

In [2]: *#LBL1*

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense,
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.callbacks import EarlyStopping, ReduceLRonPlateau
from tensorflow.keras.utils import to_categorical
import cv2
```

In [3]: *#LBL2*

```
def explore_dataset(file_path):
    # Загрузка данных
    data = np.load(file_path)
    images = data['data'] # Извлечение изображений
    labels = data['labels'] # Извлечение меток

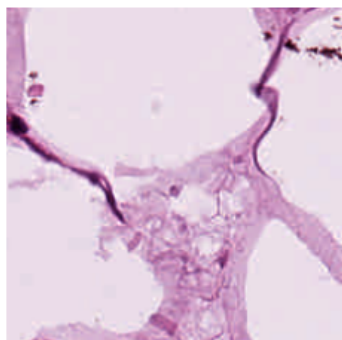
    print(f"Файл: {file_path}")
    print(f"Количество изображений: {images.shape[0]}")
    print(f"Размер каждого изображения: {images[0].shape}")
    print(f"Количество меток: {len(labels)}")
    print(f"Пример меток: {np.unique(labels)}")

    # Визуализация нескольких изображений
    plt.figure(figsize=(10, 10))
    for i in range(9): # Покажем первые 9 изображений
        plt.subplot(3, 3, i + 1)
        plt.imshow(images[i])
        plt.title(f"Label: {labels[i]}")
