## 1. section1 強化学習

マルコフ決定過程(Markov decision process: MDP)

```
class MDP:
         _init__(self, init, actlist, terminals, gamma=.9):
        self.init = init
        self.actlist = actlist
self.terminals = terminals
        if not (0 <= gamma < 1):
            raise ValueError("An MDP must have 0 <= gamma < 1")
        self.gamma = gamma
        self.states = set()
        self.reward = {}
    def R(self, state):
    return self.reward[state]
    def T(self, state, action):
        raise NotImplementedError
    def actions(self, state):
        if state in self.terminals:
            return [None]
        else:
            return self.actlist
```

```
class GridMDP(MDP):
           init__(self, grid, terminals, init=(0, 0), gamma=.9):
         grid.reverse() # because we want row 0 on bottom, not on top
         MDP.__init__(self, init, actlist=orientations,
                        terminals=terminals, gamma=gamma)
         self.grid = grid
         self.rows = len(grid)
self.cols = len(grid[0])
         for x in range(self.cols):
              for y in range(self.rows):
                  self.reward[x, y] = grid[y][x]
                  if grid[y][x] is not None:
                       self.states.add((x, y))
    def T(self, state, action):
         if action is None:
              return [(0.0, state)]
             def go(self, state, direction):
         state1 = vector_add(state, direction)
return state1 if state1 in self.states else state
    def to_grid(self, mapping):
         return list(reversed([[mapping.get((x, y), None)
                                    for x in range(self.cols)]
                                   for y in range(self.rows)]))
    def to_arrows(self, policy):
    chars = {(1, 0): '>', (0, 1): '^', (-1, 0): '<', (0, -1): 'v', None:
    return self.to_grid({s: chars[a] for (s, a) in policy.items()})</pre>
```

## 2. section2\_AlphaGo

### AlphaGo Zero モンテカルロ木探索の実装

```
from _future__ import annotations
from abc import ABC, abstractmethod
from typing import List

class IState(ABC):
    @abstractmethod
    def legal_actions(self) -> List[int]:
    pass

@abstractmethod
def random_action(self) -> int:
    pass

@abstractmethod
def next(self, action: int) -> IState:
    pass

@abstractmethod
def is_lose(self) -> bool:
    pass

@abstractmethod
def is_draw(self) -> bool:
    pass

@abstractmethod
def is_draw(self) -> bool:
    pass

@abstractmethod
def is_draw(self) -> bool:
    pass

@abstractmethod
def is_done(self) -> bool:
    pass

@abstractmethod
def is_done(self) -> bool:
    pass

@abstractmethod
def is_done(self) -> bool:
    pass

@abstractmethod
def is_first_player(self) -> bool:
    pass
```

```
| class Node:
| def __init__(self, state: IState, expand_base: int = 10) -> None:
| self.state: IState = state | self.w: int = 0 # 部間回数 | self.m: int = 0 # 部間回数 | self.expand_base: int = expand_base | self.children: Optional[List[Node]] = None | | | | | | | | | |
| def evaluate(self) -> float: | """self (current Node) の評価値を計算して更新する.""" |
| if self.state.is_done(): | value = -1 if self.state.is_lose() else 0 | self.w += value | self.m += 1 | return value |
| # self (current Node) (で子ノードがない場合 | if not self.children: | # ラグダムにプレイする | v = Node.playout(self.state) | self.w += v | self.m += 1 | # 十分に self (current Node) がプレイされたら展開(1ノード振り進める)する | if self.m == self.expand_base: | self.expand() | return v | else: | v = -self.next_child_based_ucb().evaluate() | self.m += 1 | return v |
```

```
def expand(self) -> None:
    """self (current Node) を展開する."""
    self.children = [Node(self.state.next(action), self.expand_base) for action in self.state.legal_actions()]

def next_child_based_ucb(self) -> Node:
    """self (current Node) のテノードから1ノード選択する."""

    # 試行回数がのクードを優先的に選ぶ
    for child in self.children:
        if child.n == 0:
            return child

# UCB1

# U
```

