Human value detection

Ferraro Francesca

11/12/2023

Step 1 - Dataset Loading

Arguments from dataset are loaded as the union of the training, validation and test data from the original dataset. Each one of the tab-separated columns (Premise, Stance and Conclusion) is considered for the goal of the project.

```
def load dataset():
    # Load the dataset into separate DataFrames for each split
    df training = pd.read csv('arguments-training.tsv', delimiter='\t')
    df validation = pd.read csv('arguments-validation.tsv', delimiter='\t')
    df test = pd.read csv('arguments-test.tsv', delimiter='\t')
    # Concatenate all the dataframes
    df = pd.concat([df training, df validation, df test])
    # Extract the argument text from each DataFrame
    arguments = df['Premise'].tolist()
    stances = df['Stance'].tolist()
    conclusions = df['Conclusion'].tolist()
    return arguments, stances, conclusions
```

Step 2 - Preprocessing and model initialization

Preprocessing is strongly linked to the choose of the model, in this case, with bert-base-uncased, preprocessing is done (on the premise and conclusion) by the words conversion to lowercase and the punctuation removal. Then, the model is loaded.

```
def process_sentences(sentences):
    processed = []
    for sentence in sentences:
        # Remove punctuation
        sentence = sentence.translate(str.maketrans('', '', string.punctuation))
        # Convert to Lowercase
        sentence = sentence.lower()
        processed.append(sentence)
    return processed
```

```
# model inizialization
model_bert = pipeline('fill-mask', model='bert-base-uncased') # Bert
```

Step 3 - Argument preparation

Each premise is mapped to the corresponding stance and conclusion to be easily used in the prompt construction.

Then a random argument is chosen from the dataset as an example.

```
# Map each argument to the corresponding stance and conclusion
if len(arguments) != len(stances) or len(arguments) != len(conclusions):
    print("Error: incompatible data")
else:
    input = {}
    for i in range(len(arguments)):
        input[arguments[i]] = (stances[i], conclusions[i])
```

```
import random

# Random argument selection
input_argument = random.choice(list(input.keys()))

argument = input_argument
stance = input[input_argument][0]
conclusion = input[input_argument][1]
```

Step 4 - Prompt and word generation

Prompt is constructed using argument's premise, stance and conclusion. Several kinds of prompt were tried and the best results to be the one with a first-person sentence, but others are worth exploring.

The model predicts a series of masked words for describing the person associated to the argument and the one with the highest score is chosen.

```
def generate_word(model) :
    # Create prompt
    prompt = f"i am {stance} the fact that {conclusion} because i think that {argument}. i am a {model.tokenizer.mask_token}."
    # Generate filling
    output = model(prompt)

# Sort output by 'score'
    output.sort(key=lambda x: x['score'], reverse=True)

# Extract 'sequence' with higher 'score'
    description = output[0]['sequence']

return description

description_bert = generate_word(model_bert)
    print(description_bert)
```

i am against the fact that the european union should be harder towards irreparable migrants those who do not value their lives in peace security and prosperity because i think that migrants have an enormous amount to offer in energy new ideas and cultures why this emphasis on imaginary takeovers. i am a socialist.

Results

Here are some examples of the model's predictions. This is not a representative sample, but rather a selection of examples. Some of the predictions appear to be appropriate, while others don't.

PROMPT	PREDICTION	EVALUATION
am in favor of the fact that payday loans should be banned because I think that they charge a crazy amount of interest on their loans driving many people into serious debt.	Libertarian	Appropriate
am in favor of the fact that we should enforce the rule of law, promoting respect for human rights without linking them to any alleged "European" identity.	Conservative	Appropriate
am against the fact that the EU should immediately end the refugee deal with Turkey.	Conservative	Appropriate
am in favor of the fact that we should ban naturopathy because I think that naturopathy s just a scam that has no scientific backing people that could truly be helped with medicine that choose naturopathy will suffer unde	Vegetarian	Not Appropriate
am against the fact that we should cancel pride parades because I think that cancelling oride parades would make people less aware of the LGBT movement.	Feminist	Appropriate
am in favor of the fact that we should limit executive compensation because I think that he executives already earn enough money to give more incentives those incentives have to go to teps below where they are necessary.	Conservative	Not Appropriate
am in favor of the fact that holocaust denial should be a criminal offence because I think that denial of this is wrong it happened and many people died.	Conservative	Appropriate
am against the fact that we should end racial profiling because I think that we shouldnt end racial profiling because it is a way to stop criminals before they commit crimes.	Feminist	Not Appropriate
am in favor of the fact that we need a common migration and asylum policy, based on respect for rights and equal treatment because! think that in the EU labor market we will need more workers from outside Europe to continue the level of economic activities we are planning.	Socialist	Appropriate
am in favor of the fact that we should ban algorithmic trading because I think that algorithmic rading has its disadvantages which include system failure risksnetwork connectivity errors timelags between trade orders and execution and most important of all imperfect algorithms.	Libertarian	Appropriate
am in favor of the fact that we should ban fast food because I think that fast food is bad for adults and children and causes obesity and the risk of diabetes in many cases.	Vegetarian	Appropriate
am in favor of the fact that we should ban whaling because I think that the whales are the ambassadors of the sea and an icon of the struggle to protect the planet they must prohibit whaling.	Vegetarian	Appropriate
am against the fact that we should subsidize stay - at - home dads because I think that the uggestion that men deserve payment for what has traditionally been womens unpaid work is problematic it sends the message that only mens labor has value.	Feminist	Appropriate
am against the fact that we should abandon marriage because I think that society needs people who commit to marriage rather than just living together as an example to children that a lifetime commitment really does mean that.	Feminist	Not Appropriate
am in favor of the fact that we should subsidize journalism because I think that subsidizing ournalism increases journalist pay allowing for better journalism.	Conservative	Not Appropriate

