Чтобы изменить содержимое ячейки, дважды нажмите на нее (или выберите "Ввод")



Подключение google disk для использовани Подключение google disk для

Подключение google disk для использования dataset

from google.colab import drive
drive.mount('/content/drive')

→ Mounted at /content/drive

Установка библиотеки

Чтобы изменить содержимое ячейки, дважды нажмите на нее (или выберите "Ввод")

!pip install petroscope

Requirement already satisfied: petroscope in /usr/local/lib/python3.11/dist Requirement already satisfied: pyyaml in /usr/local/lib/python3.11/dist-pac Requirement already satisfied: numpy<2.0.0,>=1.16 in /usr/local/lib/python3 Requirement already satisfied: pillow in /usr/local/lib/python3.11/dist-pac Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packa Requirement already satisfied: scipy in /usr/local/lib/python3.11/dist-pack Requirement already satisfied: loguru in /usr/local/lib/python3.11/dist-pac Requirement already satisfied: prettytable in /usr/local/lib/python3.11/dis Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-p Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.1 Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/di Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3. Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3. Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11 Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.1 Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/pytho Requirement already satisfied: wcwidth in /usr/local/lib/python3.11/dist-pa Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/p Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/di Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3 Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3 Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-p

Импорт неободимых библиотек

```
from pathlib import Path
from petroscope.segmentation.classes import ClassSet, LumenStoneClasses
from petroscope.segmentation.utils import load_image, load_mask
from petroscope.segmentation import GeoSegmModel
import numpy as np
from tqdm import tqdm
import tensorflow as tf
from tensorflow.keras.models import Model, load_model
from tensorflow.keras.layers import Input, Conv2D, MaxPooling2D, UpSampling2D,
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.utils import to_categorical
from skimage.color import rgb2lab
import os
```

Путь к dataset

```
ds_path = Path('/content/drive/MyDrive/PhotoSets/')
```

Инциализация набора классов для сегментации

```
classset = LumenStoneClasses.S1v1()

for cl in classset.classes:
    print(cl)

       [0, bg (background), color: #000000]
       [1, ccp/kub (chalcopyrite/cubanite), color: #ffa500]
       [2, gl (galena), color: #9acd32]
```

[4, brt (bornite), color: #00bfff]
[6, py/mrc (pyrite/marcasite), color: #2f4f4f]

[8, sph (sphalerite), color: #ee82ee]

[11, tnt/ttr (tenantite/tetrahedrite), color: #483d8b]

Формирование путей к изображениям и маскам

```
train_img_mask_p = [
    (img_p, ds_path / "masks" / "train" / f"{img_p.stem}.png")
    for img_p in sorted((ds_path / "imgs" / "train").iterdir())
]

test_img_mask_p = [
    (img_p, ds_path / "masks" / "test" / f"{img_p.stem}.png")
    for img_p in sorted((ds_path / "imgs" / "test").iterdir())
]
```

