Практическое задание №2

Общая терминология по используемым данным

Предоставляемые данные для разработки моделей и алгоритмов трекинга мяча в теннисе представляют собор набор игр (game), состоящих из нескольких клипов (clip), каждый из которых состоит из набора кадров (frame). Обратите внимание на структуру организации файлов внутри предоставляемого датасета для полного понимания.

Большинство алгоритмов трекинга объектов работают с несколькими последовательными кадрами, и в данном задании также подразумевается использование этого приема. Последовательность нескольких кадров будем именовать стопкой (stack), размер стопки (stack_s) является гиперпараметром разрабатываемого алгоритма.

Заготовка решения

Загрузка датасета

Для работы с данными в ноутбуке kaggle необходимо подключить датасет. File -> Add or upload data, далее в поиске написать tennis-tracking-assignment и выбрать датасет. Если поиск не работает, то можно добавить датасет по url: https://www.kaggle.com/xubiker/tennistrackingassignment. После загрузки данные датасета будут примонтированы в .../input/tennistrackingassignment.

Установка и импорт зависимостей

Установка необходимых пакетов (не забудьте "включить интернет" в настройках ноутбука kaggle):

In []:

!pip install moviepy --upgrade

!pip install gdown

После установки пакетов для корректной работы надо обязательно перезагрузить ядро. Run -> Restart and clear cell outputs. Без сего действа будет ошибка при попытке обращения к библиотеке moviepy при сохранении визуализации в виде видео. Может когда-то авторы библиотеки это починят...

Импорт необходимых зависимостей:

In [2]:

from pathlib import Path

from typing import List, Tuple, Sequence

import numpy as np

from numpy import unravel_index

from PIL import Image, ImageDraw, ImageFont

from tqdm import tqdm, notebook

from moviepy.video.io.lmageSequenceClip import ImageSequenceClip

import math

from scipy.ndimage import gaussian filter

import gc

import time

import random

import csv

from keras.layers import Input, Conv2D, MaxPooling2D, UpSampling2D, Concatenate, BatchNormalization, Activation from keras.models import Model

import tensorflow as tf

from pathlib import Path

from scipy.ndimage import label

from skimage.measure import regionprops

Набор функций для загрузки данных из датасета

Функция load_clip_data загружает выбранный клип из выбранной игры и возвращает его в виде numpy массива [n_frames, height, width, 3] типа uint8. Для ускорения загрузки используется кэширование - однажды загруженные клипы хранятся на диске в виде npz архивов, пpи последующем обращении к таким клипам происходит загрузка npz архива.

Также добавлена возможность чтения клипа в половинном разрешении 640х360, вместо оригинального 1280х720 для упрощения и ускорения разрабатываемых алгоритмов.

```
Функция load_clip_labels загружает референсные координаты мяча в клипе в виде numpy массива [n_frames, 4], где в каждой строке массива содержатся значения [code, x, y, q]. x, y соответствуют координате центра мяча на кадре, q не используется в данном задании, code oписывает статус мяча:

• code = 0 - мяча в кадре нет
• code = 1 - мяч присутствует в кадре и легко идентифицируем
• code = 2 - мяч присутствует в кадре, но сложно идентифицируем
• code = 3 - мяч присутствует в кадре, но заслонен другими объектами.

При загрузке в половинном разрешении координаты x, y делятся на 2.

Функция load_clip загружает выбранный клип и соответствующий массив координат и возвращает их в виде пары.

In [3]:

def get_num_clips(path: Path, game: int) -> int:
    return len(list((path / fgame{game}/").iterdir()))
```

def get_game_clip_pairs(path: Path, games: List[int]) -> List[Tuple[int, int]]:

print(floading clip data (game {game}, clip {clip}) {suffix}')

clip data = np.load(cache path / cached data name)['clip data']

img = img.resize((img.width // 2, img.height // 2),)

np.savez compressed(cache path / cached data name, clip data=clip data)

def load clip labels(path: Path, game: int, clip: int, downscale: bool, quiet=False):

cached data name = f'{game} {clip}{resize code}.npz'

suffix = 'downscaled' if downscale else "

if (cache path / cached data name).exists():

clip_path = path / f'game{game}/clip{clip}'
n imgs = len(list(clip path.iterdir())) - 1

for i in notebook.tqdm(range(n_imgs)):
 img = Image.open(clip_path / f'{i:04d}.jpg')

imgs[i] = np.array(img, dtype=np.uint8)

cache path.mkdir(exist ok=True, parents=True)

print(f'loading clip labels (game {game}, clip {clip})')

values = np.array([-1 if i == " else int(i) for i in line[1:]])

def load clip(path: Path, game: int, clip: int, downscale: bool, quiet=False):

data = load_clip_data(path, game, clip, downscale, quiet) labels = load_clip_labels(path, game, clip, downscale, quiet)

clip_path = path / f'game{game}/clip{clip}'

lines = list(csv.reader(csvfile))

for line in lines[1:]:

if downscale:
 values[1] //= 2
 values[2] //= 2
labels.append(values)

return np.stack(labels)

return data, labels

with open(clip_path / 'labels.csv') as csvfile:

if not quiet:

else:

cache_path = path / 'cache'
cache_path.mkdir(exist_ok=True)
resize code = ' ds2' if downscale else "

imgs = [None] * n imgs

clip data = np.stack(imgs)

if downscale:

return clip data

if not quiet:

labels = []

return [(game, c) for game in games for c in range(1, get num clips(path, game) + 1)]

def load clip data(path: Path, game: int, clip: int, downscale: bool, quiet=False) -> np.ndarray:

Набор дополнительных функций

Еще несколько функций, немного облегчающих выполнение задания:

- ргераге_expariment создает новую директорию в out_path для хранения результатов текущего эксперимента. Нумерация выполняется автоматически, функция возвращает путь к созданной директории эксперимента;
- ball_gauss_template создает "шаблон" мяча, может быть использована в алгоритмах поиска мяча на изображении по корреляции;
- create_masks принимает набор кадров и набор координат мяча, и генерирует набор масок, в которых помещает шаблон мяча на заданные координаты. Может быть использована при обучении нейронной сети семантической сегментации;

```
In [4]:
def prepare experiment(out path: Path) -> Path:
  out path.mkdir(parents=True, exist ok=True)
  dirs = [d for d in out path.iterdir() if d.is dir() and d.name.startswith('exp')]
  experiment_id = max(int(d.name.split('_')[1]) for d in dirs) + 1 if dirs else 1
  exp path = out path / f'exp {experiment id}'
  exp path.mkdir()
  return exp path
def ball gauss template(rad, sigma):
  x, y = np.meshgrid(np.linspace(-rad, rad, 2 * rad + 1), np.linspace(-rad, rad, 2 * rad + 1))
  dst = np.sqrt(x * x + y * y)
  gauss = np.exp(-(dst ** 2 / (2.0 * sigma ** 2)))
  return gauss
def create_masks(data: np.ndarray, labels: np.ndarray, resize):
  rad = 64 #25
  sigma = 10
  if resize:
     rad //= 2
  ball = ball gauss template(rad, sigma)
  n frames = data.shape[0]
  sh = rad
  masks = []
  for i in range(n frames):
     label = labels[i, ...]
     frame = data[i, ...]
     if 0 < label[0] < 3:
       x, y = label[1:3]
       mask = np.zeros((frame.shape[0] + 2 * rad + 2 * sh, frame.shape[1] + 2 * rad + 2 * sh), np.float32)
       mask[y + sh : y + sh + 2 * rad + 1, x + sh : x + sh + 2 * rad + 1] = ball
       mask = mask[rad + sh : -rad - sh, rad + sh : -rad - sh]
       masks.append(mask)
     else:
       masks.append(np.zeros((frame.shape[0], frame.shape[1]), dtype=np.float32))
  return np.stack(masks)
```

Набор функций, предназначенных для визуализации результатов

Функция visualize_prediction принимает набор кадров, набор координат детекции мяча (можно подавать как референсные значения, так и предсказанные) и создает видеоклип, в котором отрисовывается положение мяча, его трек, номер кадра и метрика качества трекинга (если она была передана в функцию). Видеоклип сохраняется в виде mp4 файла. Кроме того данная функция создает текстовый файл, в который записывает координаты детекции мяча и значения метрики качества трекинга.