I am in favor of the fact that we should ban missionary work because I think that missionary work often only benefits the missionary those that are targeted are often left worse off after the missionary leaves having been given a taste of first world things then left to rot.	Christian	Not Appropriate
I am against the fact that we should ban naturopathy because I think that naturapathy tends to have less side effects and can lead to a healthier lifestyle.	Vegetarian	Appropriate
I am against the fact that we should adopt an austerity regime because I think that juggling debt in never a good option compared to paying it off properly.	Conservative	Not Appropriate
I am against the fact that we should abolish capital punishment because I think that some people can not be rehabilitated and will only continue their violent acts unless they are put to death.	Feminist	Not Appropriate
I am against the fact that we should ban cosmetic surgery for minors because I think that minors should have a choice in what to do with their bodies.	Feminist	Not Appropriate
I am in favor of the fact that we should end racial profiling because I think that it enforces the stereotype that all people of a certain race are dangerous criminals.	Conservative	Not Appropriate
I am against the fact that we should ban missionary work because I think that missionaries help the less fortunate in third world countries such as setting up schools and clinics.	Christian	Appropriate
I am in favor of the fact that the eppo's jurisdiction should be extended to all member states because I think that the EU must take a strong position against corruption and be in favor of an independent supranational prosecutor that is the eppo.	Democrat	Appropriate
I am in favor of the fact that we should oppose collectivism because I think that collectivism reduces productivity from individuals as they have no incentive to do better.	Libertarian	Appropriate
I am in favor of the fact that esperanto should replace english as the first foreign language in schools and universities because! I think that to build a common european understanding and to create a european identity we need our own common language.	Conservative	Not Appropriate
I am against the fact that we should legalize prostitution because I think that legalizing prostitution can many women as well as children vulnerable to the dangerous and immoral legal acceptance of prostitution.	Feminist	Not Appropriate
am in favor of the fact that homeschooling should be banned because I think that homeschooling doesnt allow the child to have the same social aspect that they would attain at a regular school.	Christian	Not Appropriate
am in favor of the fact that we should fight urbanization because I think that urbanization would have a very negative effect on the environment because of pollution having clean air and water are always essential for a healthy society.	Vegetarian	Appropriate
I am against the fact that we should abandon marriage because I think that marriage is an important covenant relationship that causes people to work harder to stay together.	Feminist	Not Appropriate
I am against the fact that we should adopt genderneutral language because I think that genderneutral language is unnecessary.	Feminist	Not Appropriate

Human value detection

Ferraro Francesca

18/12/2023

Step 5 - Free approach

Under this "free approach," the model is allowed to select any word that it seems appropriate for completing the prompt. For each decision, the model provides a list of the chosen words along with their probabilities.

```
def generate word(model):
    # Initialize an empty list to store the results
    results = []
    # Create prompt
    prompt = f"i am {stance} the fact that {conclusion} because i think that {argument}. i am a {model.tokenizer.mask token}."
    # Generate the filling
    output = model(prompt)
    # Sort output by 'score'
    output.sort(key=lambda x: x['score'], reverse=True)
    # Create a dictionary of other predicted words with their scores
    predictions = {result['token str']: result['score'] for result in output}
    # Extract 'sequence' with higher 'score'
    description = output[0]['sequence']
    return predictions, description
generate word(model bert)
({'conservative': 0.22405116260051727,
  'libertarian': 0.13920997083187103,
  'journalist': 0.06951353698968887,
  'liberal': 0.04904218018054962,
  'republican': 0.044862568378448486},
 'i am in favor of the fact that we should subsidize journalism because i think that subsidizing journalism would help crowd
out the every increasing false news market if the false news market could not compete with legitimate news less people would
be exposed to dangerous false news. i am a conservative.')
```