Загрузка и конвертация обучающих изображений в пространство LAB

```
for img_p, _ in train_img_mask_p:
    img = load_image(img_p, normalize=True)
    img lab = rgb2lab(img)
   print(f"Image {img_p.name}: {img_lab.shape}, {img_lab.dtype}")
→ Image train_01.jpg: (2547, 3396, 3), float32
    Image train_02.jpg: (2547, 3396, 3), float32
    Image train_03.jpg: (2547, 3396, 3), float32
    Image train 04.jpg: (2547, 3396, 3), float32
    Image train_06.jpg: (2547, 3396, 3), float32
    Image train_07.jpg: (2547, 3396, 3), float32
    Image train_08.jpg: (2547, 3396, 3), float32
    Image train_09.jpg: (2547, 3396, 3), float32
    Image train_10.jpg: (2547, 3396, 3), float32
    Image train_11.jpg: (2547, 3396, 3), float32
    Image train_12.jpg: (2547, 3396, 3), float32
    Image train_13.jpg: (2547, 3396, 3), float32
    Image train_14.jpg: (2547, 3396, 3), float32
    Image train_15.jpg: (2547, 3396, 3), float32
    Image train_16.jpg: (2547, 3396, 3), float32
    Image train_17.jpg: (2547, 3396, 3), float32
    Image train_18.jpg: (2547, 3396, 3), float32
    Image train_19.jpg: (2547, 3396, 3), float32
    Image train_20.jpg: (2547, 3396, 3), float32
    Image train 21.jpg: (2547, 3396, 3), float32
    Image train_22.jpg: (2547, 3396, 3), float32
    Image train_23.jpg: (2547, 3396, 3), float32
    Image train_24.jpg: (2547, 3396, 3), float32
    Image train_25.jpg: (2547, 3396, 3), float32
    Image train_26.jpg: (2547, 3396, 3), float32
    Image train_27.jpg: (2547, 3396, 3), float32
    Image train_28.jpg: (2547, 3396, 3), float32
    Image train_29.jpg: (2547, 3396, 3), float32
    Image train_30.jpg: (2547, 3396, 3), float32
    Image train_31.jpg: (2547, 3396, 3), float32
    Image train_32.jpg: (2547, 3396, 3), float32
    Image train_33.jpg: (2547, 3396, 3), float32
```

```
Image train 34.jpg: (2547, 3396, 3), float32
Image train_35.jpg: (2547, 3396, 3), float32
Image train_36.jpg: (2547, 3396, 3), float32
Image train_37.jpg: (2547, 3396, 3), float32
Image train_38.jpg: (2547, 3396, 3), float32
Image train_39.jpg: (2547, 3396, 3), float32
Image train 40.jpg: (2547, 3396, 3), float32
Image train_41.jpg: (2547, 3396, 3), float32
Image train_42.jpg: (2547, 3396, 3), float32
Image train_43.jpg: (2547, 3396, 3), float32
Image train_44.jpg: (2547, 3396, 3), float32
Image train_45.jpg: (2547, 3396, 3), float32
Image train_46.jpg: (2547, 3396, 3), float32
Image train 47.jpg: (2547, 3396, 3), float32
Image train_48.jpg: (2547, 3396, 3), float32
Image train_49.jpg: (2547, 3396, 3), float32
Image train_50.jpg: (2547, 3396, 3), float32
Image train_51.jpg: (2547, 3396, 3), float32
Image train_52.jpg: (2547, 3396, 3), float32
Image train_53.jpg: (2547, 3396, 3), float32
Image train_54.jpg: (2547, 3396, 3), float32
Image train_55.jpg: (2547, 3396, 3), float32
Image train_56.jpg: (2547, 3396, 3), float32
Image train_57.jpg: (2547, 3396, 3), float32
Image train_58.jpg: (2547, 3396, 3), float32
Image train 59.jpg: (2547, 3396, 3), float32
```

Загрузка масок без one-hot кодирования

```
for _, mask_p in train_img_mask_p:
   mask = load_mask(mask_p, classes=classset, one_hot=False)
   print(f"Mask {mask_p.name}: {mask.shape}, {mask.dtype}")
   Mask train 01.png: (2547, 3396), uint8
    Mask train_02.png: (2547, 3396), uint8
    Mask train_03.png: (2547, 3396), uint8
    Mask train_04.png: (2547, 3396), uint8
    Mask train 06.png: (2547, 3396), uint8
    Mask train_07.png: (2547, 3396), uint8
    Mask train_08.png: (2547, 3396), uint8
    Mask train_09.png: (2547, 3396), uint8
    Mask train_10.png: (2547, 3396), uint8
    Mask train_11.png: (2547, 3396), uint8
    Mask train_12.png: (2547, 3396), uint8
    Mask train_13.png: (2547, 3396), uint8
    Mask train_14.png: (2547, 3396), uint8
    Mask train_15.png: (2547, 3396), uint8
    Mask train_16.png: (2547, 3396), uint8
    Mask train_17.png: (2547, 3396), uint8
    Mask train 18.png: (2547, 3396), uint8
    Mask train_19.png: (2547, 3396), uint8
```

```
Mask train 20.png: (2547, 3396), uint8
Mask train_21.png: (2547, 3396), uint8
Mask train_22.png: (2547, 3396), uint8
Mask train_23.png: (2547, 3396), uint8
Mask train_24.png: (2547, 3396), uint8
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Mask train_33.png: (2547, 3396), uint8
Mask train_34.png: (2547, 3396), uint8
Mask train_35.png: (2547, 3396), uint8
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Mask train_46.png: (2547, 3396), uint8
Mask train_47.png: (2547, 3396), uint8
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Mask train_49.png: (2547, 3396), uint8
Mask train_50.png: (2547, 3396), uint8
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Mask train_55.png: (2547, 3396), uint8
Mask train_56.png: (2547, 3396), uint8
Mask train_57.png: (2547, 3396), uint8
Mask train_58.png: (2547, 3396), uint8
Mask train_59.png: (2547, 3396), uint8
```

