```
!pip install python-chess

→ Collecting python-chess
      Downloading python_chess-1.999-py3-none-any.whl.metadata (776 bytes)
    Collecting chess<2,>=1 (from python-chess)
      Downloading chess-1.11.1.tar.gz (156 kB)
                                                 - 156.5/156.5 kB 6.4 MB/s eta 0:00:00
      Preparing metadata (setup.py) ... done
    Downloading python_chess-1.999-py3-none-any.whl (1.4 kB)
    Building wheels for collected packages: chess
      Building wheel for chess (setup.py) ... done
      Created wheel for chess: filename=chess-1.11.1-py3-none-any.whl size=148497 sha256=677a2bcc384f4ba1c0dcf91fe91f9
      Stored in directory: /root/.cache/pip/wheels/2e/2d/23/1bfc95db984ed3ecbf6764167dc7526d0ab521cf9a9852544e
    Successfully built chess
    Installing collected packages: chess, python-chess
    Successfully installed chess-1.11.1 python-chess-1.999
import chess
import chess.pgn
from chess import Board, Move
from enum import Enum
from google.colab import drive
import io
import itertools
import os
import pandas as pd
from pandas import DataFrame
import numpy as np
import sys
import torch
import torch.nn as nn
import torch.nn.functional as F
from torch.nn.utils.rnn import pad_sequence
import torch.optim as optim
from torch.utils.data import Dataset, DataLoader, random_split
from typing import Iterator, List
from google.colab import drive
drive.mount('/content/drive')
→ Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_
token = "github pat 11ABBPRPA0YkTnI8AiAWk0 7zZ3Y7ZSLgYkaPJp6TlBipd4oxbZz1ckKxYh0APGplUECIONAVPWK05U924"
rm -rf /content/chess-ml
!git clone https://{token}@github.com/johnsonlarryl/chess-ml.git

→ Cloning into 'chess-ml'...
    remote: Enumerating objects: 158, done.
    remote: Counting objects: 100% (158/158), done.
    remote: Compressing objects: 100% (104/104), done.
    remote: Total 158 (delta 62), reused 134 (delta 44), pack-reused 0 (from 0)
    Receiving objects: 100% (158/158), 101.61 KiB | 732.00 KiB/s, done.
    Resolving deltas: 100% (62/62), done.
os.environ["POSTGRES_HOSTNAME"]=''
os.environ["POSTGRES_DATABASE"]=''
os.environ["POSTGRES_USERNAME"]=''
os.environ["POSTGRES_PASSWORD"]=''
os.environ["POSTGRES_TOTAL_GAMES"]=''
sys.path.append("/content/chess-ml/chess-ml-dao")
from chess_ml_dao.dao.postgres import PGNDAO
from chess_ml_dao.algo import ChessMoveModel
```

```
from chess_ml_dao.model.transformation import PlayerTurn
from chess_ml_dao.util.transformation import get_board_array, \
                                             get move mask, \
                                             get_players_turn_array, \
                                             generate all possible white promotion moves, \
                                             generate all possible black promotion moves, \
                                             generate_black_castling, \
                                             generate_white_castling, \
                                             POSSIBLE_MOVES
class ChessMoveModel(nn.Module):
    def __init__(self, number_of_possible_moves):
        super(ChessMoveModel, self).__init__()
        # Convolutional Layers
        self.conv1 = nn.Conv2d(in_channels=14, out_channels=32, kernel_size=3, stride=1, padding=1)
        self.bn1 = nn.BatchNorm2d(32)
        self.conv2 = nn.Conv2d(in_channels=32, out_channels=64, kernel_size=3, stride=2, padding=1) # Stride reduces
        self.bn2 = nn.BatchNorm2d(64)
        self.conv3 = nn.Conv2d(in channels=64, out channels=128, kernel_size=3, stride=1, padding=1)
       self.bn3 = nn.BatchNorm2d(128)
        self.conv4 = nn.Conv2d(in_channels=128, out_channels=256, kernel_size=3, stride=2, padding=1) # Stride reduce
       self.bn4 = nn.BatchNorm2d(256)
        self.conv5 = nn.Conv2d(in_channels=256, out_channels=512, kernel size=3, stride=1, padding=1)
       self.bn5 = nn.BatchNorm2d(512)
        self.conv6 = nn.Conv2d(in channels=512, out channels=512, kernel size=3, stride=1, padding=1)
        self.bn6 = nn.BatchNorm2d(512)
        # Global Average Pooling
        self.global_pool = nn.AdaptiveAvgPool2d((1, 1)) # Global pooling
        self.fc1 = nn.Linear(512, 512)
        self.bn_fc = nn.BatchNorm1d(512)
       self.fc2 = nn.Linear(512, number_of_possible_moves)
    def forward(self, x):
       x = F.relu(self.bn1(self.conv1(x)))
        x = F.relu(self.bn2(self.conv2(x))) # Stride reduces spatial dimensions
       x = F.relu(self.bn3(self.conv3(x)))
       x = F.relu(self.bn4(self.conv4(x))) # Stride reduces spatial dimensions
       x = F.relu(self.bn5(self.conv5(x)))
       x = F.relu(self.bn6(self.conv6(x)))
       # Global Average Pooling
       x = self.global_pool(x) # Shape: (batch_size, 512, 1, 1)
       x = x.view(x.size(0), -1) # Flatten to (batch_size, 512)
       # Fully Connected Layers
       x = F.relu(self.bn_fc(self.fc1(x)))
       x = self.fc2(x)
        return x
# Load your trained model
number_of_possible_moves=len(POSSIBLE_MOVES)
model = ChessMoveModel(number_of_possible_moves=number_of_possible_moves)
model.load_state_dict(torch.load('/content/drive/MyDrive/UNT/CSCE 5218/Semester Project/models/cnn.12.14.2024.11.46.pyt
model.eval() # Set the model to evaluation mode
   <ipython-input-30-02908f1794d8>:4: FutureWarning: You are using `torch.load` with `weights_only=False` (the curren
      model.load_state_dict(torch.load('/content/drive/MyDrive/UNT/CSCE 5218/Semester Project/models/cnn.12.14.2024.11
    ChessMoveModel(
      (conv1): Conv2d(14, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
```

```
(bn1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv2): Conv2d(32, 64, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1))