        plt.axis('off')
    plt.show()
```

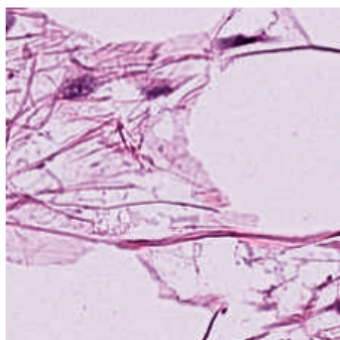
In [4]: *# Просмотр train и test данных*  
explore\_dataset('train.npz') *#LBL2*  
explore\_dataset('test.npz') *#LBL2*

Файл: train.npz  
Количество изображений: 18000  
Размер каждого изображения: (224, 224, 3)  
Количество меток: 18000  
Пример меток: [0 1 2 3 4 5 6 7 8]

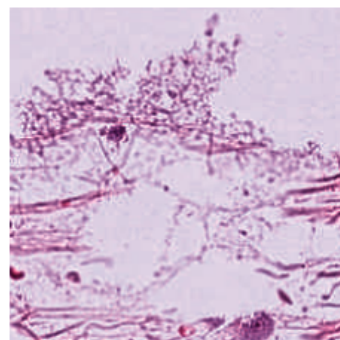
Label: 0



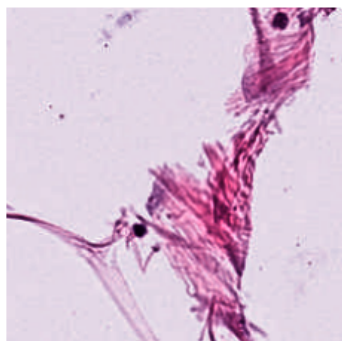
Label: 0



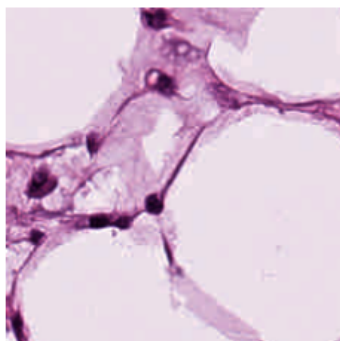
Label: 0



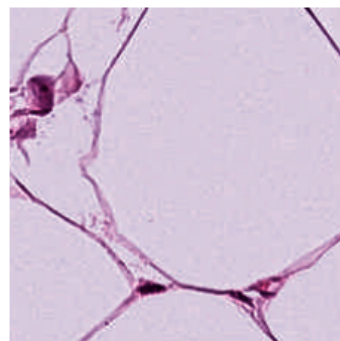
Label: 0



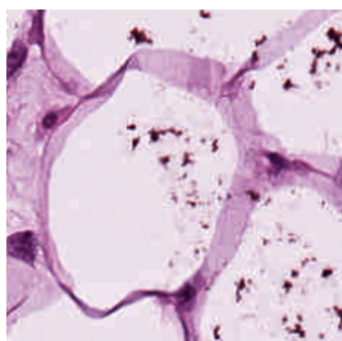
Label: 0



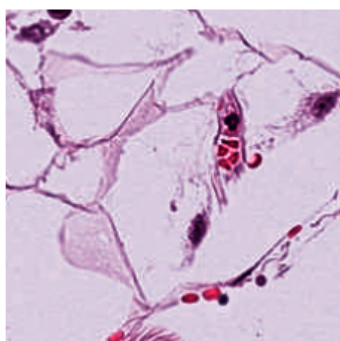
Label: 0



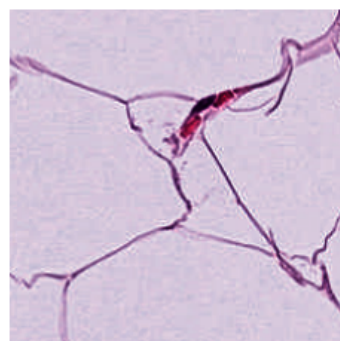
Label: 0



Label: 0



Label: 0



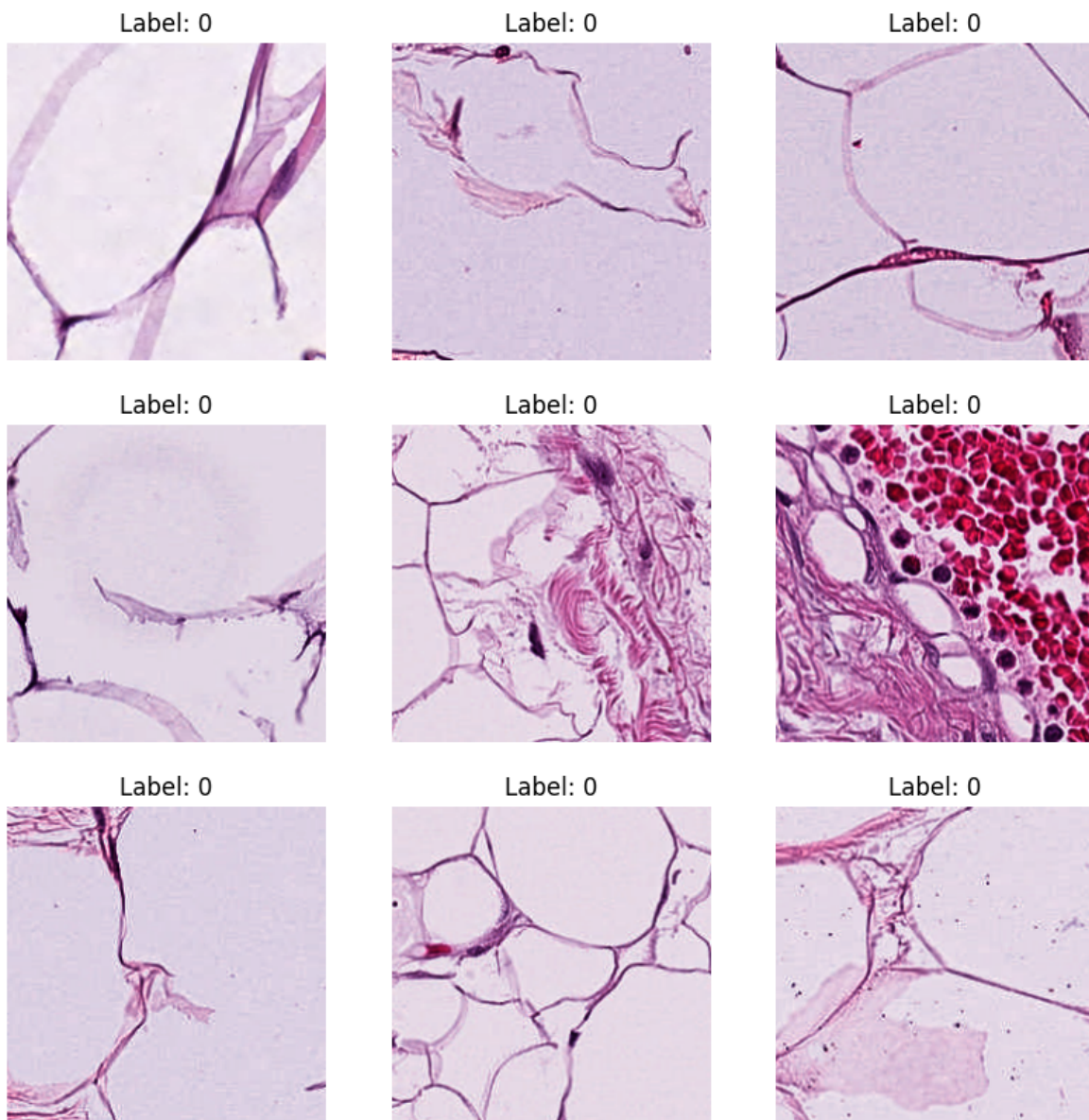
Файл: test.npz

Количество изображений: 4500

Размер каждого изображения: (224, 224, 3)

Количество меток: 4500

Пример меток: [0 1 2 3 4 5 6 7 8]



In [5]: #LBL3

