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Safetensors ①

♦ Inference API ①

4,041,901

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☐ Use this model

♦ Cold ~

幻 Deploy ~

Model size | 109M params | Tensor type | I64 · F32 | 7

S Train >

```
from transformers import BertTokenizer, BertForSequenceClassification
def personality_detection(text):
   tokenizer = BertTokenizer.from_pretrained("Minej/bert-base-personality")
   model = BertForSequenceClassification.from_pretrained("Minej/bert-base-person
   inputs = tokenizer(text, truncation=True, padding=True, return_tensors="pt")
   outputs = model(**inputs)
   predictions = outputs.logits.squeeze().detach().numpy()
   label_names = ['Extroversion', 'Neuroticism', 'Agreeableness', 'Conscientious
   result = {label_names[i]: predictions[i] for i in range(len(label_names))}
    return result
```

based on the given input text.

Result Format

Extroversion: A value between 0 and 1 representing the predicted extroversion trait. Neuroticism: A value between 0 and 1 representing the predicted neuroticism trait. Agreeableness: A value between

The dictionary contains the following personality traits with their corresponding predicted values:

The personality_detection function returns a dictionary containing the predicted personality traits

0 and 1 representing the predicted agreeableness trait. Conscientiousness: A value between 0 and 1 representing the predicted conscientiousness trait. Openness: A value between 0 and 1 representing

the predicted openness trait. text_input = "I am feeling excited about the upcoming event." personality_prediction = personality_detection(text_input)

```
Output:
      "Extroversion": 0.535,
```

"Neuroticism": 0.576, "Agreeableness": 0.399,

print(personality_prediction)

"Conscientiousness": 0.253,

"Openness": 0.563

```
Note: The values in the example output are just placeholders and may not reflect the actual
predictions.
You can modify the example code and the result format to match your specific use case and desired
output format.
```

Transfer Learning for Big Five Personality Prediction In machine learning, training accurate models can be challenging when labeled data is limited.

Transfer learning offers a solution by leveraging pre-existing labeled data from a similar task or

domain. By transferring knowledge learned from one task to another, we can overcome data scarcity and train more effective models.

In this project, we used transfer learning with the BERT BASE UNCASED model to predict Big Five

personality traits. The model was fine-tuned on a curated dataset for personality traits, learning

Model Description

patterns between input text and personality characteristics. By applying transfer learning, we improved the accuracy of personality trait predictions.

By leveraging transfer learning and fine-tuning BERT BASE UNCASED, we accurately predict an individual's Big Five personality traits based on their input text. This approach addresses the challenges of limited labeled data in personality prediction, providing insights into individuals' personalities.

Language(s) (NLP): English License: MIT Finetuned from model [optional]: https://huggingface.co/bert-base-uncased

This project showcases the power of transfer learning in machine learning and highlights the

effectiveness of BERT BASE UNCASED for predicting Big Five personality traits.

Direct Use

predictions for the Big Five personality traits.

standalone personality prediction model.

Model type: BERT BASE UNCASED

- The personality prediction model can be used directly by individuals who are interested in gaining
- insights into their own personality traits based on their input text. Users can input text and receive

Uses

Downstream Use

Out-of-Scope Use

matters.

Bias, Risks, and Limitations

The personality prediction model, like any machine learning model, has certain limitations and

This model is not suitable for uses beyond personality prediction. It should not be used for making

critical decisions or judgments about individuals in areas such as employment, education, or legal

This model is not intended for downstream use or fine-tuning for specific tasks. It is designed as a

potential biases that should be taken into account:

Limited Context:

Privacy Concerns:

Interpret with Context:

Data Privacy and Security:

Promote Ethical Use:

considerations.

import torch

Generalization: The model predicts personality traits based on patterns learned from a specific di Ethical Considerations:

Personality prediction models should be used responsibly, with an understanding t

The model relies on user-provided input text, which may contain sensitive or personal