Step 6 - Guided approach

This "guided approach" uses a list of political adjectives to guide the selection process. For each word, is calculated the probability.

```
def generate word adj(model, tokenizer, adjective list):
    # Create prompt
   prompt = f"i am {stance} the fact that {conclusion} because i think that {argument}. i am a {tokenizer.mask_token}."
    # Initialize a dictionary to store the probabilities
    probabilities = {}
   # For each word in the list, generate a score
   for word in adjective list:
       # Replace the mask token with the word
       new prompt = prompt.replace(tokenizer.mask token, word)
                                                                                          generate word adj(model, tokenizer, adjective list)[1:]
       # Encode the new prompt
       inputs = tokenizer.encode plus(new prompt, return tensors='pt')
                                                                                          ({'republican': 1.5237325290407e-10,
       # Generate the filling
                                                                                            'conservative': 9.366677572453241e-11,
       outputs = model(**inputs)
                                                                                            'liberal': 8.617905694618955e-11,
       logits = outputs.logits
                                                                                            'democrat': 4.991623087091668e-11,
       # Calculate the softmax probabilities from logits
                                                                                            'communist': 1.2314340519514744e-11,
                                                                                            'socialist': 1.1868956338589864e-11,
       softmax probs = torch.nn.functional.softmax(logits, dim=-1)
                                                                                            'capitalist': 7.94609256865586e-12,
       # Get the probability of the word
                                                                                            'libertarian': 6.847594540004831e-12.
       word id = tokenizer.encode(word, add special tokens=False)[0]
                                                                                            'progressive': 1.3555856740940508e-12.
       word_prob = softmax_probs[0, -1, word_id].item()
                                                                                            'centrist': 6.949596063551833e-13,
                                                                                            'anarchist': 4.94443516078219e-13}.
       # Store the probability of the word
       probabilities[word] = word prob
                                                                                          be exposed to dangerous false news. i am a republican.')
    # Sort the probabilities dictionary by value in descending order
    sorted probabilities = {k: v for k, v in sorted(probabilities.items(), kev=lambda item: item[1], reverse=True)}
    # Choose the word with the highest probability
    top word = next(iter(sorted probabilities))
    # Replace the mask token in the original prompt with the top word
    description = prompt.replace(tokenizer.mask token, top word)
    return probabilities, sorted probabilities, description
```

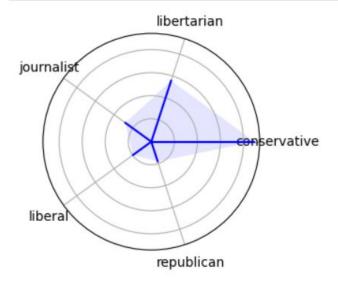
```
# model inizialization for adjectiveList-based approach
                                       def load model() :
                                           # Load the model and tokenizer
                                           model = BertForMaskedLM.from pretrained('bert-base-uncased')
                                           tokenizer = BertTokenizer.from pretrained('bert-base-uncased')
                                           return model, tokenizer
                                       model, tokenizer = load model()[0], load model()[1]
                                              adjective list =
                                                   "conservative", "republican", "capitalist",
                                                   "libertarian", "centrist", "democrat",
                                                   "liberal", "progressive", "socialist",
                                                  "communist", "anarchist",
'i am in favor of the fact that we should subsidize journalism because i think that subsidizing journalism would help crowd
out the every increasing false news market if the false news market could not compete with legitimate news less people would
```

Step 7 - Plots for data visualization

Radial plots for a better comprehension of the results.

"Free approach"

radial plot(generate word(model bert)[0])



"Guided approach"

Results are shown in order of political proximity, indicating if the model learns towards a specific political side

radial_plot(generate_word_adj(model, tokenizer, adjective_list)[0])



Results

Here are some examples of the model's predictions, containing the two approaches compared.

PROMPT	PREDICTION	PREDICTION w/ ADJECTIVE
am in favor of the fact that we should fight against urbanization because i think that urbanization results in increased population density which leads to pressure on resources such as and water medical professionals and police.	Conservative	Republican
i am in favor of the fact that we should legalize cannabis because i think that in many cases cannabis has wonderful health results such as taking away pain from ill people it should be legal for all.	Vegetarian	Liberal
am against the fact that we should adopt genderneutral language because i think that gender neutral anguage can be very confusing and simply serves to emphasize the difference between those who consider themselves binary and those who do not.	Feminist	Liberal
a m in favor of the fact that we should abandon marriage because I think that marriage is antiquated so many people now live together as lifelong partners therefore marriage is an unnecessary act.	Feminist	Republican
i am against the fact that we should subsidize hydrogen vehicles because i think that if the hydrogen fuel escapes into the air it can increase global warming.	Vegetarian	Republican
i am in favor of the fact that we should ban algorithmic trading because i think that algorithmic trading will put many stock brokers and analysts out of business.	Libertarian	Republican
i am against the fact that we should ban factory farming because i think that factory farming is an efficient way to produce cheap food for the masses.	Vegetarian	Republican
i am against the fact that we should ban naturopathy because i think that it is safer than traditional medicine and does work.	Vegetarian	Republican
i am against the fact that we should subsidize wikipedia because i think that wikipedia has always been free to use and should continue to be free without subsidies.	Conservative	Republican
i am against the fact that we need a common migration and asylum policy based on respect for rights and equal treatment because it hink that giving immigrants more rights will lead to the forced displacement and eventual erasure of indigenous europeans this is criminal if whats happening to the uyghurs and tibetans in china is genocide then so is this this idea proposal is not grassroots its astroturfed lies to hide the fact that our migration policies are totally undemocratic and forced onto native europeans.	Conservative	Liberal
i am in favor of the fact that the eu should have common taxation and social policy because i think that the future is work mobility thus we need to simplify taxation enhance remote working opportunities and enable intercountry recruitment for companies.	Democrat	Liberal
i am in favor of the fact that payday loans should be banned because i think that payday loans should be banned because they are not held to the same standards as loans for middle class people.	Conservative	Republican
i am in favor of the fact that we should abandon television because i think that we should abandon television because it is a waste of time people could get so much more done without it.	Conservative	Republican
i am against the fact that consumerism brings more harm than good because i think that the middle class benefited tremendously from consumerism increased job opportunities wages and cheaper goods helped raise their living standards tremendously the new world order gave them hope that even they could have luxuries like the elite through business they could climb up the social adder.	Conservative	Liberal
i am in favor of the fact that we should ban the use of child actors because i think that children should be learning and playing not working.	Conservative	Liberal