```
class Dataset:
    def __init__(self, name, resize_to=(64, 64)): # Размер 64x64
        self.name = name
        self.is_loaded = False
        output = f'{name}.npz'
        print(f'Loading dataset {self.name} from npz.')
        np_obj = np.load(output)
        resized_images = [cv2.resize(img, resize_to) for img in np_obj['d']]
        self.images = np.array(resized_images, dtype=np.float32) / 255.0
        self.labels = np_obj['labels']
        self.n_files = self.images.shape[0]
        self.is_loaded = True
        print(f'Done. Dataset {self.name} consists of {self.n_files} images')

    def get_data_and_labels(self):
        return self.images, to_categorical(self.labels, num_classes=9)
```

In [6]: #LBL4

```
class CNNModel:
    def __init__(self):
```

```

self.model = Sequential([
    Conv2D(32, (3, 3), activation='relu', input_shape=(64, 64, 3)),
    BatchNormalization(),
    MaxPooling2D((2, 2)),
    Conv2D(64, (3, 3), activation='relu'),
    BatchNormalization(),
    MaxPooling2D((2, 2)),
    Conv2D(128, (3, 3), activation='relu'),
    BatchNormalization(),
    MaxPooling2D((2, 2)),
    Flatten(),
    Dense(256, activation='relu'),
    Dropout(0.5),
    Dense(9, activation='softmax')
])
self.model.compile(optimizer=Adam(learning_rate=0.0001),
                    loss='categorical_crossentropy',
                    metrics=['accuracy'])

def train(self, X_train, y_train, X_val, y_val, epochs=20, batch_size
early_stopping = EarlyStopping(monitor='val_accuracy', patience=5
lr_reduction = ReduceLROnPlateau(monitor='val_loss', patience=3,

print("Начало обучения модели CNN...")
history = self.model.fit(X_train, y_train,
                        validation_data=(X_val, y_val),
                        epochs=epochs,
                        batch_size=batch_size,
                        callbacks=[early_stopping, lr_reduction])
print("Обучение завершено.")
return history

def save(self, name: str):
    self.model.save(f'{name}.h5')
    print(f'Model saved at: {name}.h5')

def load(self, name: str):
    self.model = tf.keras.models.load_model(f'{name}.h5')
    print(f'Model loaded from: {name}.h5')

def evaluate(self, X_test, y_test):
    test_loss, test_accuracy = self.model.evaluate(X_test, y_test)
    print(f'Test Accuracy: {test_accuracy:.4f}')
    return test_accuracy

```