Функция visualize_prob принимает набор кадров и набор предсказанных карт вероятности и создает клип с наложением предсказанных карт вероятности на исходные карты. Области "подсвечиваются" желтым, клип сохраняется в виде mp4 видеофайла. Данная функция может быть полезна при наличии в алгоритме трекинга сети, осуществляющей семантическую сегментацию.

```
In [5]:
def _add_frame_number(frame: np.ndarray, number: int) -> np.ndarray:
  fnt = ImageFont.load_default() # ImageFont.truetype("arial.ttf", 25)
  img = Image.fromarray(frame)
  draw = ImageDraw.Draw(img)
  draw.text((10, 10), f'frame {number}', font=fnt, fill=(255, 0, 255))
  return np.array(img)
```

```
n_frames = data.shape[0]
  frames_res = []
  fnt = ImageFont.load_default() # ImageFont.truetype("arial.ttf", 25)
  for i in range(n frames):
     img = Image.fromarray(data[i, ...])
     draw = ImageDraw.Draw(img)
     txt = f'frame {i}'
     if metrics is not None:
       txt += f', SiBaTrAcc: {metrics[i]:.3f}'
     draw.text((10, 10), txt, font=fnt, fill=(255, 0, 255))
     label = lbls[i]
     if label[0] != 0: # the ball is clearly visible
       px, py = label[1], label[2]
       draw.ellipse((px - ball rad, py - ball rad, py + ball rad, py + ball rad), outline=color, width=2)
       for q in range(track length):
          if lbls[i-q-1][0] == 0:
            break
          if i - q > 0:
            draw.line((lbls[i - q - 1][1], lbls[i - q - 1][2], lbls[i - q][1], lbls[i - q][2]), fill=color)
     frames_res.append(np.array(img))
  return frames res
def save clip(frames: Sequence[np.ndarray], path: Path, fps):
  assert path.suffix in ('.mp4', '.gif')
  clip = ImageSequenceClip(frames, fps=fps)
  if path.suffix == '.mp4':
     clip.write videofile(str(path), fps=fps, logger=None)
  else:
     clip.write_gif(str(path), fps=fps, logger=None)
def to yellow heatmap(frame: np.ndarray, pred frame: np.ndarray, alpha=0.4):
  img = Image.fromarray((frame * alpha).astype(np.uint8))
  maskR = (pred_frame * (1 - alpha) * 255).astype(np.uint8)
  maskG = (pred frame * (1 - alpha) * 255).astype(np.uint8)
  maskB = np.zeros like(maskG, dtype=np.uint8)
  mask = np.stack([maskR, maskG, maskB], axis=-1)
  return img + mask
def vis pred heatmap(data full: np.ndarray, pred prob: np.ndarray, display frame number):
  n frames = data full.shape[0]
  v_frames = []
  for i in range(n_frames):
     frame = data_full[i, ...]
     pred = pred_prob[i, ...]
     hm = _to_yellow_heatmap(frame, pred)
     if display_frame_number:
       hm = _add_frame_number(hm, i)
     v_frames.append(hm)
  return v frames
def visualize prediction(data full: np.ndarray, labels pr: np.ndarray, save path: Path, name: str, metrics=None, fps=15):
  with open(save_path / f'{name}.txt', mode='w') as f:
     if metrics is not None:
       f.write(f'SiBaTrAcc: {metrics[-1]} \n')
     for i in range(labels pr.shape[0]):
       v = vis clip(data full, labels pr, metrics)
  save clip(v, save path / f'{name}.mp4', fps=fps)
def visualize_prob(data: np.ndarray, pred_prob: np.ndarray, save_path: Path, name: str, frame_number=True, fps=15):
  v pred = vis pred heatmap(data, pred prob, frame number)
  _save_clip(v_pred, save_path / f'{name}_prob.mp4', fps=fps)
```

print('perforing clip visualization')

Класс DataGenerator

Класс, отвечающий за генерацию данных для обучения модели. Принимает на вход путь к директории с играми, индексы игр, используемые для генерации данных, и размер стопки. Хранит в себе автоматически обновляемый пул с клипами игр.

В пуле содержится pool_s клипов. DataGenerator позволяет генерировать батч из стопок (размера stack_s) последовательных кадров. Выбор клипа для извлечения данных взвешенно-случайный: чем больше длина клипа по сравнению с другими клипами в пуле, тем вероятнее, что именно из него будет сгенерирована стопка кадров. Выбор стопки кадров внтури выбранного клипа полностью случаен. Кадры внутри стопки конкатенируются по последнему измерению (каналам).

После генерирования количества кадров равного общему количеству кадров, хранимых в пуле, происходит автоматическое обновление пула: из пула извлекаются pool_update_s случайных клипов, после чего в пул загружается pool_update_s случайных клипов, не присутствующих в пуле. В случае, если размер пула pool_s больше или равен суммарному количеству клипов в играх, переданных в конструктор, все клипы сразу загружаются в пул, и автообновление не производится.

Использование подобного пула позволяет работать с практически произвольным количеством клипов, без необходимости загружать их всех в оперативную память.

Для вашего удобства функция извлечения стопки кадров из пула помимо самой стопки также создает и возвращает набор сгенерированных масок с мячом исходя из референсных координат мяча в клипе.

Функция random_g принимает гиперпараметр размера стопки кадров и предоставляет генератор, возвращающий стопки кадров и соответствующие им маски. Данный генератор может быть использован при реализации решения на tensorflow. Обновление пула происходит автоматически, об этом беспокоиться не нужно.

def __init__(self, path: Path, games: List[int], stack_s, downscale, pool_s=30, pool_update_s=10, pool_autoupdate=**True**, quiet=**False**) -:

In [6]:

```
class DataGenerator:
```

print(f'items in pool: {len(self.pool)} - {self.pool.keys()}')

```
self.path = path
  self.stack_s = stack_s
  self.downscale = downscale
  self.pool_size = pool_s
  self.pool update size = pool update s
  self.pool autoupdate = pool autoupdate
  self.quiet = quiet
  self.data = []
  self.masks = []
  self.frames in pool = 0
  self.produced frames = 0
  self.game clip pairs = get game clip pairs(path, list(set(games)))
  self.game clip pairs loaded = []
  self.game clip pairs not loaded = list.copy(self.game clip pairs)
  self.pool = {}
  self._first_load()
def first load(self):
  # --- if all clips can be placed into pool at once, there is no need to refresh pool at all ---
  if len(self.game_clip_pairs) <= self.pool_size:</pre>
    for gcp in self.game_clip_pairs:
       self._load(gcp)
     self.game clip pairs loaded = list.copy(self.game clip pairs)
    self.game clip pairs not loaded.clear()
    self.pool_autoupdate = False
    self. load to pool(self.pool size)
  self. update clip weights()
def load(self, game clip pair):
  game, clip = game_clip_pair
  data, labels = load clip(self.path, game, clip, self.downscale, quiet=self.quiet)
  masks = create masks(data, labels, self.downscale)
  weight = data.shape[0] if data.shape[0] >= self.stack s else 0
  self.pool[game_clip_pair] = (data, labels, masks, weight)
  self.frames_in_pool += data.shape[0] - self.stack_s + 1
  # print(f'items in pool: {len(self.pool)} - {self.pool.keys()}')
def remove(self, game clip pair):
  value = self.pool.pop(game clip pair)
  self.frames in pool -= value[0].shape[0] - self.stack s + 1
```

```
def _update_clip_weights(self):
     weights = [self.pool[pair][-1] for pair in self.game_clip_pairs_loaded]
     tw = sum(weights)
     self.clip weights = [w / tw for w in weights]
     # print(f'clip weights: {self.clip weights}')
  def remove from pool(self, n):
     # --- remove n random clips from pool ---
     if len(self.game clip pairs loaded) >= n:
       remove pairs = random.sample(self.game clip pairs loaded, n)
       for pair in remove pairs:
          self. remove(pair)
          self.game clip pairs loaded.remove(pair)
          self.game clip pairs not loaded.append(pair)
       gc.collect()
  def _load_to_pool(self, n):
     # --- add n random clips to pool ---
     gc.collect()
     add_pairs = random.sample(self.game_clip_pairs_not_loaded, n)
     for pair in add_pairs:
       self._load(pair)
       self.game clip pairs not loaded.remove(pair)
       self.game clip pairs loaded.append(pair)
  def update pool(self):
     self._remove_from_pool(self.pool_update_size)
     self. load to pool(self.pool update size)
     self. update clip weights()
  def get random stack(self):
     pair idx = np.random.choice(len(self.game clip pairs loaded), 1, p=self.clip weights)[0]
     game clip pair = self.game clip pairs loaded[pair idx]
     d, , m, = self.pool[game clip pair]
     start = np.random.choice(d.shape[0] - self.stack s, 1)[0]
     frames stack = d[start : start + self.stack_s, ...]
     frames stack = np.squeeze(np.split(frames stack, indices or sections=self.stack s, axis=0))
     frames stack = np.concatenate(frames stack, axis=-1)
     mask = m[start + self.stack s - 1, ...]
     return frames stack, mask
  def get random batch(self, batch s):
     imgs, masks = [], []
     while len(imgs) < batch_s:
       frames_stack, mask = self.get_random_stack()
       imgs.append(frames_stack)
       masks.append(mask)
     if self.pool_autoupdate:
       self.produced_frames += batch_s
       # print(f'produced frames: {self.produced_frames} from {self.frames_in_pool}')
       if self.produced_frames >= self.frames_in_pool:
          self.update pool()
          self_produced frames = 0
     return np.stack(imgs), np.stack(masks)
  def random g(self, batch s):
     while True:
       imgs batch, masks batch = self.get random batch(batch s)
       yield imgs batch, masks batch
Пример использования DataGenerator
```

```
In [7]:
from pathlib import Path
print(Path('/kaggle/input/tennistrackingassignment/train').exists())
```

Рекомендованный размер пула pool s=10 в случае использования уменьшенных вдвое изображений. При большем размере пула есть большая вероятность нехватки имеющихся 13G оперативной памяти. Используйте параметр quiet=True в конструкторе DataGenerator, если хотите скрыть все сообщения о чтении данных и обновлении пула.

In [8]:

```
imgs, masks = train gen.get random batch(batch s)
   print(imgs.shape, imgs.dtype, masks.shape, masks.dtype)
loading clip data (game 4, clip 6) downscaled
loading clip labels (game 4, clip 6)
loading clip data (game 3, clip 7) downscaled
loading clip labels (game 3, clip 7)
loading clip data (game 2, clip 4) downscaled
loading clip labels (game 2, clip 4)
loading clip data (game 1, clip 13) downscaled
loading clip labels (game 1, clip 13)
loading clip data (game 2, clip 1) downscaled
loading clip labels (game 2, clip 1)
loading clip data (game 1, clip 4) downscaled
loading clip labels (game 1, clip 4)
loading clip data (game 2, clip 7) downscaled
loading clip labels (game 2, clip 7)
loading clip data (game 4, clip 15) downscaled
loading clip labels (game 4, clip 15)
loading clip data (game 4, clip 9) downscaled
loading clip labels (game 4, clip 9)
loading clip data (game 2, clip 9) downscaled
loading clip labels (game 2, clip 9)
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
(4, 360, 640, 9) uint8 (4, 360, 640) float32
import matplotlib.pyplot as plt
stack s = 3
train gen = DataGenerator(Path('../input/tennistrackingassignment/train/'), [1], stack s=stack s, downscale=True, pool s=10, pool update
stack, mask = train gen.get random stack()
print(stack.shape, mask.shape)
for i in range(stack_s):
  plt.figure()
```

train_gen = DataGenerator(Path('../input/tennistrackingassignment/train/'), [1, 2, 3, 4], stack_s=stack_s, downscale=**True**, pool_s=10, pool_

 $stack_s = 3$ $batch_s = 4$

for i in range(10):

plt.imshow(stack[:, :, 3 * i: 3 * i + 3])

loading clip data (game 1, clip 2) downscaled loading clip labels (game 1, clip 2) loading clip data (game 1, clip 9) downscaled loading clip labels (game 1, clip 9) loading clip data (game 1, clip 8) downscaled loading clip labels (game 1, clip 8) loading clip data (game 1, clip 6) downscaled loading clip labels (game 1, clip 6) loading clip data (game 1, clip 3) downscaled loading clip labels (game 1, clip 3) loading clip data (game 1, clip 13) downscaled loading clip labels (game 1, clip 13) loading clip data (game 1, clip 5) downscaled loading clip labels (game 1, clip 5) loading clip data (game 1, clip 12) downscaled loading clip labels (game 1, clip 12) loading clip data (game 1, clip 4) downscaled loading clip labels (game 1, clip 4) loading clip data (game 1, clip 10) downscaled loading clip labels (game 1, clip 10) (360, 640, 9) (360, 640)







Класс Metrics

Класс для вычисления метрики качества трекинга SiBaTrAcc. Функция evaluate_predictions принимает массив из референсных и предсказанных координат мяча для клипа и возвращает массив аккумулированных значений SiBaTrAcc (может быть полезно для визуализации результатов предсказания) и итоговое значение метрики SiBaTrAcc.

```
In [10]: class Metrics:
```

```
@staticmethod
def position error(label gt: np.ndarray, label pr: np.ndarray, step=8, alpha=1.5, e1=5, e2=5):
  # at codes:
  # 0 - the ball is not within the image
  #1 - the ball can easily be identified
  #2 - the ball is in the frame, but is not easy to identify
  #3 - the ball is occluded
  if label_gt[0] != 0 and label_pr[0] == 0:
    return e1
  if label gt[0] == 0 and label pr[0] != 0:
    return e2
  dist = math.sqrt((label_gt[1] - label_pr[1]) ** 2 + (label_gt[2] - label_pr[2]) ** 2)
  pe = math.floor(dist / step) ** alpha
  pe = min(pe, 5)
  return pe
@staticmethod
def evaluate_predictions(labels_gt, labels_pr) -> Tuple[List[float], float]:
  pe = [Metrics.position error(labels gt[i, ...], labels pr[i, ...]) for i in range(len(labels gt))]
  SIBATRACC = []
  for i, in enumerate(pe):
    SIBATRACC.append(1 - sum(pe[: i + 1]) / ((i + 1) * 5))
  SIBATRACC_total = 1 - sum(pe) / (len(labels_gt) * 5)
  return SIBATRACC, SIBATRACC total
```

Основной класс модели SuperTrackingModel

Реализует всю логику обучения, сохранения, загрузки и тестирования разработанной модели трекинга. Этот класс можно и нужно расширять.

В качестве примера вам предлагается заготовка модели, в которой трекинг осуществляется за счет предсказания маски по входному батчу и последующему предсказанию координат мяча по полученной маски. В данном варианте вызов функции предсказания координат по клипу (predict) повлечет за собой разбиение клипа на батчи, вызов предсказания маски для каждого батча, склеивание результатов в последовательность масок, вызов функции по вычислению координат мяча по маскам и возвращения результата. Описанные действия уже реализованы, вам остается только написать функции predict_on_bath и get_labels_from_prediction. Эта же функция predict используется и в вызове функции test, дополнительно вычисляя метрику качества трекинга и при необходимости визуализируя результат тестирования. Обратите внимание, что в результирующем питру массиве с координатами помимо значений х и у первым значением в каждой строке должно идти значение соde (0, если мяча в кадре нет и > 0, если мяч в кадре есть) для корректного вычисления качества трекинга.

