  
50/50 ————— 0s 3ms/step - loss: 0.1000 - mae: 0.2224 - val_lo  
Epoch 21/100  
50/50 ————— 0s 3ms/step - loss: 0.0930 - mae: 0.2115 - val_lo  
Epoch 22/100  
50/50 ————— 0s 3ms/step - loss: 0.0860 - mae: 0.2061 - val_lo  
Epoch 23/100  
50/50 ————— 0s 4ms/step - loss: 0.0681 - mae: 0.1839 - val_lo  
Epoch 24/100  
50/50 ————— 0s 3ms/step - loss: 0.0694 - mae: 0.1827 - val_lo  
Epoch 25/100  
50/50 ————— 0s 3ms/step - loss: 0.0653 - mae: 0.1766 - val_lo  
Epoch 26/100  
50/50 ————— 0s 3ms/step - loss: 0.0547 - mae: 0.1633 - val_lo  
Epoch 27/100  
50/50 ————— 0s 4ms/step - loss: 0.0612 - mae: 0.1715 - val_lo
```

[illegible]

50/50	0s 4ms/step	loss: 0.0085	mae: 0.0685	val_loss: 0.0085
Epoch 57/100				
50/50	0s 4ms/step	loss: 0.0083	mae: 0.0654	val_loss: 0.0083
Epoch 58/100				
50/50	0s 3ms/step	loss: 0.0088	mae: 0.0679	val_loss: 0.0088
Epoch 59/100				
50/50	0s 3ms/step	loss: 0.0068	mae: 0.0610	val_loss: 0.0068
Epoch 60/100				
50/50	0s 4ms/step	loss: 0.0061	mae: 0.0591	val_loss: 0.0061
Epoch 61/100				
50/50	0s 3ms/step	loss: 0.0075	mae: 0.0666	val_loss: 0.0075
Epoch 62/100				
50/50	0s 3ms/step	loss: 0.0068	mae: 0.0617	val_loss: 0.0068
Epoch 63/100				
50/50	0s 4ms/step	loss: 0.0060	mae: 0.0616	val_loss: 0.0060
Epoch 64/100				
50/50	0s 4ms/step	loss: 0.0055	mae: 0.0566	val_loss: 0.0055
Epoch 65/100				
50/50	0s 3ms/step	loss: 0.0059	mae: 0.0581	val_loss: 0.0059
Epoch 66/100				
50/50	0s 4ms/step	loss: 0.0054	mae: 0.0552	val_loss: 0.0054
Epoch 67/100				
50/50	0s 3ms/step	loss: 0.0047	mae: 0.0519	val_loss: 0.0047
Epoch 68/100				
50/50	0s 4ms/step	loss: 0.0047	mae: 0.0532	val_loss: 0.0047
Epoch 69/100				
50/50	0s 4ms/step	loss: 0.0052	mae: 0.0555	val_loss: 0.0052
Epoch 70/100				
50/50	0s 4ms/step	loss: 0.0056	mae: 0.0564	val_loss: 0.0056
Epoch 71/100				
50/50	0s 3ms/step	loss: 0.0050	mae: 0.0555	val_loss: 0.0050
Epoch 72/100				
50/50	0s 3ms/step	loss: 0.0044	mae: 0.0522	val_loss: 0.0044
Epoch 73/100				
50/50	0s 3ms/step	loss: 0.0041	mae: 0.0499	val_loss: 0.0041
Epoch 74/100				
50/50	0s 3ms/step	loss: 0.0044	mae: 0.0529	val_loss: 0.0044
Epoch 75/100				
50/50	0s 4ms/step	loss: 0.0041	mae: 0.0496	val_loss: 0.0041
Epoch 76/100				
50/50	0s 3ms/step	loss: 0.0043	mae: 0.0521	val_loss: 0.0043
Epoch 77/100				
50/50	0s 4ms/step	loss: 0.0038	mae: 0.0486	val_loss: 0.0038
Epoch 78/100				
50/50	0s 3ms/step	loss: 0.0036	mae: 0.0467	val_loss: 0.0036
Epoch 79/100				
50/50	0s 5ms/step	loss: 0.0036	mae: 0.0471	val_loss: 0.0036
Epoch 80/100				
50/50	0s 5ms/step	loss: 0.0033	mae: 0.0449	val_loss: 0.0033
Epoch 81/100				
50/50	0s 5ms/step	loss: 0.0045	mae: 0.0521	val_loss: 0.0045
Epoch 82/100				
50/50	0s 5ms/step	loss: 0.0041	mae: 0.0506	val_loss: 0.0041
Epoch 83/100				
50/50	0s 5ms/step	loss: 0.0036	mae: 0.0473	val_loss: 0.0036
Epoch 84/100				

50/50 ————— 0s 7ms/step - loss: 0.0039 - mae: 0.0494 - val\_loss: 0.0494  
Epoch 85/100

50/50 ————— 0s 5ms/step - loss: 0.0037 - mae: 0.0479 - val\_loss: 0.0479  
Epoch 86/100

50/50 ————— 0s 5ms/step - loss: 0.0038 - mae: 0.0482 - val\_loss: 0.0482  
Epoch 87/100

50/50 ————— 0s 4ms/step - loss: 0.0035 - mae: 0.0478 - val\_loss: 0.0478  
Epoch 88/100

50/50 ————— 0s 4ms/step - loss: 0.0035 - mae: 0.0467 - val\_loss: 0.0467  
Epoch 89/100

50/50 ————— 0s 4ms/step - loss: 0.0035 - mae: 0.0461 - val\_loss: 0.0461  
Epoch 90/100

50/50 ————— 0s 3ms/step - loss: 0.0033 - mae: 0.0468 - val\_loss: 0.0468  
Epoch 91/100

50/50 ————— 0s 3ms/step - loss: 0.0034 - mae: 0.0457 - val\_loss: 0.0457  
Epoch 92/100

50/50 ————— 0s 3ms/step - loss: 0.0032 - mae: 0.0442 - val\_loss: 0.0442  
Epoch 93/100

50/50 ————— 0s 3ms/step - loss: 0.0039 - mae: 0.0508 - val\_loss: 0.0508  
Epoch 94/100

50/50 ————— 0s 3ms/step - loss: 0.0035 - mae: 0.0476 - val\_loss: 0.0476  
Epoch 95/100

50/50 ————— 0s 4ms/step - loss: 0.0033 - mae: 0.0458 - val\_loss: 0.0458  
Epoch 96/100

50/50 ————— 0s 3ms/step - loss: 0.0031 - mae: 0.0442 - val\_loss: 0.0442  
Epoch 97/100

50/50 ————— 0s 3ms/step - loss: 0.0033 - mae: 0.0469 - val\_loss: 0.0469  
Epoch 98/100

50/50 ————— 0s 3ms/step - loss: 0.0033 - mae: 0.0462 - val\_loss: 0.0462  
Epoch 99/100

50/50 ————— 0s 4ms/step - loss: 0.0033 - mae: 0.0464 - val\_loss: 0.0464  
Epoch 100/100

50/50 ————— 0s 3ms/step - loss: 0.0034 - mae: 0.0472 - val\_loss: 0.0472