i am against the fact that we should fight urbanization because i think that urbanization provides better schools for people who might live outside of a city.	Conservative	Republican
i am against the fact that we should link aadhaar card with the voter id because i think that there are lake and duplicate aadhaar cards too so linking aadhaar cards with voter ids may complicate the issue.	Conservative	Conservative
i am in favor of the fact that we should ban the use of child actors because i think that we should ban the use of child actors because they should be in school and learning.	Feminist	Republican
am in favor of the fact that the eu should recognize taiwan as an independent sovereign state because i think that china is acting against european values and violating intellectual property rights.	Conservative	Republican
am against the fact that we should ban fast food because I think that fast food is a cheap treat and to san it would be to take the joy out of life for many people.	Vegetarian	Republican
am in favor of the fact that we should legalize cannabis because i think that it will stop people from buying the drug from the streets it will also mean that regulations are put in place to ensure the quality of the drug is safe to use it will also increase tax revenue.	Republican	Liberal
am in favor of the fact that there should be a ban on bandhs because i think that though right to protest is a fundamental right people do not have right to take away the fundamental rights of others such as going to their work doing their personal works et con teveryone is against to the changes by government some people may support it and may not support the ban forcing them to stay at home is undemocratic.	Conservative	Conservative
am in favor of the fact that holocaust denial should be a criminal offence because I think that holocaust denial should be a criminal offence because its main goal is to make jewish people feel in danger making it a hate crime.	Conservative	Liberal
i am against the fact that we should abolish the right to keep and bear arms because i think that we should not abolish the right to keep and bear arms because people would be left defenseless against criminals who are going to find a way to get a gue.	Libertarian	Liberal
i am against the fact that we should abolish capital punishment because i think that capital punishment is a real deterant to crime and should not be abolished.	Feminist	Republican
i am against the fact that we should cancel pride parades because I think that we should not cancel pride parades because people should have the right to express what they want and how they feel about a certain issue.	Feminist	Liberal
i am against the fact that we should subsidize space exploration because i think that space exploration should not be subsidized because it is not affordable there is not enough money available in mostcountries to even support health care and security without space exploration too.	Libertarian	Conservative
i am in favor of the fact that we should introduce compulsory voting because i think that compulsory voting gives everyone in society their say in politics.	Conservative	Republican
i am against the fact that we should ban the use of child actors because i think that not all tv and movies is going to be adults only we need child actors to shows the world as it is.	Feminist	Liberal
am against the fact that we should ban telemarketing because i think that if someone does not want to engage with telemarketers they can just hang up or sign off.	Feminist	Republican
i am in favor of the fact that we should legalize polygamy because i think that polygamy has been going on for hundreds of years.	Feminist	Republican
i am in favor of the fact that all workers and unemployed should have access to social protection including young people because i think that we cannot afford a lost generation so we need to invest in quality lobs and young people for our future.	Conservative	Liberal

Human value detection

Ferraro Francesca

04/01/2024

Step 8 - Mapping words to the vocabulary

A word-value mapping vocabulary was used for associating the predicted words to one or more human values. For the words that weren't present into the vocabulary, it was calculated the closest word and the corresponding human values.

```
en path = 'C:/Users/ffraa/Desktop/Università/Tesi/vast-value-map-main/valuemap/Data/wiki.en.vec'
n max = 100000
eng = Model(en path, n max=n max)
oracle = MultiModel(eng=eng)
vocabulary path = 'C:/Users/ffraa/Desktop/Università/Tesi/vast-value-map-main/valuemap/Refined dictionary.txt'
value map = ValueMap.from vocabulary(vocabulary path)
value oracle = ValueSearch(valuemap=value map, oracle=oracle, k=10, depth=2)
# Load pretrained word2vec model
path = 'C:/Users/ffraa/Desktop/Università/Tesi/vast-value-map-main//valuemap/GoogleNews-vectors-negative300.bin'
similarity model = KeyedVectors.load word2vec format(path, binary=True)
def calculate oracle answer(word):
    answers = value oracle.search(word, lang='eng')
    aggregated = value oracle.aggregated search(word, lang='eng')
    if answers is not None:
        # Convert the aggregated values to a series and normalize them
        aggregated series = pd.Series(aggregated).sort values(ascending=False) / sum(aggregated.values())
        return answers, aggregated series
```