## 3. section3 軽量化 高速化技術

#### 蒸留実装

```
# 生徒モデル
with tf.device('/cpu:0'):
    student_model = Model(inputs=input_layer, outputs=output)
     # 入力として学習データの正解ラベルを入れる
    input_true = Input(name='input_true', shape=[im_height, im_width, im_chan])
# 教師モデル + 生徒モデル
# 自作損失関数をレイヤーとして組込み
output_loss = Lambda(knowledge_distillation_loss, output_shape=(1,), name='kd_|)(
    [output, input_true, teacher_probabilities_T, probabilities_T]
# input_layer:入力 input_true:学習データの正解ラベル inputs = [input_layer, input_true] with tf.device('/cpu:0'):
# 損失値を出力とする
    train_model = Model(inputs=inputs, outputs=output_loss)
# 出力が loss になるように設定
train_model.compile(optimizer='adam', loss= lambda y_true, y_pred: y_pred)
#損失関数の作成
from keras.losses import binary_crossentropy as logloss
lambda_ = 0.9
def knowledge_distillation_loss(input_distillation):
    y_pred, y_true, y_soft, y_pred_soft = input_distillation
return (1 - lambda_) * logloss(y_true, y_pred) + lambda_*T*logloss(y_soft, y_pred_soft)
```

# 4. section4\_応用モデル

#### DenseNet 実装

```
from keras.layers import Conv2D, Activation, BatchNormalization, Concatenate, AveragePooling2D, Input, GlobalAveragePooling2D, Dense from keras.models import Model from keras.optimizers import cifar10 from keras.utils import to_categorical from keras.utils import to_categorical from keras.preprocessing.image import ImageDataGenerator import pickle import numpy as np
# DenseBlockのLayer
def dense_block(self, input_tensor, input_channels, nb_blocks):
    x = input_tensor
    n_channels = input_channels
    for i in range(nb_blocks):
    # 分歧前の本線
                                      # 分收制的本級
main = X
# DenseBlock側の分歧
X = BatchNormalization()(X)
X = Activation("relu")(X)
# Bottle-Neck 1X1畳み込み
                                      # Bottle-Neck 1x1畳み込み

X = Conv2D(125, (1, 1))(x)

X = BatchNormalization()(x)

X = Activation("relu")(X)

3 3x2畳み込み フィルター数は成長率

X = Conv2D(3elf.k, (3, 3), padding="same")(X)

# 本級と結合
                        x = Concatenate()([main, x])
n_channels += self.k
return x, n_channels
                     # Transaction Layer(self, input_tensor, input_channels):
    n_channels = int(input_channels * self.compression)
    # 1x1畳み込みで圧縮
                                   x = Conv2D(n_channels, (1, 1))(input_tensor)
                                  x = AveragePooling2D((2, 2))(x)
return x, n_channels
                   # モデルの作成

def make_model(self, blocks):
    # blocks=[6,12,24,16]とするとDenseNet-121の設定に準じる
    input = Input(shape=(32,32,3))
    # 媚教を出さないようにフィルター教 16にする
                                  n = 16
x = Conv2D(n, (1,1))(input)
                                   x, n = self.transition_layer(x, n)
# DenseBlock
x, n = self.dense_block(x, n, blocks[i])
# GlobalAveragePooling(チャンネル単位の全平均)
x = GlobalAveragePooling2D()(X)
                                   output = Dense(10, activation="softmax")(x)
                                  model = Model(input, output)
return model
                    # ਜ਼ਗ਼ਖ਼ਸ਼
def train(self, X_train, y_train, X_val, y_val):
                                            f.model.compile(optimizer=Adam(), loss="categorical_crossentropy", metrics=["acc"])
                              self.model.comple.self.model.comple.self.model.comple.self.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.comple.c
                               # 別報
# 別報
# #history = self.model.fit(X_train, y_train, batch_size=128, epochs=1, validation_data=(X_val, y_val)).history
# 水増しありの訓練
                                # 水地(地方の創練
history = self.model.fit_generator(datagen.flow(X_train, y_train, batch_size=128),
steps_per_epoch=len(X_train) / 128, validation_data=(X_val, y_val), epochs=1).history
# 保存
with open("history.dat", "wb") as fp:
                                            h open("history.dat", "wb") as fp:
pickle.dump(history, fp)
      if __name__ == "__main__":
# k=16の場合
densenet = DenseNetSimple(16)
# densenet.model.summary()
                   # CIFAR-100ថ្នៃការ៉េក
(X_train, y_train), (X_test, y_test) = cifar10.load_data()
| kasta = (X_train / 255.0).astype('float32")
                   X_test = (X_test / 255.0).astype("float32")
y_train, y_test = to_categorical(y_train), to_categorical(y_test)
                   densenet.train(X_train, y_train, X_test, y_test)
```