```

In [7]: #LBL5
d_train = Dataset('train')
d_test = Dataset('test')

X_train, y_train = d_train.get_data_and_labels()
X_test, y_test = d_test.get_data_and_labels()

#LBL6
X_train_split, X_val, y_train_split, y_val = train_test_split(X_train, y

#LBL7
model = CNNModel()
history = model.train(X_train_split, y_train_split, X_val, y_val, epochs=

#LBL8

```

```
model.save('cnn_model')

#LBL9
test_accuracy = model.evaluate(X_test, y_test)
```

Loading dataset train from npz.

Done. Dataset train consists of 18000 images with size (64, 64).

Loading dataset test from npz.

Done. Dataset test consists of 4500 images with size (64, 64).

/home/fantom/jupyter\_env/lib/python3.12/site-packages/keras/src/layers/convolutional/base\_conv.py:107: UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When using Sequential models, prefer using a n `Input(shape)` object as the first layer in the model instead.

super().\_\_init\_\_(activity\_regularizer=activity\_regularizer, \*\*kwargs)

2024-12-02 23:51:39.196954: E external/local\_xla/xla/stream\_executor/cuda/cuda\_driver.cc:152] failed call to cuInit: INTERNAL: CUDA error: Failed call to cuInit: UNKNOWN ERROR (303)

Начало обучения модели CNN...

Epoch 1/20

2024-12-02 23:51:39.478198: W external/local\_xla/xla/tsl/framework/cpu\_allocator\_impl.cc:83] Allocation of 707788800 exceeds 10% of free system memory.



225/225 ————— 68s 291ms/step - accuracy: 0.4280 - loss: 1.9  
289 - val\_accuracy: 0.1147 - val\_loss: 8.3743 - learning\_rate: 1.0000e-04  
Epoch 2/20

225/225 ————— 63s 278ms/step - accuracy: 0.6386 - loss: 0.9  
795 - val\_accuracy: 0.2528 - val\_loss: 3.7920 - learning\_rate: 1.0000e-04  
Epoch 3/20

225/225 ————— 67s 298ms/step - accuracy: 0.7138 - loss: 0.7  