Step 9 - Similarity calculation

For finding the closest words, they were used word2vec and cosine similarity.

```
def find closest word(word, dictionary, model):
   # Get the embedding for the predicted word
   word embedding = model[word]
   max similarity = -1
   closest word = None
   # Iterate over all words in the dictionary
   for dict word in dictionary.keys():
       try:
            # Get the embedding for the word from the dictionary
            dict word embedding = model[dict word]
           # Calculate the cosine similarity between the two embeddings
            similarity = cosine similarity([word embedding], [dict word embedding])
           # If the similarity is higher than the current maximum, update the maximum
           if similarity > max similarity:
               max similarity = similarity
               closest word = dict word
        except KeyError:
            continue # Continue if the word isn't into the model
    return closest word
```

Step 10 - Human Value Detection

return normalized_values

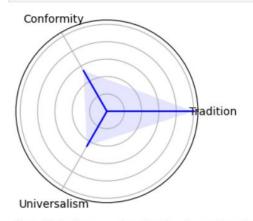
```
def human value detection(word, dictionary, model, aggregated values):
    try:
        # Calculate answers
        print(f"\n{word}")
        answers, aggregated = calculate oracle answer(word)
        if not answers :
             print(f"No values available for the word. Finding the closest word in the dictionary...")
             closest word = find closest word(word, dictionary, model)
             print(f"The closest word in the dictionary for {word} is {closest word}.")
             # Closest word's answers
             answers, aggregated = calculate_oracle_answer(closest word)
        # Print answers
        sorted answers = sorted(answers, key=lambda x: x['similarity'], reverse=True)
        for answer in sorted answers:
                                                                                    aggregated values = {}
             print(f"{answer}")
                                                                                    for word in generate word(model bert)[0].kevs():
                                                                                        aggregated values = human value detection(word, value map, similarity model, aggregated values)
        # Add agaregate values
        for value, probability in aggregated.items():
                                                                                    conservative
             if value in aggregated values:
                                                                                    {'value': 'Conformity', 'word': 'conservative', 'similarity': 1.0, 'depth': 0}
                 aggregated values[value] += probability
                                                                                    {'value': 'Universalism', 'word': 'liberal', 'similarity': 0.8551696815099271, 'depth': 0}
             else:
                                                                                    libertarian
                 aggregated values[value] = probability
                                                                                    No values available for the word. Finding the closest word in the dictionary...
    except KeyError as e:
                                                                                    The closest word in the dictionary for libertarian is liberal.
        print(e)
                                                                                    {'value': 'Universalism', 'word': 'liberal', 'similarity': 1.0, 'depth': 0}
                                                                                    {'value': 'Conformity', 'word': 'conservative', 'similarity': 0.855169681509927, 'depth': 0}
    # Normalize agaregated values
                                                                                    {'value': 'Universalism', 'word': 'democrats', 'similarity': 0.6624991470620656, 'depth': 0}
    total probability = sum(aggregated values.values())
    normalized values = {value: probability / total probability for value, probability in aggregated values.items()}
```

Step 11 - Human Value Detection Plots

Firstly, for both of the approaches, the similarity method was used for mapping words into values, and the results were plotted with the normalized probabilities.

"Free approach"

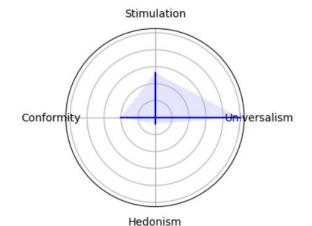
```
sorted_aggregated_values = sort_elements(aggregated_values)
radial_plot(sorted_aggregated_values)
print(sorted_aggregated_values)
```



{'Tradition': 0.5, 'Conformity': 0.2695171255672143, 'Universalism': 0.23048287443278567}

"Guided approach"

```
sorted_aggregated_values_adj = sort_elements(aggregated_values_adj)
radial_plot(sorted_aggregated_values_adj)
print(sorted_aggregated_values_adj)
```



{'Universalism': 0.49996862164606326, 'Stimulation': 0.26426107451247777, 'Conformity': 0.20064113234152348, 'Hedonism': 0.0351291714999355}