## 5. section5 Transformer

Transformer, Encoder, Decoder

```
import tensorflow datasets as tfds
      import tensorflow as tf
      import time
      import numpy as np
      import matplotlib.pyplot as plt
      class Encoder(tf.keras.layers.Layer):
        def __init__(self, num_layers, d_model, num_heads, dff, input_vocab_size,
                     maximum_position_encoding, rate=0.1):
          super(Encoder, self).__init__()
          self.d_model = d_model
          self.num_layers = num_layers
          self.embedding = tf.keras.layers.Embedding(input_vocab_size, d_model)
          self.pos_encoding = positional_encoding(maximum_position_encoding,
                                                  self.d model)
          self.enc_layers = [EncoderLayer(d_model, num_heads, dff, rate)
                             for _ in range(num_layers)]
          self.dropout = tf.keras.layers.Dropout(rate)
        def call(self, x, training, mask):
26
27
         def call(self, x, training, mask):
           seq_len = tf.shape(x)[1]
           # 埋め込みと位置エンコーディングを合算する
           x = self.embedding(x) # (batch_size, input_seq_len, d_model)
           x *= tf.math.sqrt(tf.cast(self.d_model, tf.float32))
           x += self.pos_encoding[:, :seq_len, :]
           x = self.dropout(x, training=training)
           for i in range(self.num_layers):
            x = self.enc_layers[i](x, training, mask)
           return x # (batch_size, input_seq_len, d_model)
```

```
class Decoder(tf.keras.layers.Layer):
 super(Decoder, self).__init__()
    self.d_model = d_model
    self.num_layers = num_layers
    self.embedding = tf.keras.layers.Embedding(target_vocab_size, d_model)
    self.pos encoding = positional encoding(maximum position encoding, d model)
    self.dec_layers = [DecoderLayer(d_model, num_heads, dff, rate)
    for _ in range(num_layers)]

self.dropout = tf.keras.layers.Dropout(rate)
  def call(self, x, enc_output, training,
           look_ahead_mask, padding_mask):
    seq_len = tf.shape(x)[1]
    attention weights = {}
    x = self.embedding(x) # (batch_size, target_seq_len, d_model)
x *= tf.math.sqrt(tf.cast(self.d_model, tf.float32))
    x += self.pos_encoding[:, :seq_len, :]
    x = self.dropout(x, training=training)
    for i in range(self.num_layers):
     x, block1, block2 = self.dec_layers[i](x, enc_output, training,
                                                look_ahead_mask, padding_mask)
      attention_weights['decoder_layer{}_block1'.format(i+1)] = block1
attention_weights['decoder_layer{}_block2'.format(i+1)] = block2
    # x.shape == (batch_size, target_seq_len, d_model)
    return x, attention weights
```

```
class Transformer(tf.keras.Model):

def __init__(self, num_layers, d_model, num_heads, dff, input_vocab_size, target_vocab_size, pe_input, pe_target, rate=0.1):

super(Transformer, self).__init__()

self.encoder = Encoder(num_layers, d_model, num_heads, dff, input_vocab_size, pe_input, rate)

self.decoder = Decoder(num_layers, d_model, num_heads, dff, target_vocab_size, pe_target, rate)

self.final_layer = tf.keras.layers.Dense(target_vocab_size)

def call(self, inp, tar, training, enc_padding_mask, look_ahead_mask, dec_padding_mask):

enc_output = self.encoder(inp, training, enc_padding_mask) # (batch_size, inp_seq_len, d_model)

# dec_output.shape == (batch_size, tar_seq_len, d_model)

dec_output, attention_weights = self.decoder(
    tar, enc_output, training, look_ahead_mask, dec_padding_mask)

final_output = self.final_layer(dec_output) # (batch_size, tar_seq_len, target_vocab_size)

return final_output, attention_weights
```