905 - val\_accuracy: 0.6969 - val\_loss: 0.8527 - learning\_rate: 1.0000e-04  
Epoch 4/20

225/225 ————— 71s 314ms/step - accuracy: 0.7615 - loss: 0.6  
607 - val\_accuracy: 0.7397 - val\_loss: 0.7220 - learning\_rate: 1.0000e-04  
Epoch 5/20

225/225 ————— 63s 280ms/step - accuracy: 0.7969 - loss: 0.5  
547 - val\_accuracy: 0.6944 - val\_loss: 0.9458 - learning\_rate: 1.0000e-04  
Epoch 6/20

225/225 ————— 62s 275ms/step - accuracy: 0.8368 - loss: 0.4  
757 - val\_accuracy: 0.7608 - val\_loss: 0.7455 - learning\_rate: 1.0000e-04  
Epoch 7/20

225/225 ————— 66s 294ms/step - accuracy: 0.8571 - loss: 0.4  
091 - val\_accuracy: 0.7128 - val\_loss: 0.8511 - learning\_rate: 1.0000e-04  
Epoch 8/20

225/225 ————— 70s 310ms/step - accuracy: 0.8855 - loss: 0.3  
307 - val\_accuracy: 0.8317 - val\_loss: 0.4972 - learning\_rate: 5.0000e-05  
Epoch 9/20

225/225 ————— 63s 280ms/step - accuracy: 0.9127 - loss: 0.2  
670 - val\_accuracy: 0.8164 - val\_loss: 0.5499 - learning\_rate: 5.0000e-05  
Epoch 10/20

225/225 ————— 63s 280ms/step - accuracy: 0.9152 - loss: 0.2  
540 - val\_accuracy: 0.8356 - val\_loss: 0.4921 - learning\_rate: 5.0000e-05  
Epoch 11/20

225/225 ————— 63s 278ms/step - accuracy: 0.9222 - loss: 0.2  
337 - val\_accuracy: 0.8319 - val\_loss: 0.4977 - learning\_rate: 5.0000e-05  
Epoch 12/20

225/225 ————— 63s 280ms/step - accuracy: 0.9317 - loss: 0.2  
061 - val\_accuracy: 0.8461 - val\_loss: 0.4650 - learning\_rate: 5.0000e-05  
Epoch 13/20

225/225 ————— 63s 280ms/step - accuracy: 0.9412 - loss: 0.1  
841 - val\_accuracy: 0.8483 - val\_loss: 0.4714 - learning\_rate: 5.0000e-05  
Epoch 14/20

225/225 ————— 63s 278ms/step - accuracy: 0.9470 - loss: 0.1  