Step 12 - Guided approach for HV Detection

For the political adjectives from the "guided approach", another mapping procedure was used: each one of the elements in the adjective list was previously associated to a set of human values.

```
# Import the human values from values.py labels
from valuemap.values import VALUE LABELS
# For the guided approach, it's used an a-priori association between adjectives and human values
mapped adjective values = {
    "conservative": ['1', '3', '2'],
    "republican": ['1', '9', '10'],
    "capitalist": ['6', '9', '10'],
    "libertarian": ['6', '7', '8'],
    "centrist": ['4', '5', '2'],
    "democrat": ['5', '4', '6'],
    "liberal": ['5', '7', '6'],
    "progressive": ['5', '7', '4'],
    "socialist": ['5', '4', '1'],
    "communist": ['5', '1', '4'],
    "anarchist": ['6', '7', '8'],
for adjective, values in mapped adjective values.items():
    print(f"{adjective}: {[VALUE LABELS[value] for value in values]}")
conservative: ['Security', 'Tradition', 'Conformity']
republican: ['Security', 'Achievement', 'Power']
capitalist: ['Self-Direction', 'Achievement', 'Power']
libertarian: ['Self-Direction', 'Stimulation', 'Hedonism']
centrist: ['Benevolence', 'Universalism', 'Conformity']
democrat: ['Universalism', 'Benevolence', 'Self-Direction']
liberal: ['Universalism', 'Stimulation', 'Self-Direction']
progressive: ['Universalism', 'Stimulation', 'Benevolence']
socialist: ['Universalism', 'Benevolence', 'Security']
communist: ['Universalism', 'Security', 'Benevolence']
```

anarchist: ['Self-Direction', 'Stimulation', 'Hedonism']

For each of them, the probability is calculated.

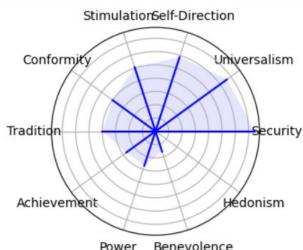
```
desc order = {}
value probs = {}
for adj, prob in desc order probabilities.items():
    human_value = mapped_adjective_values.get(adj)
   if human value is not None:
        human_value = [VALUE_LABELS.get(k) for k in human_value]
        if None not in human value:
            desc order[tuple(human value)] = prob
for values, prob in desc order.items():
    for value in values:
        if value not in value probs:
            value_probs[value] = 0
        value probs[value] += prob
total prob = sum(value probs.values())
# Normalize
normalized probs = {k: v / total prob for k, v in value probs.items()}
```

Step 13 - Guided approach Plot

Plot of the guided approach 's results.

In this scenario, each one of the human values that appears in this plot, was associated to one or more of the political adjectives in the list and, for each of them, the normalized sum of probabilities derived from the adjective probabilities is showed.

radial plot(normalized probs) normalized probs



Benevolence Power

{'Security': 0.18611482657363146, 'Universalism': 0.1658513487918742, 'Self-Direction': 0.14645849072887018, 'Stimulation': 0.12655240169024354, 'Conformity': 0.09943543460690181, 'Tradition': 0.09929246981114936, 'Achievement': 0.06796276988558812, 'Power': 0.06796276988558812, 'Benevolence': 0.04014274318143164, 'Hedonism': 0.00022674484472167382}

Human value detection

Ferraro Francesca

23/01/2024

Validation set loading

For the comparison between the results and the SemEval's ground truth, the entire validation set was loaded, with also the resulting labels for each argument.

Then, data were processed by iteration on the dataset contents.

```
def load_dataset():
    # Load the arguments and labels into separate DataFrames
    df_arguments = pd.read_csv('arguments-validation.tsv', delimiter='\t')
    df_labels = pd.read_csv('labels-validation.tsv', delimiter='\t')

# Merge the two DataFrames on the 'Argument ID' column
    df = pd.merge(df_arguments, df_labels, on='Argument ID')

# Extract the argument text from each DataFrame
    id = df_arguments['Argument ID'].tolist()
    arguments = df_arguments['Premise'].tolist()
    stances = df_arguments['Stance'].tolist()
    conclusions = df_arguments['Conclusion'].tolist()

return arguments, stances, conclusions, id, df
```

SemEval data extraction

SemEval results will be used as ground truth for evaluating the proposed method, so they are stored into a dictionary containing argument IDs and their associated human values.

```
def get all human values(df):
    # Initialize an empty dictionary to store the human values for each argument ID
   human values dict = {}
   # Iterate over all rows in the DataFrame
   for _, row in df.iterrows():
       # Get the argument ID
       argument id = row['Argument ID']
       # Extract the human values
       values dict = row.iloc[4:].to dict() # Adjust the column index as needed
       # Get a list of human values that have value 1
       human values = [key.split(':')[0] for key, value in values dict.items() if value == 1]
       # Convert the list to a set to remove duplicates, then convert it back to a list
       human values = list(set(human values))
       # Store the human values in the dictionary
       human values dict[argument id] = human values
    return human values dict
semeval validation set results = get all human values(load dataset()[4])
print(semeval validation set results)
{'A01001': ['Security'], 'A01012': ['Universalism'], 'A02001': ['Security', 'Universalism'], 'A02002': ['Self-direction'], 'A02009': ['Universalism'.
'Conformity'], 'A02018': ['Universalism'], 'A03005': ['Humility', 'Conformity'], 'A03014': ['Security', 'Conformity'], 'A03018': ['Security', 'Face'],
'A04005': ['Achievement', 'Universalism'], 'A04012': ['Security', 'Universalism'], 'A04019': ['Self-direction', 'Security'], 'A05002': ['Self-directio
n', 'Benevolence', 'Security', 'Universalism'], 'A05004': ['Benevolence', 'Security'], 'A05007': ['Security', 'Face'], 'A05010': ['Self-direction', 'S
```