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv3): Conv2d(64, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
      (bn3): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv4): Conv2d(128, 256, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1))
      (bn4): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv5): Conv2d(256, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
      (bn5): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv6): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
      (bn6): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (global_pool): AdaptiveAvgPool2d(output_size=(1, 1))
      (fc1): Linear(in_features=512, out_features=512, bias=True)
      (bn_fc): BatchNorm1d(512, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (fc2): Linear(in_features=512, out_features=4548, bias=True)
def prepare_input_tensor(board: chess.Board) -> torch.Tensor:
    board_array = get_board_array(board)
   players_turn = np.full((8, 8), 1 if board.turn == chess.WHITE else −1, dtype=np.float32)
    if board.turn:
     current_player = chess.WHITE
   else:
     current_player = chess.BLACK
   players_turn = get_players_turn_array(board, current_player)
   pieces = PGNDAO.get_piece_channels(board)
   board_features = np.vstack([
        np.expand_dims(board_array, axis=0),
        np.expand_dims(players_turn, axis=0),
        pieces
    ])
    input tensor = torch.tensor(board features, dtype=torch.float32).unsqueeze(0) # Shape: (1, 14, 8, 8)
    return input_tensor
def make_move(board: chess.Board) -> str:
    Generate a move for the computer using legal moves.
    legal_moves = list(board.legal_moves) # Get all legal moves
    if not legal_moves:
        raise ValueError("No legal moves available.")
   # Pick a random legal move (or implement your logic here)
    chosen_move = legal_moves[0] # Example: pick the first legal move
    return chosen_move.uci() # Return the move in UCI format
def filter model outputs(outputs: torch.Tensor, possible moves: list, original moves: list) -> torch.Tensor:
    # Map reduced moves to their indices in the original move set
    indices = [original_moves.index(move) for move in possible_moves]
   # Filter the logits to include only reduced moves
    return outputs[:, indices]
def generate_original_moves() -> list:
   # Standard moves
    standard_moves = generate_standard_possible_moves()
   # Promotions
   white_promotions = generate_all_possible_white_promotion_moves()
   black_promotions = generate_all_possible_black_promotion_moves()
```

```
# Castling
   castling moves = generate white castling() + generate black castling()
    # Combine all moves
    all moves = standard moves + white promotions + black promotions + castling moves
    # Ensure the total matches 4612
   while len(all_moves) < len(POSSIBLE_MOVES):</pre>
        all_moves.append(f"placeholder_move_{len(all_moves)}")
    return all moves
# Define ORIGINAL MOVES
ORIGINAL_MOVES = generate_original_moves()
print(f"Number of ORIGINAL_MOVES: {len(ORIGINAL_MOVES)}") # Should print 4612
→ Number of ORIGINAL_MOVES: 4548
def make_move(board: chess.Board) -> str:
    input_tensor = prepare_input_tensor(board)
    # Get model predictions
    with torch.no grad():
        outputs = model(input_tensor) # Shape: (1, num_original_moves)
    # Filter outputs to match the reduced move set
    outputs = filter_model_outputs(outputs, POSSIBLE_MOVES, ORIGINAL_MOVES)
    # Generate the move mask
    move_mask = get_move_mask(board)
   move_mask_tensor = torch.tensor(move_mask, dtype=torch.float32)
    # Apply the mask to outputs
   masked_outputs = outputs * move_mask_tensor
    # Handle edge case: no valid moves
    if move_mask_tensor.sum() == 0:
        if board.is_checkmate():
            raise ValueError("Checkmate: No valid moves available.")
        elif board.is_stalemate():
            raise ValueError("Stalemate: No valid moves available.")
            raise ValueError("No valid moves available; the game is over or an error occurred.")
    # Get the best move index
    _, best_move_index = masked_outputs.max(dim=1)
    # Map the index to a UCI move
   predicted_move = POSSIBLE_MOVES[best_move_index.item()]
    # Validate the predicted move
    legal moves = {move.uci() for move in board.legal moves}
    if predicted move not in legal moves:
        print(f"Invalid move generated: {predicted_move}. Falling back to best legal move.")
        # Find the best legal move
        legal_moves_indices = [POSSIBLE_MOVES.index(move) for move in legal_moves]
        legal_logits = outputs[0, legal_moves_indices] # Logits for legal moves only
        best legal index = legal moves indices[legal logits.argmax().item()]
        predicted_move = POSSIBLE_MOVES[best_legal_index]
    return predicted_move
def human_move(board: Board, move: str) -> Board:
 move = chess.Move.from_uci(move)
 board.push(move)
  return board
```

```
def computer_move(board: chess.Board) -> chess.Board:
   move = make_move(board)
   print(f"Computer: {move}")
       board.push(chess.Move.from_uci(move))
   except chess.InvalidMoveError:
       print(f"Invalid move generated: {move}")
   return board
Start coding or generate with AI.
def get_move_mask(board: chess.Board) -> np.ndarray:
   # Initialize the move mask
   move_mask = np.zeros(len(POSSIBLE_MOVES), dtype=np.float32)
   # Get all legal moves in UCI format (filtered by python-chess to handle checks)
   legal_moves = {move.uci() for move in board.legal_moves}
   # Iterate over all possible moves and mark legal ones
   for i, move in enumerate(POSSIBLE_MOVES):
       if move in legal_moves:
           move_mask[i] = 1
   return move_mask
```