Загрузка масок с one-hot кодированием

```
for _, mask_p in train_img_mask_p:
    mask_one_hot = load_mask(mask_p, classes=classset, one_hot=True)
    print(f"Mask one-hot {mask_p.name}: {mask_one_hot.shape}, {mask_one_hot.dt})

→ Mask one-hot train_01.png: (2547, 3396, 7), float32
    Mask one-hot train_02.png: (2547, 3396, 7), float32
    Mask one-hot train_03.png: (2547, 3396, 7), float32
    Mask one-hot train_04.png: (2547, 3396, 7), float32
```

3306

Mack and-hot train 06 nng: 12517

```
riask ulic-liut trail_wu.plig. (2547, 550, 77, 110at52
Mask one-hot train_07.png: (2547, 3396, 7), float32
Mask one-hot train_08.png: (2547, 3396, 7), float32
Mask one-hot train 09.png: (2547, 3396, 7), float32
Mask one-hot train_10.png: (2547, 3396, 7), float32
Mask one-hot train_11.png: (2547, 3396, 7), float32
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Mask one-hot train_13.png: (2547, 3396, 7), float32
Mask one-hot train_14.png: (2547, 3396, 7), float32
Mask one-hot train 15.png: (2547, 3396, 7), float32
Mask one-hot train_16.png: (2547, 3396, 7), float32
Mask one-hot train_17.png: (2547, 3396, 7), float32
Mask one-hot train_18.png: (2547, 3396, 7), float32
Mask one-hot train_19.png: (2547, 3396, 7), float32
Mask one-hot train 20.png: (2547, 3396, 7), float32
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Mask one-hot train_31.png: (2547, 3396, 7), float32
Mask one-hot train_32.png: (2547, 3396, 7), float32
Mask one-hot train_33.png: (2547, 3396, 7), float32
Mask one-hot train 34.png: (2547, 3396, 7), float32
Mask one-hot train_35.png: (2547, 3396, 7), float32
Mask one-hot train_36.png: (2547, 3396, 7), float32
Mask one-hot train_37.png: (2547, 3396, 7), float32
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Mask one-hot train_39.png: (2547, 3396, 7), float32
Mask one-hot train_40.png: (2547, 3396, 7), float32
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Mask one-hot train_42.png: (2547, 3396, 7), float32
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Mask one-hot train_46.png: (2547, 3396, 7), float32
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Mask one-hot train_53.png: (2547, 3396, 7), float32
Mask one-hot train_54.png: (2547, 3396, 7), float32
Mask one-hot train_55.png: (2547, 3396, 7), float32
Mask one-hot train_56.png: (2547, 3396, 7), float32
Mask one-hot train_57.png: (2547, 3396, 7), float32
Mask one-hot train_58.png: (2547, 3396, 7), float32
Mask one-hot train_59.png: (2547, 3396, 7), float32
```