Results evaluation for "free approach"

The same structure (dict.) was used for collecting the results obtained by each of the proposed approaches. First of all, the "free approach":

```
# Initialize an empty dictionary to store the human values for each argument ID
human values dict = {}
# Iterate over all arguments, stances, and conclusions
for argument, stance, conclusion, argument id in zip(arguments, stances, conclusions, id):
              # Generate words using model
              generated_words = generate_word(stance, conclusion, argument, model_bert)
              # Plot the generated words
                                                                                                                                                                                                                                transformed_dict = {k: list(v.keys()) for k, v in human_values_dict.items()}
             radial plot(generated words[0])
                                                                                                                                                                                                                                print(transformed dict)
                                                                                                                                                                                                                                 {'A01001': ['Universalism', 'Conformity'], 'A01012': ['Universalism', 'Conformity'], 'A02001': ['Universalism', 'Conformity'], 'A02002': ['Conformit
              # Initialize an empty dictionary for agaregated values
                                                                                                                                                                                                                                y', 'Benevolence', 'Universalism'], 'A02009': ['Stimulation', 'Conformity', 'Universalism'], 'A02018': ['Conformity', 'Universalism'], 'Self-Directio
                                                                                                                                                                                                                                n'], 'A03005': ['Hedonism', 'Tradition', 'Universalism', 'Conformity'], 'A03014': ['Universalism', 'Conformity'], 'A03018': ['Tradition', 'Universalism', 'Conformity'], 'A03018': ['Universalism', 'Conformity'], 'A03018': ['Tradition', 'Universalism', 'Conformity'], 'A03018': ['Universalism', 'Conformity', 'Conformity', 'Conformity', 'Conformity', 'Conformity', 'Conformity', 'Conform
              aggregated values = {}
                                                                                                                                                                                                                                m', 'Conformity', 'Hedonism'], 'A04005': ['Conformity', 'Self-Direction', 'Tradition', 'Universalism'], 'A04012': ['Benevolence', 'Security', 'Traditi
                                                                                                                                                                                                                                on', 'Universalism', 'Conformity'], 'A04019': ['Tradition', 'Hedonism', 'Universalism', 'Conformity'], 'A05002': ['Universalism', 'Conformity', 'Conformity', 'Conformity', 'Conformity', 'Conformity', 'Conformity', 'Conformity', 'Conformity'
              # Detect human values for each generated word
              for word in generated words[0].keys():
                            aggregated values = human value detection(word, value map, similarity model, aggregated values)
              # Sort the agareaated values
              sorted aggregated values = sort elements(aggregated values)
              # Plot the sorted agaregated values
              radial plot(sorted aggregated values)
              # Print the sorted aggregated values
              print(sorted aggregated values)
              # Store the sorted aggregated values in the dictionary
              human values dict[argument id] = sorted aggregated values
```

Results evaluation for "guided approach"

Then for "guided approach":

human values dict guided[argument id] = sorted aggregated values

```
# Initialize an empty dictionary to store the human values for each argument ID
human values dict guided = {}
# Iterate over all arguments, stances, and conclusions
for argument, stance, conclusion, argument id in zip(arguments, stances, conclusions, id):
              # Generate words using model
              generated words = generate word adj(stance, conclusion, argument, model, tokenizer, adjective list)[1]
              # Plot the generated words
              radial plot(generate word adj(stance, conclusion, argument, model, tokenizer, adjective list)[0])
                                                                                                                                                                                                                                      transformed_dict_guided = {k: list(v.keys()) for k, v in human_values_dict_guided.items()}
              # Initialize an empty dictionary for aggregated values
                                                                                                                                                                                                                                      print(transformed dict guided)
              aggregated values adj = {}
                                                                                                                                                                                                                                      {'A01001': ['Universalism', 'Conformity', 'Stimulation', 'Hedonism', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Conformity', 'Hedonism', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Hedonism', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Conformity', 'Hedonism', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Hedonism', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Conformity', 'Hedonism', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Tradition'], 'A01012': ['Universalism', 'Stimulation', 'Tradition'], 'Tradition', 'Tradi
                                                                                                                                                                                                                                       'Tradition'], 'A02001': ['Universalism', 'Conformity', 'Stimulation', 'Hedonism', 'Tradition'], 'A02002': ['Stimulation', 'Universalism', 'Conformit
              # Detect human values for each generated word
                                                                                                                                                                                                                                      y', 'Hedonism', 'Tradition'], 'A02009': ['Universalism', 'Stimulation', 'Conformity', 'Hedonism', 'Tradition'], 'A02018': ['Universalism', 'Stimulation']
                                                                                                                                                                                                                                      n', 'Conformity', 'Hedonism', 'Tradition'], 'A03005': ['Stimulation', 'Universalism', 'Conformity', 'Hedonism', 'Tradition'], 'A03014': ['Universalism', 'Tradition'], 'A03014': ['Universalism
              for word in generated words.kevs():
                            aggregated values adj = human value detection(word, value map, similarity model, aggregated values adj)
              # Sort the aggregated values
              sorted aggregated values = sort elements(aggregated values adi)
              # Plot the sorted agaregated values
              radial plot(sorted aggregated values)
              # Print the sorted agareaated values
              print(sorted aggregated values)
              # Store the sorted aggregated values in the dictionary
```