The model makes predictions based on input text alone and may not capture the full

```
False Positives/Negatives:
 The model's predictions may not always align perfectly with an individual's actual
Recommendations
To mitigate risks and limitations associated with personality prediction models, the following
recommendations are suggested:
  Awareness and Education:
  Users should be informed about the limitations and potential biases of the model.
  Avoid Stereotyping and Discrimination:
  Users should be cautious about making judgments or decisions solely based on pred
```

Interpret the model's predictions in the appropriate context and consider addition

Ensure that user data is handled securely and with respect to privacy regulations

Encourage responsible use of personality prediction models and discourage misuse

It is important to note that the above recommendations are general guidelines, and further context-

specific recommendations should be developed based on the particular use case and ethical

How to Download the Model If you would like to download the model files and use them instead of the Hosted inference API, then you can follow the code snippet provided below:

from transformers import BertForSequenceClassification, BertTokenizer

model = BertForSequenceClassification.from_pretrained(".", num_labels=5)

tokenizer = BertTokenizer.from_pretrained('.', do_lower_case=True)

"Extroversion": 0, "Neuroticism": 1, "Agreeableness": 2,

model.config.label2id = {

"Conscientiousness": 3,

Initialization of the model values

"Openness": 4, model.config.id2label = {

```
"0": "Extroversion",
      "1": "Neuroticism",
      "2": "Agreeableness",
      "3": "Conscientiousness",
      "4": "Openness",
  def personality_detection(model_input: str) -> dict:
      Performs personality prediction on the given input text
      Args:
          model_input (str): The text conversation
      Returns:
          dict: A dictionary where keys are speaker labels and values are their per
      1.1.1
      if len(model_input) == 0:
          ret = {
               "Extroversion": float(0),
              "Neuroticism": float(0),
              "Agreeableness": float(0),
              "Conscientiousness": float(0),
               "Openness": float(0),
          return ret
      else:
          dict_custom = {}
          preprocess_part1 = model_input[:len(model_input)]
          dict1 = tokenizer.encode_plus(preprocess_part1, max_length=1024, padding=
          dict_custom['input_ids'] = [dict1['input_ids'], dict1['input_ids']]
          dict_custom['token_type_ids'] = [dict1['token_type_ids'], dict1['token_type_ids']
          dict_custom['attention_mask'] = [dict1['attention_mask'], dict1['attention_mask']
          outs = model(torch.tensor(dict_custom['input_ids']), token_type_ids=None,
          b_logit_pred = outs[0]
          pred_label = torch.sigmoid(b_logit_pred)
          ret = {
              "Extroversion": float(pred_label[0][0]),
              "Neuroticism": float(pred_label[0][1]),
              "Agreeableness": float(pred_label[0][2]),
              "Conscientiousness": float(pred_label[0][3]),
              "Openness": float(pred_label[0][4]),
          return ret
  personality_prediction = personality_detection(text_input)
Make sure you have the required dependencies installed (transformers and torch). This code snippet
initializes the model, tokenizer, and configuration. It then defines the personality_detection
function, which takes a text conversation as input and returns a dictionary with personality
predictions for each speaker.
You can call the personality_detection function with your input text to obtain the personality
predictions. The personality_prediction variable will hold the resulting dictionary.
```

Please note that this code assumes you have already downloaded the necessary model files (config.json, pytorch_model.bin, special_tokens_map.json, tokenizer_config.json, vocab.txt) and placed them in the current directory (indicated by "."). Adjust the paths and filenames accordingly if needed.

Citation

@article{DBLP:journals/corr/abs-1810-04805,

= {Jacob Devlin and

TOS

Ming{-}Wei Chang and

Kenton Lee and Kristina Toutanova}, = {{BERT:} Pre-training of Deep Bidirectional Transformers for Langua title Understanding}, = {CoRR}, journal $= \{abs/1810.04805\},$ volume = {2018}, year = {http://arxiv.org/abs/1810.04805}, url archivePrefix = {arXiv}, = {1810.04805}, eprint timestamp = {Tue, 30 Oct 2018 20:39:56 +0100}, = {https://dblp.org/rec/journals/corr/abs-1810-04805.bib}, biburl bibsource = {dblp computer science bibliography, https://dblp.org}

Privacy

About

TBA

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