Загрузка цветных масок

```
for img_p, _ in train_img_mask_p:
   mask_colored_path = ds_path / "masks_colored_png" / "train" / f"{img_p.sten
   mask colored = load image(mask colored path, normalize=False)
   print(f"Colored mask {mask_colored_path.name}: {mask_colored.shape}, {mask_
   Colored mask train_01.png: (2547, 3396, 3), uint8
    Colored mask train_02.png: (2547, 3396, 3), uint8
    Colored mask train_03.png: (2547, 3396, 3), uint8
    Colored mask train_04.png: (2547, 3396, 3), uint8
    Colored mask train_06.png: (2547, 3396, 3), uint8
    Colored mask train_07.png: (2547, 3396, 3), uint8
    Colored mask train 08.png: (2547, 3396, 3), uint8
    Colored mask train_09.png: (2547, 3396, 3), uint8
    Colored mask train_10.png: (2547, 3396, 3), uint8
    Colored mask train_11.png: (2547, 3396, 3), uint8
    Colored mask train_12.png: (2547, 3396, 3), uint8
    Colored mask train_13.png: (2547, 3396, 3), uint8
    Colored mask train_14.png: (2547, 3396, 3), uint8
    Colored mask train_15.png: (2547, 3396, 3), uint8
    Colored mask train_16.png: (2547, 3396, 3), uint8
    Colored mask train_17.png: (2547, 3396, 3), uint8
    Colored mask train_18.png: (2547, 3396, 3), uint8
    Colored mask train_19.png: (2547, 3396, 3), uint8
    Colored mask train_20.png: (2547, 3396, 3), uint8
    Colored mask train 21.png: (2547, 3396, 3), uint8
    Colored mask train_22.png: (2547, 3396, 3), uint8
    Colored mask train_23.png: (2547, 3396, 3), uint8
    Colored mask train_24.png: (2547, 3396, 3), uint8
    Colored mask train_25.png: (2547, 3396, 3), uint8
    Colored mask train_26.png: (2547, 3396, 3), uint8
    Colored mask train_27.png: (2547, 3396, 3), uint8
    Colored mask train_28.png: (2547, 3396, 3), uint8
    Colored mask train_29.png: (2547, 3396, 3), uint8
    Colored mask train_30.png: (2547, 3396, 3), uint8
    Colored mask train_31.png: (2547, 3396, 3), uint8
    Colored mask train_32.png: (2547, 3396, 3), uint8
```

Colored mask train_33.png: (2547, 3396, 3), uint8 Colored mask train_34.png: (2547, 3396, 3), uint8 Colored mask train_35.png: (2547, 3396, 3), uint8 Colored mask train_36.png: (2547, 3396, 3), uint8 Colored mask train_37.png: (2547, 3396, 3), uint8 Colored mask train_38.png: (2547, 3396, 3), uint8 Colored mask train_39.png: (2547, 3396, 3), uint8 Colored mask train 40.png: (2547, 3396, 3), uint8 Colored mask train_41.png: (2547, 3396, 3), uint8 Colored mask train_42.png: (2547, 3396, 3), uint8 Colored mask train_43.png: (2547, 3396, 3), uint8 Colored mask train_44.png: (2547, 3396, 3), uint8

```
Colored mask train_45.png: (2547, 3396, 3), uint8
Colored mask train_46.png: (2547, 3396, 3), uint8
Colored mask train_47.png: (2547, 3396, 3), uint8
Colored mask train_48.png: (2547, 3396, 3), uint8
Colored mask train_49.png: (2547, 3396, 3), uint8
Colored mask train_50.png: (2547, 3396, 3), uint8
Colored mask train_51.png: (2547, 3396, 3), uint8
Colored mask train_52.png: (2547, 3396, 3), uint8
Colored mask train_53.png: (2547, 3396, 3), uint8
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Colored mask train_55.png: (2547, 3396, 3), uint8
Colored mask train_56.png: (2547, 3396, 3), uint8
Colored mask train_57.png: (2547, 3396, 3), uint8
Colored mask train_58.png: (2547, 3396, 3), uint8
Colored mask train_59.png: (2547, 3396, 3), uint8
Colored mask train_59.png: (2547, 3396, 3), uint8
```