Results evaluation for "guided approach" v.2

print(normalized probs)

human values dict guidedValues[argument id] = normalized probs

For the "guided approach" that uses a "guided mapping", the same method was used (only for the human value detection step, since the word generation is the same as the guided approach with free mapping) and were considered the first 3 human values detected (sorted by score):

```
human values dict guidedValues = {}
for argument, stance, conclusion, argument id in zip(arguments, stances, conclusions, id):
    desc order = {}
    value probs = {}
    for adj, prob in generate word adj(stance, conclusion, argument, model, tokenizer, adjective list)[1].items():
         human value = mapped adjective values.get(adj)
         if human value is not None:
             human value = [VALUE LABELS.get(k) for k in human value]
             if None not in human value:
                  desc order[tuple(human value)] = prob
                                                              transformed dict guidedValues = {k: list(v.kevs())[:3] for k, v in human values dict guidedValues.items()}
    for values, prob in desc order.items():
                                                             print(transformed dict guidedValues)
         for value in values:
             if value not in value probs:
                                                              {'A01001': ['Security', 'Achievement', 'Power'], 'A01012': ['Security', 'Tradition', 'Conformity'], 'A02001': ['Security', 'Achievement', 'Power'], 'A
                                                              02002': ['Security', 'Achievement', 'Power'], 'A02009': ['Security', 'Achievement', 'Power'], 'A02018': ['Security', 'Achievement', 'Power'], 'A0300
                  value probs[value] = 0
                                                             5': ['Universalism', 'Stimulation', 'Self-Direction'], 'A03014': ['Security', 'Achievement', 'Power'], 'A03018': ['Security', 'Tradition', 'Conformit
             value probs[value] += prob
                                                              y'], 'A04005': ['Universalism', 'Stimulation', 'Self-Direction'], 'A04012': ['Universalism', 'Stimulation', 'Self-Direction'], 'A04019': ['Security',
    total prob = sum(value probs.values())
    # Normalize
    normalized probs = {k: v / total prob for k, v in value probs.items()}
    radial plot(normalized probs)
```

Combined results evaluation

For the final evaluation, F1 and accuracy metrics were calculated for each argument ID, and then the model performances were evaluated on the average scores for each of the 3 methods.

```
def calculate average metrics(predicted dict, ground truth dict):
    total f1 = 0
    total accuracy = 0
    count = 0
   for key in ground truth dict.keys():
        if key in predicted dict:
            # Converting human values to binary format for calculation
           human_values = list(set(ground_truth_dict[key] + predicted_dict[key]))
           y true binary = [1 if value in ground truth dict[key] else 0 for value in human values]
           y pred binary = [1 if value in predicted dict[key] else 0 for value in human values]
            # Calculating F1 Score
           f1 = f1 score(y true binary, y pred binary, average='binary')
           # Calculating Accuracy
            accuracy = accuracy score(y true binary, y pred binary)
            total_f1 += f1
            total accuracy += accuracy
            count += 1
   # Calculating average F1 and Accuracy
    avg f1 = total f1 / count if count > 0 else 0
    avg accuracy = total accuracy / count if count > 0 else 0
    return avg f1, avg accuracy
```

Results

GUIDED APPROACH - GUIDED MAPPING = F1 Score: 0.3067080920720138, Accuracy: 0.2030697458308209

```
S 0.15
f1 free, accuracy free = calculate average metrics(transformed dict, semeval validation set results)
print(f"FREE APPROACH = F1 Score: {f1}, Accuracy: {accuracy}")
                                                                                                                         0.10
FREE APPROACH = F1 Score: 0.3150269649478491, Accuracy: 0.20982589989710146
                                                                                                                         0.05
f1 guided, accuracy guided = calculate average metrics(transformed dict guided, semeval validation set results)
print(f"GUIDED APPROACH - FREE MAPPING = F1 Score: {f1}, Accuracy: {accuracy}")
                                                                                                                         0.00
                                                                                                                              Free approach
                                                                                                                                                 Guided approach
                                                                                                                                                                  Guided Mapping approach
GUIDED APPROACH - FREE MAPPING = F1 Score: 0.2547531933607879, Accuracy: 0.15838065327780462
                                                                                                                                                     Models
f1 guidedMapping, accuracy guidedMapping = calculate average metrics(transformed dict guidedValues, semeval validation set results)
print(f"GUIDED APPROACH - GUIDED MAPPING = F1 Score: {f1}, Accuracy: {accuracy}")
```

Scores by model

Accuracy

0.30

0.25

0.20

As can be seen, the model that performs better is the one with the "free approach", followed by the "guided mapping approach" (i.e. with the list of human values associated to each political adjective), and the worst is the guided approach with free mapping.