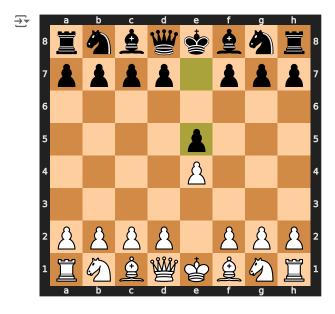
# Create a new chess board (starting position) board = chess.Board() board







human\_move(board, "e7e5")

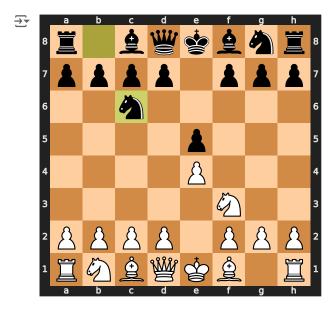


computer\_move(board)





human\_move(board, "b8c6")

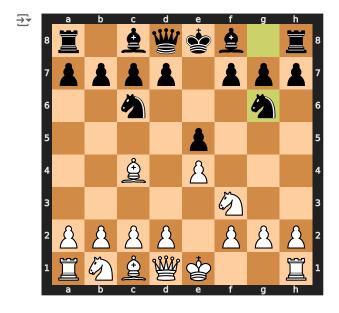


computer\_move(board)





human\_move(board, "g8g6")

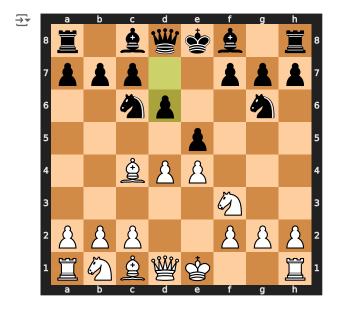


computer\_move(board)