Функция для загрузки и предобработки изображений и масок

```
def load_and_preprocess(img_path, mask_path, classes, img_size=(256, 256)):
    img = load_image(img_path, normalize=True)
    img = tf.image.resize(img, img_size)

mask = load_mask(mask_path, classes=classes, one_hot=False)
    mask = tf.image.resize(mask[..., np.newaxis], img_size, method='nearest')
    mask = to_categorical(mask, num_classes=len(classes))

return img, mask
```

Унаследованный класс от GeoSegmModel, метод сегментации - нейронная сеть U-Net

```
class UNetSegmModel(GeoSegmModel):
    def __init__(self, classes: ClassSet, input_size=(256, 256, 3), save_path='
        super().__init__()
        self.classes = classes
        self.input_size = input_size
        self.save_path = Path(save_path)
        self.model = self._build_model()

def _build_model(self):
    inputs = Input(self.input_size)

    conv1 = Conv2D(64, 3, activation='relu', padding='same')(inputs)
    conv1 = Conv2D(64, 3, activation='relu', padding='same')(conv1)
    pool1 = MaxPooling2D(pool_size=(2, 2))(conv1)
    conv2 = Conv2D(128, 3, activation='relu', padding='same')(pool1)
    conv2 = Conv2D(128, 3, activation='relu', padding='same')(conv2)
```

```
pool2 = MaxPooling2D(pool size=(2, 2))(conv2)
    conv3 = Conv2D(256, 3, activation='relu', padding='same')(pool2)
    conv3 = Conv2D(256, 3, activation='relu', padding='same')(conv3)
    pool3 = MaxPooling2D(pool_size=(2, 2))(conv3)
    conv4 = Conv2D(512, 3, activation='relu', padding='same')(pool3)
    conv4 = Conv2D(512, 3, activation='relu', padding='same')(conv4)
    up5 = concatenate([UpSampling2D(size=(2, 2))(conv4), conv3], axis=-1)
    conv5 = Conv2D(256, 3, activation='relu', padding='same')(up5)
    conv5 = Conv2D(256, 3, activation='relu', padding='same')(conv5)
    up6 = concatenate([UpSampling2D(size=(2, 2))(conv5), conv2], axis=-1)
    conv6 = Conv2D(128, 3, activation='relu', padding='same')(up6)
    conv6 = Conv2D(128, 3, activation='relu', padding='same')(conv6)
    up7 = concatenate([UpSampling2D(size=(2, 2))(conv6), conv1], axis=-1)
    conv7 = Conv2D(64, 3, activation='relu', padding='same')(up7)
    conv7 = Conv2D(64, 3, activation='relu', padding='same')(conv7)
    outputs = Conv2D(len(self.classes), 1, activation='softmax')(conv7)
    model = Model(inputs=[inputs], outputs=[outputs])
    model.compile(optimizer=Adam(), loss='categorical_crossentropy', metric
    return model
def train(self, img_mask_paths, epochs=20, batch_size=4, validation_split=0
    train_data = [load_and_preprocess(img_p, mask_p, self.classes) for img_
    train_images, train_masks = zip(*train_data)
    train_images = np.array(train_images)
    train_masks = np.array(train_masks)
    self.model.fit([train_images], [train_masks], batch_size=batch_size, er
    self.save()
def predict image(self, image: np.ndarray) -> np.ndarray:
    img_resized = tf.image.resize(image, self.input_size[:2])
    pred = self.model.predict(np.expand_dims(img_resized, axis=0))
    pred = np.argmax(pred[0], axis=-1)
    return pred.astype(np.uint8)
def save(self):
    self.save_path.mkdir(parents=True, exist_ok=True)
    self.model.save(self.save_path / "unet_model.keras")
def load(self, saved_path: Path, **kwargs):
    if not saved_path.exists():
        raise FileNotFoundError(f"Model file not found at {saved_path}")
    self.model = load_model(saved_path / "unet_model.keras")
```