Вам разрешается менять логику работы класса модели, (например, если решение не подразумевает использование масок), но при этом логика и работа функций load и test должна остаться неизменной! In [11]:

```
# Define U-Net architecture
def unet_block(input_layer, filters, kernel_size=(3, 3), activation="relu"):
  conv = Conv2D(filters, kernel_size, padding="same")(input_layer)
  conv = Activation(activation)(conv)
  conv = BatchNormalization()(conv)
  return conv
def UNetForTrack(input_size):
  inputs = Input(shape=input size)
  # Encoder
  conv1 = unet block(inputs, 64)
  conv2 = unet_block(conv1, 64)
  pool1 = MaxPooling2D(pool_size=(2, 2))(conv2)
  conv3 = unet block(pool1, 128)
  conv4 = unet_block(conv3, 128)
  pool2 = MaxPooling2D(pool_size=(2, 2))(conv4)
  conv5 = unet block(pool2, 256)
  conv6 = unet block(conv5, 256)
  pool3 = MaxPooling2D(pool_size=(2, 2))(conv6)
  # Bottleneck
  conv7 = unet block(pool3, 512)
  conv8 = unet_block(conv7, 512)
  # Decoder
  up1 = UpSampling2D(size=(2, 2))(conv8)
  concat1 = Concatenate()([up1, conv6])
  conv9 = unet block(concat1, 256)
  conv10 = unet_block(conv9, 256)
  up2 = UpSampling2D(size=(2, 2))(conv10)
  concat2 = Concatenate()([up2, conv4])
  conv11 = unet block(concat2, 128)
  conv12 = unet_block(conv11, 128)
  up3 = UpSampling2D(size=(2, 2))(conv12)
  concat3 = Concatenate()([up3, conv2])
  conv13 = unet_block(concat3, 64)
  conv14 = unet_block(conv13, 64)
  outputs = Conv2D(1, (1, 1), activation="sigmoid")(conv14)
  model = Model(inputs, outputs)
  return model
In [12]:
import tensorflow as tf
class SaveEveryNEpochs(tf.keras.callbacks.Callback):
  def __init__(self, save_interval, save_path):
     super(SaveEveryNEpochs, self).__init__()
     self.save_interval = save_interval
     self.save_path = save_path
  def on_epoch_end(self, epoch, logs=None):
     if (epoch + 1) % self.save_interval == 0:
       print(f"Epoch {epoch + 1}: Saving model...")
       self.model.save weights(f'{self.save path} epoch {epoch + 1}.weights.h5')
In [26]:
class SuperTrackingModel:
  def init (self, batch s, stack s, out path, downscale, height=720, width=1280):
     self.height = height
     self.width = width
     if downscale:
       self.height //= 2
       self.width //= 2
     self.batch_s = batch_s
     self.stack s = stack s
     self.model = UNetForTrack((self.height, self.width, 3 * self.stack s))
     self.out_path = out_path
     self.downscale = downscale
```

```
def save(self, name: str):
  print("Saving to folder /working")
  self.model.save_weights(f'/kaggle/working/{name}.weights.h5')
def load(self, name: str):
  print(f"Loading model weights for {name}")
  self.model.load weights(f'/kaggle/working/{name}.weights.h5')
  print('Loading model done.')
def predict on batch(self, batch: np.ndarray) -> np.ndarray:
  predictions = self.model.predict(batch).reshape((batch.shape[0], batch.shape[1], batch.shape[2]))
  return predictions
def predict prob on clip(self, clip: np.ndarray) -> np.ndarray:
  print('doing predictions')
  n_frames = clip.shape[0]
  # --- get stacks ---
  stacks = []
  for i in range(n frames - self.stack s + 1):
    stack = clip[i : i + self.stack_s, ...]
    stack = np.squeeze(np.split(stack, self.stack_s, axis=0))
    stack = np.concatenate(stack, axis=-1)
    stacks.append(stack)
  # --- round to batch size ---
  add stacks = 0
  while len(stacks) % self.batch s != 0:
    stacks.append(stacks[-1])
    add stacks += 1
  # --- group into batches ---
  batches = []
  for i in range(len(stacks) // self.batch_s):
    batch = np.stack(stacks[i * self.batch s: (i + 1) * self.batch s])
    batches.append(batch)
  stacks.clear()
  # --- perform predictions ---
  predictions = []
  for batch in batches:
     pred = np.squeeze(self.predict on batch(batch))
    predictions.append(pred)
  # --- crop back to source length ---
  predictions = np.concatenate(predictions, axis=0)
  if (add stacks > 0):
    predictions = predictions[:-add_stacks, ...]
  batches.clear()
  # --- add (stack s - 1) null frames at the begining ---
  start_frames = np.zeros((stack_s - 1, predictions.shape[1], predictions.shape[2]), dtype=np.float32)
  predictions = np.concatenate((start_frames, predictions), axis=0)
  print('predictions are made')
  return predictions
def get_labels_from_prediction(self, pred_prob: np.ndarray, upscale_coords: bool) -> np.ndarray:
  n frames = pred prob.shape[0]
  coords = np.zeros([n frames, 3])
  for i in range(n frames):
    prediction mask = pred prob[i]
    if prediction mask.sum() < 1:</pre>
       coords[i] = [0, -1, -1]
       prediction mask[prediction mask < 0.5] = 0
       prediction mask[prediction mask >= 0.5] = 1
       code = 0
       x, y = -1, -1
       labeled, num_features = label(prediction_mask)
       if num features > 0:
          regions = regionprops(labeled)
          largest_region = max(regions, key=lambda r: r.area)
          y, x = largest_region.centroid
          code = 1
       if upscale coords:
          x, y = 2 * x, 2 * y
```

```
coords[i] = [code, x, y]
     return coords
  def predict(self, clip: np.ndarray, upscale coords=True) -> tuple[np.ndarray, np.ndarray]:
     prob pr = self. predict prob on clip(clip)
     labels pr = self.get labels from prediction(prob pr, upscale coords)
     return labels pr, prob pr
   def test(self, data path: Path, games: list, do visualization=False, test name='test'):
     game clip pairs = get game clip pairs(data path, games)
     SIBATRACC vals = []
     for game, clip in game clip pairs:
        data = load clip data(data path, game, clip, downscale=self.downscale)
       if do visualization:
          data_full = load_clip_data(data_path, game, clip, downscale=False) if self.downscale else data
       labels_gt = load_clip_labels(data_path, game, clip, downscale=False)
       labels_pr, prob_pr = self.predict(data, upscale_coords=True)
       SIBATRACC_per_frame, SIBATRACC_total = Metrics.evaluate_predictions(labels_gt, labels_pr)
       SIBATRACC vals.append(SIBATRACC total)
       if do visualization:
          visualize_prediction(data_full, labels_pr, self.out_path, f'{test_name}_g{game}_c{clip}', SIBATRACC_per_frame)
          visualize_prob(data, prob_pr, self.out_path, f'{test_name}_g{game}_c{clip}')
          del data full
       del data, labels gt, labels pr, prob pr
        gc.collect()
     SIBATRACC_final = sum(SIBATRACC_vals) / len(SIBATRACC_vals)
     return SIBATRACC_final
  def train(self, train generator, val gen, epochs=15, save interval=5):
     self.epochs = epochs
     save path = '/kaggle/working/unet model'
     save callback = SaveEveryNEpochs(save interval, save path)
     self.model.compile(optimizer=tf.keras.optimizers.Adam(learning rate=1e-4),
                  loss=tf.keras.losses.BinaryCrossentropy())
     self.model.fit(train generator, validation data=val gen, epochs=self.epochs,
               steps_per_epoch=1000 // self.batch_s, validation_steps=200 // self.batch_s,
               callbacks=[save callback])
     self.save('unet last model')
     print('Training done')
Пример пайплайна для обучения модели:
In [13]:
batch s = 5
stack s = 3
downscale = True
output path = prepare experiment(Path('/kaggle/working'))
train_gen = DataGenerator(Path('../input/tennistrackingassignment/train/'), [1, 2, 3, 5], stack_s=stack_s, downscale=True, pool_s=10, pool_
val gen = DataGenerator(Path("../input/tennistrackingassignment/train/"), [4, 5], stack s=stack s, downscale=True, pool s=4, pool update
loading clip data (game 5, clip 5) downscaled
loading clip labels (game 5, clip 5)
loading clip data (game 2, clip 8) downscaled
loading clip labels (game 2, clip 8)
loading clip data (game 5, clip 2) downscaled
loading clip labels (game 5, clip 2)
loading clip data (game 3, clip 1) downscaled
loading clip labels (game 3, clip 1)
loading clip data (game 3, clip 6) downscaled
loading clip labels (game 3, clip 6)
loading clip data (game 2, clip 3) downscaled
loading clip labels (game 2, clip 3)
loading clip data (game 1, clip 2) downscaled
loading clip labels (game 1, clip 2)
loading clip data (game 2, clip 1) downscaled
loading clip labels (game 2, clip 1)
loading clip data (game 2, clip 4) downscaled
loading clip labels (game 2, clip 4)
loading clip data (game 3, clip 4) downscaled
loading clip labels (game 3, clip 4)
loading clip data (game 5, clip 1) downscaled
loading clip labels (game 5, clip 1)
loading clip data (game 5, clip 5) downscaled
loading clip labels (game 5, clip 5)
loading clip data (game 4, clip 7) downscaled
loading clip labels (game 4, clip 7)
loading clip data (game 5, clip 6) downscaled
```

loading clip labels (game 5, clip 6)

```
In [21]:
model = SuperTrackingModel(batch s, stack s, out path=output path, downscale=downscale)
model.train(train gen.random g(batch s), val gen.random g(batch s))
Epoch 1/15
200/200
                                    299s 1s/step - loss: 0.6981 - val loss: 0.6251
Epoch 2/15
136/200
                                    59s 934ms/step - loss: 0.5753 loading clip data (game 2, clip 3) downscaled
137/200
                                    58s 934ms/step - loss: 0.5752loading clip labels (game 2, clip 3)
loading clip data (game 1, clip 5) downscaled
loading clip labels (game 1, clip 5)
loading clip data (game 1, clip 9) downscaled
loading clip labels (game 1, clip 9)
loading clip data (game 5, clip 9) downscaled
loading clip labels (game 5, clip 9)
200/200
                                    205s 1s/step - loss: 0.5661 - val_loss: 0.4714
Epoch 3/15
200/200
                                    199s 995ms/step - loss: 0.4487 - val_loss: 0.3498
Epoch 4/15
42/200
                                   2:27 933ms/step - loss: 0.3563loading clip data (game 2, clip 9) downscaled
43/200
                                   2:26 933ms/step - loss: 0.3562loading clip labels (game 2, clip 9)
loading clip data (game 5, clip 5) downscaled
44/200
                                   2:25 933ms/step - loss: 0.3561loading clip labels (game 5, clip 5)
loading clip data (game 2, clip 8) downscaled
loading clip labels (game 2, clip 8)
loading clip data (game 3, clip 5) downscaled
loading clip labels (game 3, clip 5)
200/200
                                    0s 955ms/step - loss: 0.3359loading clip data (game 4, clip 2) downscaled
loading clip labels (game 4, clip 2)
loading clip data (game 4, clip 15) downscaled
loading clip labels (game 4, clip 15)
200/200
                                    205s 1s/step - loss: 0.3358 - val loss: 0.2636
Epoch 5/15
119/200
                                    1:15 934ms/step - loss: 0.2516loading clip data (game 3, clip 2) downscaled
123/200
                                    1:11 934ms/step - loss: 0.2512loading clip labels (game 3, clip 2)
loading clip data (game 1, clip 5) downscaled
loading clip labels (game 1, clip 5)
loading clip data (game 1, clip 12) downscaled
loading clip labels (game 1, clip 12)
loading clip data (game 1, clip 4) downscaled
loading clip labels (game 1, clip 4)
                                    0s 955ms/step - loss: 0.2438Epoch 5: Saving model...
200/200
200/200
                                    203s 1s/step - loss: 0.2437 - val loss: 0.1836
Epoch 6/15
200/200
                                    199s 997ms/step - loss: 0.1759 - val_loss: 0.1324
Epoch 7/15
 7/200
                                   3:00 933ms/step - loss: 0.1376loading clip data (game 1, clip 6) downscaled
loading clip labels (game 1, clip 6)
loading clip data (game 1, clip 8) downscaled
 8/200
                                   2:59 934ms/step - loss: 0.1375loading clip labels (game 1, clip 8)
loading clip data (game 1, clip 1) downscaled
 9/200
                                   2:58 933ms/step - loss: 0.1374loading clip labels (game 1, clip 1)
loading clip data (game 3, clip 6) downscaled
12/200
                                   2:55 934ms/step - loss: 0.1373loading clip labels (game 3, clip 6)
200/200
                                    0s 931ms/step - loss: 0.1282loading clip data (game 4, clip 10) downscaled
loading clip labels (game 4, clip 10)
loading clip data (game 4, clip 5) downscaled
loading clip labels (game 4, clip 5)
200/200
                                    200s 1s/step - loss: 0.1282 - val loss: 0.0995
Epoch 8/15
33/200
                                   2:36 936ms/step - loss: 0.1006loading clip data (game 1, clip 10) downscaled
loading clip labels (game 1, clip 10)
loading clip data (game 3, clip 1) downscaled
                                   2:29 937ms/step - loss: 0.1003loading clip labels (game 3, clip 1)
loading clip data (game 3, clip 5) downscaled
41/200
                                   2:28 937ms/step - loss: 0.1003loading clip labels (game 3, clip 5)
loading clip data (game 1, clip 7) downscaled
loading clip labels (game 1, clip 7)
200/200
                                   - 202s 1s/step - loss: 0.0948 - val_loss: 0.0769
Epoch 9/15
200/200
                                    0s 936ms/step - loss: 0.0718loading clip data (game 5, clip 3) downscaled
loading clip labels (game 5, clip 3)
loading clip data (game 5, clip 1) downscaled
loading clip labels (game 5, clip 1)
200/200
                                    200s 999ms/step - loss: 0.0718 - val. loss: 0.0595
Epoch 10/15
                                   2:33 938ms/step - loss: 0.0581loading clip data (game 2, clip 1) downscaled
36/200
37/200
                                   2:32 938ms/step - loss: 0.0581loading clip labels (game 2, clip 1)
loading clip data (game 2, clip 5) downscaled
loading clip labels (game 2, clip 5)
loading clip data (game 5, clip 2) downscaled
38/200
                                   2:31 938ms/step - loss: 0.0581loading clip labels (game 5, clip 2)
loading clip data (game 5, clip 8) downscaled
39/200
                                   2:31 938ms/step - loss: 0.0581loading clip labels (game 5, clip 8)
200/200
                                    0s 937ms/step - loss: 0.0556Epoch 10: Saving model...
200/200
                                    200s 1s/step - loss: 0.0555 - val_loss: 0.0483
Epoch 11/15
```

```
5s 936ms/step - loss: 0.0443loading clip data (game 3, clip 2) downscaled
197/200
                                   2s 936ms/step - loss: 0.0442loading clip labels (game 3, clip 2)
loading clip data (game 1, clip 5) downscaled
loading clip labels (game 1, clip 5)
198/200
                                   1s 936ms/step - loss: 0.0442loading clip data (game 2, clip 3) downscaled
199/200
                                   0s 936ms/step - loss: 0.0442loading clip labels (game 2, clip 3)
200/200
                                   0s 936ms/step - loss: 0.0442loading clip data (game 3, clip 7) downscaled
loading clip labels (game 3, clip 7)
                                   200s 998ms/step - loss: 0.0442 - val_loss: 0.0407
200/200
Epoch 12/15
200/200
                                   0s 936ms/step - loss: 0.0359loading clip data (game 4, clip 7) downscaled
loading clip labels (game 4, clip 7)
loading clip data (game 5, clip 6) downscaled
loading clip labels (game 5, clip 6)
200/200
                                   200s 1s/step - loss: 0.0359 - val_loss: 0.0335
Epoch 13/15
200/200
                                   200s 1s/step - loss: 0.0296 - val_loss: 0.0262
Epoch 14/15
                                   47s 939ms/step - loss: 0.0252loading clip data (game 1, clip 12) downscaled
149/200
loading clip labels (game 1, clip 12)
                                   46s 939ms/step - loss: 0.0252loading clip data (game 5, clip 3) downscaled
150/200
loading clip labels (game 5, clip 3)
loading clip data (game 1, clip 1) downscaled
151/200
                                   45s 939ms/step - loss: 0.0252loading clip labels (game 1, clip 1)
loading clip data (game 1, clip 5) downscaled
loading clip labels (game 1, clip 5)
                                   0s 938ms/step - loss: 0.0249loading clip data (game 4, clip 5) downscaled
200/200
loading clip labels (game 4, clip 5)
loading clip data (game 5, clip 8) downscaled
loading clip labels (game 5, clip 8)
200/200
                                   200s 1s/step - loss: 0.0249 - val_loss: 0.0229
Epoch 15/15
200/200
                                   0s 934ms/step - loss: 0.0213Epoch 15: Saving model...
200/200
                                   199s 997ms/step - loss: 0.0213 - val_loss: 0.0199
Saving to folder /working
Пример пайплайна для тестирования обученной модели:
In [22]:
output_path = prepare_experiment(Path('/kaggle/working'))
new_model = SuperTrackingModel(batch_s, stack_s, out_path=output_path, downscale=downscale)
new_model.load('unet_last_model')
sibatracc final = new model.test(Path('../input/tennistrackingassignment/test/'), [1], do visualization=False, test name='test')
print(f'SiBaTrAcc final value: {sibatracc final}')
Loading model weights for unet last model
Loading model done.
loading clip data (game 1, clip 1) downscaled
loading clip labels (game 1, clip 1)
doing predictions
1/1

    1s 974ms/step

1/1
                              0s 29ms/step
                              0s 28ms/step
1/1
1/1
                              0s 29ms/step
                              0s 28ms/step
1/1
1/1

    0s 28ms/step

1/1
                              0s 29ms/step
1/1
                              0s 39ms/step
1/1
                              0s 28ms/step
                              0s 29ms/step
1/1
1/1
                              0s 28ms/step
1/1
                              0s 22ms/step
1/1

    0s 22ms/step

1/1
                              0s 26ms/step
1/1
                              0s 36ms/step
1/1
                              0s 22ms/step
1/1
                              0s 29ms/step
1/1
                              0s 22ms/step
1/1

    0s 22ms/step

1/1

    0s 23ms/step

1/1
                              0s 23ms/step
                              0s 30ms/step
1/1
1/1
                              0s 23ms/step
                              0s 22ms/step
1/1
1/1
                              0s 22ms/step
1/1
                              0s 23ms/step
                              0s 22ms/step
1/1
                              0s 29ms/step
1/1
                              0s 29ms/step
1/1
1/1
                              0s 22ms/step
                              0s 22ms/step
1/1
                              0s 28ms/step
1/1
1/1
                              0s 28ms/step
```

1/1

1/1 1/1 0s 28ms/step0s 29ms/step

0s 23ms/step

0s 22ms/step

1/1 ————	0s 29ms/step
1/1	0s 28ms/step
1/1	- 0e 22me/etan
1/1	_ 0e 22me/etan
1/1	- 0e 20me/etan
1/1	03 23/113/3(ep
	- 0s 24ms/step
1/1	- us 28ms/step
1/1	- 0s 22ms/step
1/1	0s 22ms/step
1/1	0s 30ms/step
1/1	- 0s 23ms/step
1/1	- 0s 22ms/step
1/1 —	- 0s 23ms/sten
1/1	03 20m3/3tep
	05 22ms/step
1/1	- 0s 22ms/step
	- us 23ms/step
1/1 ————	- 0s 22ms/step
1/1	0s 22ms/step
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predictions are made SiBaTrAcc final value: 0.78872	00040000447
טום וואטט וווומו value: 0.788/2	.7043300311/

SiBaTrAcc final value: 0.79 - можно и ещё пообучать, тем более что лосс падает

снизу прописываю модель заново, чтобы переопределить метод трейн - добавляю выгрузку весов из директории воркинг, продолжаю обучение, ниже см модель для выгрузки из гугл драйва

```
class SuperTrackingModel:
  def init (self, batch s, stack s, out path, downscale, height=720, width=1280):
     self.height = height
     self.width = width
     if downscale:
       self.height //= 2
       self.width //= 2
     self.batch_s = batch_s
     self.stack_s = stack_s
     self.model = UNetForTrack((self.height, self.width, 3 * self.stack s))
     self.out path = out path
     self.downscale = downscale
  def save(self, name: str):
     print("Saving to folder /working")
     self.model.save weights(f'/kaggle/working/{name}.weights.h5')
  def load(self, name: str):
     print(f"Loading model weights for {name}")
     self.model.load weights(f'/kaggle/working/{name}.weights.h5')
     print('Loading model done.')
  def predict_on_batch(self, batch: np.ndarray) -> np.ndarray:
     predictions = self.model.predict(batch).reshape((batch.shape[0], batch.shape[1], batch.shape[2]))
     return predictions
  def _predict_prob_on_clip(self, clip: np.ndarray) -> np.ndarray:
     print('doing predictions')
     n_frames = clip.shape[0]
     # --- get stacks ---
     stacks = []
     for i in range(n_frames - self.stack_s + 1):
       stack = clip[i : i + self.stack_s, ...]
       stack = np.squeeze(np.split(stack, self.stack s, axis=0))
       stack = np.concatenate(stack, axis=-1)
       stacks.append(stack)
     # --- round to batch size ---
     add stacks = 0
     while len(stacks) % self.batch s != 0:
       stacks.append(stacks[-1])
       add_stacks += 1
     # --- group into batches ---
     batches = []
     for i in range(len(stacks) // self.batch s):
       batch = np.stack(stacks[i * self.batch_s : (i + 1) * self.batch_s])
       batches.append(batch)
     stacks.clear()
     # --- perform predictions ---
     predictions = []
     for batch in batches:
       pred = np.squeeze(self.predict_on_batch(batch))
       predictions.append(pred)
     # --- crop back to source length ---
     predictions = np.concatenate(predictions, axis=0)
     if (add_stacks > 0):
       predictions = predictions[:-add_stacks, ...]
     batches.clear()
```

```
# --- add (stack_s - 1) null frames at the begining ---
     start_frames = np.zeros((stack_s - 1, predictions.shape[1], predictions.shape[2]), dtype=np.float32)
     predictions = np.concatenate((start_frames, predictions), axis=0)
     print('predictions are made')
     return predictions
  def get labels from prediction(self, pred prob: np.ndarray, upscale coords: bool) -> np.ndarray:
     n frames = pred prob.shape[0]
     coords = np.zeros([n frames, 3])
     for i in range(n frames):
       prediction mask = pred prob[i]
       if prediction mask.sum() < 1:
          coords[i] = [0, -1, -1]
          prediction mask[prediction mask < 0.5] = 0
          prediction mask[prediction mask >= 0.5] = 1
          code = 0
          x, y = -1, -1
          labeled, num_features = label(prediction_mask)
          if num_features > 0:
            regions = regionprops(labeled)
            largest region = max(regions, key=lambda r: r.area)
            y, x = largest region centroid
            code = 1
          if upscale coords:
            x, y = 2 * x, 2 * y
          coords[i] = [code, x, y]
     return coords
  def predict(self, clip: np.ndarray, upscale coords=True) -> tuple[np.ndarray, np.ndarray]:
     prob pr = self. predict prob on clip(clip)
     labels pr = self.get labels from prediction(prob pr, upscale coords)
     return labels pr, prob pr
  def test(self, data path: Path, games: list, do visualization=False, test name='test'):
     game clip pairs = get game clip pairs(data path, games)
     SIBATRACC vals = []
     for game, clip in game clip pairs:
       data = load clip data(data path, game, clip, downscale=self.downscale)
       if do visualization:
          data_full = load_clip_data(data_path, game, clip, downscale=False) if self.downscale else data
       labels gt = load clip labels(data path, game, clip, downscale=False)
       labels pr, prob pr = self.predict(data, upscale coords=True)
       SIBATRACC_per_frame, SIBATRACC_total = Metrics.evaluate_predictions(labels_gt, labels_pr)
       SIBATRACC_vals.append(SIBATRACC_total)
       if do visualization:
          visualize_prediction(data_full, labels_pr, self.out_path, f'{test_name}_g{game}_c{clip}', SIBATRACC_per_frame)
          visualize_prob(data, prob_pr, self.out_path, f'{test_name}_g{game}_c{clip}')
          del data_full
       del data, labels gt, labels pr, prob pr
       ac.collect()
     SIBATRACC final = sum(SIBATRACC vals) / len(SIBATRACC vals)
     return SIBATRACC final
  def train(self, train generator, val gen, epochs=15, save interval=5, load weights=False):
     self.epochs = epochs
     save path = '/kaggle/working/unet model'
     if load weights:
       self.load('unet_model_epoch_15')
     save callback = SaveEveryNEpochs(save interval, save path)
     self.model.compile(optimizer=tf.keras.optimizers.Adam(learning rate=1e-4),
                 loss=tf.keras.losses.BinaryCrossentropy())
     self.model.fit(train_generator, validation_data=val_gen, epochs=self.epochs,
              steps per epoch=1000 // self.batch s, validation steps=200 // self.batch s,
              callbacks=[save callback])
     self.save('unet last model')
     print('Training done')
In [28]:
```

```
model = SuperTrackingModel(batch_s, stack_s, out_path=output_path, downscale=downscale)
model.train(train_gen.random_g(batch_s), val_gen.random_g(batch_s), epochs=5, save_interval=5, load_weights=True)
Loading model weights for unet model epoch 15
Loading model done.
Epoch 1/5
127/200
                                  1:09 950ms/step - loss: 0.0180loading clip data (game 5, clip 7) downscaled
128/200
                                 1:08 950ms/step - loss: 0.0180loading clip labels (game 5, clip 7)
loading clip data (game 1, clip 3) downscaled
loading clip labels (game 1, clip 3)
loading clip data (game 2, clip 9) downscaled
loading clip labels (game 2, clip 9)
loading clip data (game 5, clip 4) downscaled
loading clip labels (game 5, clip 4)
200/200
                                  0s 974ms/step - loss: 0.0173loading clip data (game 5, clip 6) downscaled
loading clip labels (game 5, clip 6)
loading clip data (game 4, clip 4) downscaled
loading clip labels (game 4, clip 4)
200/200
                                  225s 1s/step - loss: 0.0172 - val_loss: 0.0141
Epoch 2/5
200/200
                                 199s 997ms/step - loss: 0.0116 - val loss: 0.0111
Epoch 3/5
200/200
                                  199s 997ms/step - loss: 0.0092 - val_loss: 0.0101
Epoch 4/5
 7/200
                                 3:00 936ms/step - loss: 0.0083loading clip data (game 2, clip 7) downscaled
 8/200
                                2:59 936ms/step - loss: 0.0083loading clip labels (game 2, clip 7)
 9/200
                                2:58 935ms/step - loss: 0.0082loading clip data (game 1, clip 9) downscaled
loading clip labels (game 1, clip 9)
loading clip data (game 2, clip 9) downscaled
loading clip labels (game 2, clip 9)
loading clip data (game 2, clip 2) downscaled
loading clip labels (game 2, clip 2)
200/200
                                  0s 942ms/step - loss: 0.0076loading clip data (game 4, clip 14) downscaled
loading clip labels (game 4, clip 14)
loading clip data (game 5, clip 2) downscaled
loading clip labels (game 5, clip 2)
200/200
                                  202s 1s/step - loss: 0.0076 - val loss: 0.0094
Epoch 5/5
200/200
                                  0s 933ms/step - loss: 0.0066Epoch 5: Saving model...
200/200
                                  199s 997ms/step - loss: 0.0066 - val loss: 0.0071
Saving to folder /working
Training done
ИТОГО НА 20 ЭПОХЕ МЕТРИКА:
In [29]:
output path = prepare experiment(Path('/kaggle/working'))
new model = SuperTrackingModel(batch s, stack s, out path=output path, downscale=downscale)
new model.load('unet_model_epoch_5') #загружаю веса после 20 эпохи,
                          #просто файл под таким же именем пересохранился
sibatracc_final = new_model.test(Path('../input/tennistrackingassignment/test/'), [1], do_visualization=False, test_name='test')
print(f'SiBaTrAcc final value: {sibatracc final}')
Loading model weights for unet_model_epoch_5
Loading model done.
loading clip data (game 1, clip 1) downscaled
loading clip labels (game 1, clip 1)
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    0s 22ms/step

predictions are made
SiBaTrAcc final value: 0.833510282282972
Во время самостоятельного тестирования попробуйте хотя бы раз сделать тестирование с визуализацией
(do_visualization=True), чтобы визуально оценить качество трекинга разработанной моделью.
output path = prepare experiment(Path('/kaggle/working'))
new model = SuperTrackingModel(batch s, stack s, out path=output path, downscale=downscale)
new_model.load('unet_model_epoch_5') #загружаю веса после 20 эпохи,
                         #просто файл под таким же именем пересохранился
sibatracc_final = new_model.test(Path('../input/tennistrackingassignment/test/'), [1], do_visualization=True, test_name='test')
print(f'SiBaTrAcc final value: {sibatracc_final}')
Loading model weights for unet_model_epoch_5
Loading model done.
loading clip data (game 1, clip 1) downscaled
loading clip data (game 1, clip 1)
loading clip labels (game 1, clip 1)
doing predictions
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                           - 0s 24ms/step
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predictions are made perfoming clip visualization

SiBaTrAcc final value: 0.833510282282972

Загрузка модели через функцию load должна происходить полностью автоматически без каких-либо действий со стороны пользователя! Один из вариантов подобной реализации с использованием google drive и пакета gdown приведен в разделе с дополнениями.

Модель для выгрузки весов из гугл-драйва:

```
def __init__(self, batch_s, stack_s, out_path, downscale, height=720, width=1280):
  self.height = height
  self.width = width
  if downscale:
     self.height //= 2
     self.width //= 2
  self.batch s = batch s
  self.stack_s = stack_s
  self.model = UNetForTrack((self.height, self.width, 3 * self.stack s))
  self.out path = out path
  self.downscale = downscale
def save(self, name: str):
  print("Saving to folder /working")
  self.model.save weights(f'/kaggle/working/{name}.weights.h5')
def load(self, name: str):
  models = {
     'test': '1yn p8P8UDzKHUEQIExS3bMPAY5OeCTd '
  }
  output = f'/kaggle/working/{name}.weights.h5'
  gdown.download(f"https://drive.google.com/file/d/{models[name]}/view?usp=drive link", output, quiet=False, fuzzy=True)
  self.model.load weights(output)
  print('Loading model done.')
def predict_on_batch(self, batch: np.ndarray) -> np.ndarray:
  predictions = self.model.predict(batch).reshape((batch.shape[0], batch.shape[1], batch.shape[2]))
  return predictions
def predict prob on clip(self, clip: np.ndarray) -> np.ndarray:
  print('doing predictions')
  n_frames = clip.shape[0]
  # --- get stacks ---
  stacks = []
  for i in range(n frames - self.stack s + 1):
    stack = clip[i:i+self.stack s, ...]
    stack = np.squeeze(np.split(stack, self.stack s, axis=0))
    stack = np.concatenate(stack, axis=-1)
    stacks.append(stack)
  # --- round to batch size ---
  add stacks = 0
  while len(stacks) % self.batch_s != 0:
    stacks.append(stacks[-1])
    add_stacks += 1
  # --- group into batches ---
  batches = []
  for i in range(len(stacks) // self.batch_s):
    batch = np.stack(stacks[i * self.batch_s : (i + 1) * self.batch_s])
    batches.append(batch)
  stacks.clear()
  # --- perform predictions ---
  predictions = []
  for batch in batches:
    pred = np.squeeze(self.predict on batch(batch))
    predictions.append(pred)
  # --- crop back to source length ---
  predictions = np.concatenate(predictions, axis=0)
  if (add stacks > 0):
    predictions = predictions[:-add stacks, ...]
  batches.clear()
  # --- add (stack_s - 1) null frames at the begining ---
  start frames = np.zeros((stack s - 1, predictions.shape[1], predictions.shape[2]), dtype=np.float32)
  predictions = np.concatenate((start_frames, predictions), axis=0)
  print('predictions are made')
  return predictions
def get labels from prediction(self, pred prob: np.ndarray, upscale coords: bool) -> np.ndarray:
  n frames = pred prob.shape[0]
  coords = np.zeros([n frames, 3])
  for i in range(n frames):
     prediction_mask = pred_prob[i]
    if prediction mask.sum() < 1:</pre>
```

class SuperTrackingModel:

```
coords[i] = [0, -1, -1]
       else:
          prediction_mask[prediction_mask < 0.5] = 0
          prediction_mask[prediction_mask >= 0.5] = 1
          code = 0
          x, y = -1, -1
          labeled, num features = label(prediction mask)
          if num features > 0:
            regions = regionprops(labeled)
            largest_region = max(regions, key=lambda r: r.area)
            y, x = largest_region.centroid
            code = 1
          if upscale_coords:
            x, y = 2 * x, 2 * y
          coords[i] = [code, x, y]
     return coords
  def predict(self, clip: np.ndarray, upscale_coords=True) -> tuple[np.ndarray, np.ndarray]:
     prob_pr = self._predict_prob_on_clip(clip)
     labels_pr = self.get_labels_from_prediction(prob_pr, upscale_coords)
     return labels_pr, prob_pr
  def test(self, data_path: Path, games: list, do_visualization=False, test_name='test'):
     game_clip_pairs = get_game_clip_pairs(data_path, games)
     SIBATRACC_vals = []
     for game, clip in game clip pairs:
       data = load clip data(data path, game, clip, downscale=self.downscale)
       if do visualization:
          data_full = load_clip_data(data_path, game, clip, downscale=False) if self.downscale else data
       labels gt = load clip labels(data path, game, clip, downscale=False)
       labels pr, prob pr = self.predict(data, upscale coords=True)
       SIBATRACC per frame, SIBATRACC total = Metrics evaluate predictions(labels gt, labels pr)
       SIBATRACC\_vals.append(SIBATRACC\_total)
       if do visualization:
          visualize prediction(data full, labels pr, self-out path, f'{test name} g{game} c{clip}', SIBATRACC per frame)
          visualize prob(data, prob pr, self.out path, f'{test name} g{game} c{clip}')
          del data full
       del data, labels_gt, labels_pr, prob_pr
       gc.collect()
     SIBATRACC_final = sum(SIBATRACC_vals) / len(SIBATRACC_vals)
     return SIBATRACC final
  def train(self, train_generator, val_gen, epochs=15, save_interval=5, load_weights=False):
     self.epochs = epochs
     save_path = '/kaggle/working/unet_model'
     if load_weights:
       self.load('unet_model_epoch_15')
     save_callback = SaveEveryNEpochs(save_interval, save_path)
     self.model.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=1e-4),
                 loss=tf.keras.losses.BinaryCrossentropy())
     self.model.fit(train_generator, validation_data=val_gen, epochs=self.epochs,
              steps per epoch=1000 // self.batch s, validation steps=200 // self.batch s,
              callbacks=[save callback])
     self.save('unet last model')
     print('Training done')
In [25]:
import gdown
output path = prepare experiment(Path('/kaggle/working'))
new_model = SuperTrackingModel(batch_s, stack_s, out_path=output_path, downscale=downscale)
new model.load('test')
sibatracc_final = new_model.test(Path('../input/tennistrackingassignment/test/'), [1], do_visualization=False, test_name='test')
print(f'SiBaTrAcc final value: {sibatracc_final}')
Downloading...
From: https://drive.google.com/uc?id=1yn_p8P8UDzKHUEQIExS3bMPAY5OeCTd_
To: /kaggle/working/test.weights.h5
                      93.7M/93.7M [00:00<00:00, 124MB/s]
100%
Loading model done.
loading clip data (game 1, clip 1) downscaled
loading clip labels (game 1, clip 1)
doing predictions
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predictions are made	
SiBaTrAcc final value: 0.8335	10080080070

SiBaTrAcc final value: 0.833510282282972

Дополнения

Иногда при записи большого количества файлов в output директорию kaggle может "тупить" и не отображать корректно структуру дерева файлов в output и не показывать кнопки для скачивания выбранного файла. В этом случае удобно будет запаковать директорию с экспериментом и выкачать ее вручную. Пример для выкачивания директории с первым экспериментом приведен ниже:

In []:

%cd /kaggle/working/

!zip -r "exp_1.zip" "exp_1"

from IPython.display import FileLink

FileLink(r'exp_1.zip')

удалить лишние директории или файлы в output тоже легко:

In []:

!rm -r /kaggle/working/exp 1

Для реализации загрузки данных рекомендуется использовать облачное хранилище google drive и пакет gdown для скачивания файлов. Пример подобного использования приведен ниже:

- 1. загружаем файл в google drive (в данном случае, это npz архив, содержащий один numpy массив по ключу 'w')
- 2. в интерфейсе google drive открываем доступ на чтение к файлу по ссылке и извлекаем из ссылки id файла
- 3. формируем url для скачивания файла
- 4. с помощью gdown скачиваем файл
- 5. распаковываем прz архив и пользуемся питру массивом

Обратите внимание, что для корректной работы нужно правильно определить ід файла. В частности, в ссылке https://drive.google.com/file/d/1kZ8CC-zfkB_TlwtBjuPcEfsPV0Jz7lPA/view?usp=sharing id файла заключен между ...d/ b /view?... и равен 1kZ8CC-zfkB_TlwtBjuPcEfsPV0Jz7lPA

In []:

import gdown

id = '1kZ8CC-zfkB TlwtBjuPcEfsPV0Jz7IPA' url = f'https://drive.google.com/uc?id={id}' output = 'sample-weights.npz' gdown.download(url, output, quiet=False)

import numpy as np

weights = np.load('/kaggle/working/sample-weights.npz')['w'] print(weights)