→ Computer: d2d4



human\_move(board, "d7d6")



Invalid move generated: a1a2. Falling back to best legal move. Computer: e1g1



human\_move(board, "f8e7")



Invalid move generated: a1a2. Falling back to best legal move. Computer: d4d5



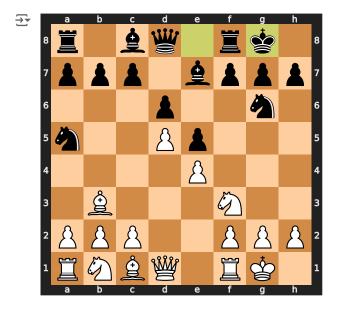
human\_move(board, "c6a5")







human\_move(board, "e8g8")

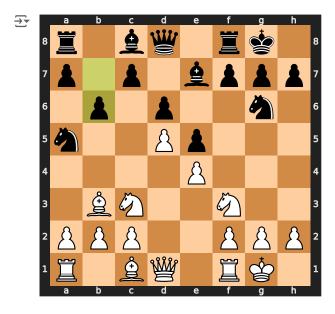


computer\_move(board)





human\_move(board, "b7b6")

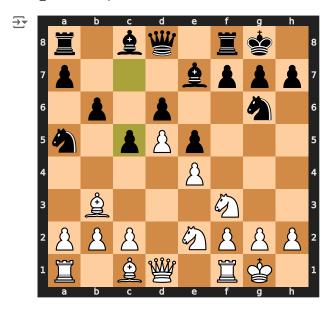


computer\_move(board)

Invalid move generated: a1a2. Falling back to best legal move. Computer: c3e2



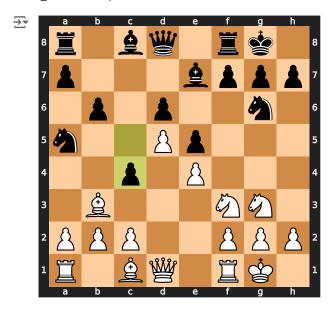
human\_move(board, "c7c5")



Invalid move generated: a1a2. Falling back to best legal move. Computer: e2g3



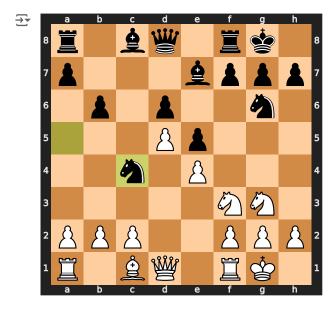
human\_move(board, "c5c4")







human\_move(board, "a5c4")



computer\_move(board)

Invalid move generated: a1a2. Falling back to best legal move. Computer: g3f5



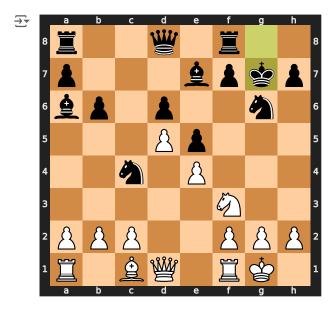
human\_move(board, "c8a6")



Invalid move generated: a1a2. Falling back to best legal move. Computer: f5g7



human\_move(board, "g8g7")





human\_move(board, "b6b5")



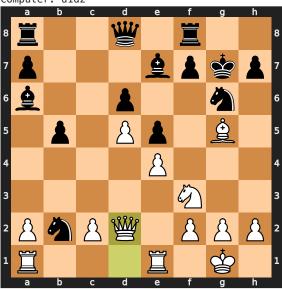
Invalid move generated: a1a2. Falling back to best legal move. Computer: c1g5



human\_move(board, "c4b2")



Invalid move generated: a1a2. Falling back to best legal move. Computer: d1d2



human\_move(board, "a8b8")



Invalid move generated: a1a2. Falling back to best legal move. Computer: g5e3

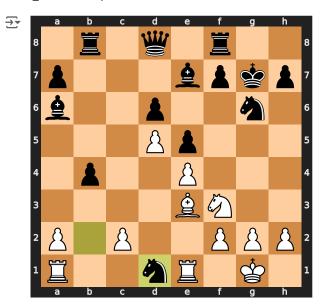


human\_move(board, "b5b4")





human\_move(board, "b2d1")



computer\_move(board)



human\_move(board, "d8a5")