Обучение модели

model = UNetSegmModel(classes=classset, input_size=(256, 256, 3), save_path="une
model.train(train_img_mask_p, epochs=200, batch_size=8, validation_split=0.1)

```
Loading data: 100%| 58/58 [00:32<00:00,
                                                     1.77it/sl
Epoch 1/200
7/7 ----
                      — 81s 6s/step - accuracy: 0.2878 - loss: 3.1463 - va
Epoch 2/200
                       - 9s 335ms/step - accuracy: 0.3057 - loss: 1.7832 -
7/7 -
Epoch 3/200
7/7 -
                       - 2s 345ms/step - accuracy: 0.4016 - loss: 1.4435 -
Epoch 4/200
7/7 -
                       - 2s 337ms/step - accuracy: 0.5407 - loss: 1.2612 -
Epoch 5/200
7/7 -
                       — 3s 336ms/step — accuracy: 0.6613 — loss: 1.1003 —
Epoch 6/200
                       - 2s 336ms/step - accuracy: 0.6560 - loss: 1.0465 -
7/7 -
Epoch 7/200
                       — 3s 335ms/step — accuracy: 0.7034 — loss: 0.9920 —
7/7 -
Epoch 8/200
7/7 -
                       - 3s 350ms/step - accuracy: 0.7238 - loss: 0.9609 -
Epoch 9/200
7/7 -
                        - 2s 339ms/step - accuracy: 0.6060 - loss: 1.1678 -
Epoch 10/200
7/7 -
                       — 2s 337ms/step - accuracy: 0.7538 - loss: 0.8598 -
Epoch 11/200
                       - 3s 346ms/step - accuracy: 0.7084 - loss: 0.9119 -
7/7 -
Epoch 12/200
7/7 -
                       - 2s 339ms/step - accuracy: 0.7086 - loss: 0.9051 -
Epoch 13/200
                        - 2s 341ms/step - accuracy: 0.7061 - loss: 0.9036 -
7/7 -
Epoch 14/200
                       - 3s 349ms/step - accuracy: 0.7462 - loss: 0.8609 -
7/7 -
Epoch 15/200
                      — 2s 339ms/step - accuracy: 0.7326 - loss: 0.8531 -
7/7 ----
Epoch 16/200
                        - 3s 339ms/step - accuracy: 0.8057 - loss: 0.7076 -
7/7 -
Epoch 17/200
                       — 3s 346ms/step — accuracy: 0.7050 — loss: 0.9211 —
7/7 -
Epoch 18/200
                       - 2s 351ms/step - accuracy: 0.7495 - loss: 0.7848 -
7/7 -
Epoch 19/200
                        - 3s 351ms/step - accuracy: 0.7498 - loss: 0.8005 -
7/7 -
Epoch 20/200
7/7 -
                       — 2s 339ms/step — accuracy: 0.7232 — loss: 0.9005 —
Epoch 21/200
7/7 -
                        - 3s 348ms/step - accuracy: 0.7883 - loss: 0.6885 -
Epoch 22/200
7/7 -
                        - 2s 341ms/step - accuracy: 0.8146 - loss: 0.6339 -
```

```
Epoch 23/200
7/7 —
                       — 2s 343ms/step - accuracy: 0.8132 - loss: 0.6357 -
Epoch 24/200
7/7 -
                        - 2s 345ms/step - accuracy: 0.7803 - loss: 0.6936 -
Epoch 25/200
7/7 -
                        - 3s 351ms/step - accuracy: 0.8023 - loss: 0.6354 -
Epoch 26/200
                        - 2s 349ms/step - accuracy: 0.8024 - loss: 0.6179 -
7/7 -
Epoch 27/200
                        - 3s 351ms/step - accuracy: 0.7927 - loss: 0.6709 -
7/7 -
Epoch 28/200
7/7 —
                        - 2s 351ms/step - accuracy: 0.7955 - loss: 0.6365 -
Epoch 29/200
7/7 -
                        • 3c 357ms/sten - accuracy: 0.8024 - loss: 0.6401 -
```

Тестирование

```
from petroscope.segmentation.eval import SegmDetailedTester
from tensorflow.keras.utils import to_categorical
tester = SegmDetailedTester(
   Path("output"),
   classes=classset,
   void pad=0,
   void_border_width=4,
   vis_plots=False,
   vis segmentation=True,
)
for img_p, mask_p in test_img_mask_p:
    img = load_image(img_p, normalize=True)
   pred = model.predict_image(img)
   print(f"Pred shape: {pred.shape}")
   mask = load_mask(mask_p, classes=classset, one_hot=False)
   mask_resized = tf.image.resize(mask[..., np.newaxis], (256, 256), method='r
   mask resized = mask resized[..., 0].numpy().astype(np.uint8)
   print(f"Mask resized shape: {mask_resized.shape}")
   pred_one_hot = to_categorical(pred, num_classes=len(classset.classes))
   mask_one_hot = to_categorical(mask_resized, num_classes=len(classset.classe
   print(f"Pred one-hot shape: {pred_one_hot.shape}")
   print(f"Mask one-hot shape: {mask_one_hot.shape}")
   metrics = tester.eval.evaluate(pred_one_hot, gt=mask_one_hot)
   metrics_void = tester.eval_void.evaluate(pred_one_hot, gt=mask_one_hot)
   print(f"Metrics for {img_p.name}:\n{metrics}")
```

```
print(f"Metrics with void borders for {img_p.name}:\n{metrics_void}")
print("-" * 50)
/usr/local/lib/python3.11/dist-packages/keras/src/models/functional.py:237:
Expected: ['keras_tensor']
Received: inputs=Tensor(shape=(1, 256, 256, 3))
  warnings.warn(msg)
1/1 -
                        4s 4s/step
Pred shape: (256, 256)
Mask resized shape: (256, 256)
Pred one-hot shape: (256, 256, 7)
Mask one-hot shape: (256, 256, 7)
Metrics for test_01.jpg:
          iou [soft]:
                  bg: 0.8814 [0.8814]
                  brt: 0.0010 [0.0010]
                  ccp/kub: 0.7368 [0.7368]
                  gl: 0.1642 [0.1642]
                  py/mrc: 0.9581 [0.9581]
                  sph: 0.7849 [0.7849]
                  tnt/ttr: 0.0074 [0.0074]
          mean iou [soft]: 0.5048 [0.5048]
          acc: 0.9280
Metrics with void borders for test_01.jpg:
          iou [soft]:
                  bg: 0.8814 [0.8814]
                  brt: 0.0010 [0.0010]
                  ccp/kub: 0.7368 [0.7368]
                  ql: 0.1642 [0.1642]
                  py/mrc: 0.9581 [0.9581]
                  sph: 0.7849 [0.7849]
                  tnt/ttr: 0.0074 [0.0074]
          mean iou [soft]: 0.5048 [0.5048]
         acc: 0.9280
                   Os 56ms/step
1/1 ———
Pred shape: (256, 256)
Mask resized shape: (256, 256)
Pred one-hot shape: (256, 256, 7)
Mask one-hot shape: (256, 256, 7)
Metrics for test 02.jpg:
          iou [soft]:
                  bq: 0.7972 [0.7972]
                  brt: 1.0000 [1.0000]
                  ccp/kub: 0.1273 [0.1273]
                  ql: 0.1905 [0.1905]
                  py/mrc: 0.9518 [0.9518]
                  sph: 0.5996 [0.5996]
                  tnt/ttr: 0.0000 [0.0000]
          mean iou [soft]: 0.5238 [0.5238]
```

acc: 0.8822

Metrics with void borders for test_02.jpg:
 iou [soft]:

bg: 0.7972 [0.7972] brt: 1.0000 [1.0000] ccp/kub: 0.1273 [0.1273] gl: 0.1905 [0.1905]

py/mrc: 0.9518 [0.9518] sph: 0.5996 [0.5996]