## 6. section6 物体検知 セグメンテーション

セマンテックセグメンテーションの実装

```
import json
       import os
       import glob
4
       import shutil
       # 画像関係
       import numpy as np
       import cv2
       from PIL import Image
       # 画像表示
       import matplotlib.pyplot as plt
       IMAGE SIZE = 256
       # データのリスト
       json_list = glob.glob('seg_dogs/*.json')
img_list = [f.replace('json', 'jpg') for f in json_list]
       print(len(json list))
       no = 1
       # アノテーションデータ読み込み
       with open(json_list[no]) as f:
            data = json.loads(f.read())
       # 1つだけ取り出す
       shape = data['shapes'][0]
       label = shape['label']
       points = shape['points']
       shape_type = shape['shape_type']
print('[label]', label)
print('[shape_type]', shape_type)
print('[points]', points)
```

```
# 画像読み込み
img = cv2.imread(img_list[no])
img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
# アノテーション部分
mask = np.zeros((img.shape[0], img.shape[1]), dtype=np.uint8)
mask = cv2.fillPoly(mask, np.int32([points]), 1)
# 横並びに表示
fig = plt.figure(figsize=(12, 6))
ax1 = fig.add_subplot(1, 2, 1)
ax2 = fig.add_subplot(1, 2, 2)
ax1.imshow(img)
ax2.imshow(mask, cmap='gray')
# フォルダ作成 trainとvalにデータを分けます
train dir = 'train'
val dir = 'val'
if not os.path.exists(train_dir):
    os.mkdir(train_dir)
    os.mkdir(train_dir + '/images')
os.mkdir(train_dir + '/masks')
if not os.path.exists(val_dir):
    os.mkdir(val_dir)
    os.mkdir(val dir + '/images')
    os.mkdir(val_dir + '/masks')
```

```
# 114個のデータを用意したので 100 と 14 に分けます
for ind, file in enumerate(json list):
    points = []
with open(file) as f:
         data = json.loads(f.read())
         for s in data['shapes']:
points.append(s['points'])
    if points:
         # 画像データを読み込み画像サイズ取得
         img_path = file.replace('json', 'jpg')
         img = cv2.imread(img_path)
         # ファイル名
file_name = os.path.basename(img_path)
         # jsonのアノテーションデータ
# 犬:1
         # 背景:0
         mask = np.zeros((img.shape[0], img.shape[1]), dtype=np.uint8)
         for p in points:
              mask = cv2.fillPoly(mask, np.int32([p]), 1)
         img = cv2.resize(img, (IMAGE_SIZE, IMAGE_SIZE), interpolation=cv2.INTER_NEAREST)
mask = cv2.resize(mask, (IMAGE_SIZE, IMAGE_SIZE), interpolation=cv2.INTER_NEAREST)
         #保存
         file_name = file_name.replace('jpg', 'png')
         if ind<100:
              maskim = Image.fromarray(np.uint8(mask))
              maskim.save(f'train/masks/{file_name}')
cv2.imwrite(f'train/images/{file_name}', img)
              maskim = Image.fromarray(np.uint8(mask))
maskim.save(f'val/masks/{file_name}')
cv2.imwrite(f'val/images/{file_name}', img)
```