652 - val\_accuracy: 0.8267 - val\_loss: 0.5644 - learning\_rate: 5.0000e-05  
Epoch 15/20

225/225 ————— 62s 276ms/step - accuracy: 0.9471 - loss: 0.1  
575 - val\_accuracy: 0.8392 - val\_loss: 0.5014 - learning\_rate: 5.0000e-05  
Epoch 16/20

225/225 ————— 63s 280ms/step - accuracy: 0.9573 - loss: 0.1  
337 - val\_accuracy: 0.8492 - val\_loss: 0.4640 - learning\_rate: 2.5000e-05  
Epoch 17/20

225/225 ————— 63s 278ms/step - accuracy: 0.9665 - loss: 0.1  
166 - val\_accuracy: 0.8589 - val\_loss: 0.4521 - learning\_rate: 2.5000e-05  
Epoch 18/20

225/225 ————— 63s 281ms/step - accuracy: 0.9657 - loss: 0.1  
091 - val\_accuracy: 0.8522 - val\_loss: 0.4687 - learning\_rate: 2.5000e-05  
Epoch 19/20

225/225 ————— 63s 279ms/step - accuracy: 0.9689 - loss: 0.1  
044 - val\_accuracy: 0.8603 - val\_loss: 0.4417 - learning\_rate: 2.5000e-05  
Epoch 20/20

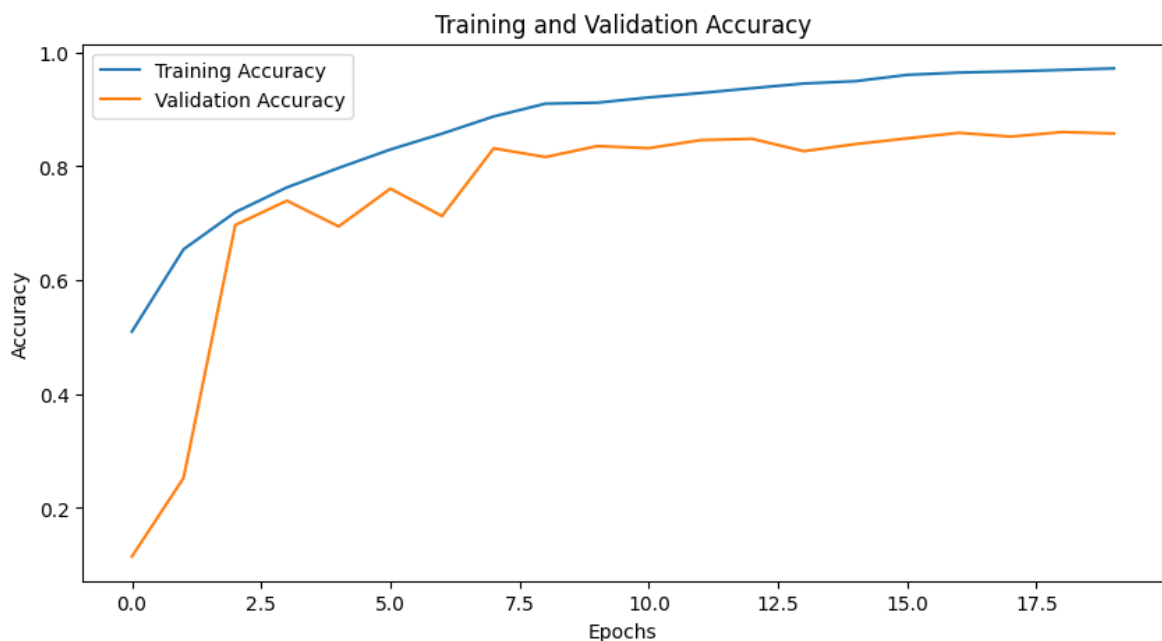
225/225 ————— 63s 280ms/step - accuracy: 0.9701 - loss: 0.1  
051 - val\_accuracy: 0.8578 - val\_loss: 0.4446 - learning\_rate: 2.5000e-05

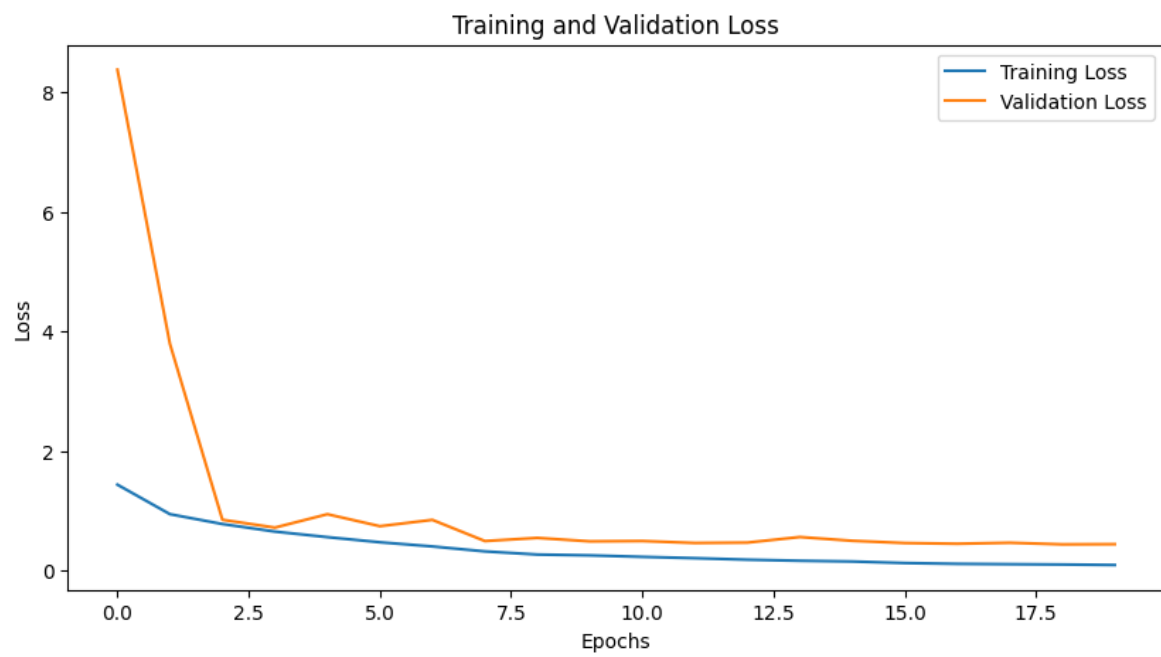
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save\_model(model)`. This file format is considered legacy . We recommend using instead the native Keras format, e.g. `model.save('my\_model.keras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

Обучение завершено.  
Model saved at: cnn\_model.h5  
**141/141** ————— **4s** 29ms/step - accuracy: 0.9101 - loss: 0.2743  
Test Accuracy: 0.8591

```
In [8]: #LBL10
plt.figure(figsize=(10, 5))
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Training and Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()

plt.figure(figsize=(10, 5))
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
```





In [ ]: