










 Anngladys / AI-Tools-Assignment



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Anngladys Winding up

2e937b0 · 2 minutes ago



496 lines (496 loc) · 19.4 KB

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```
In [1]: # Cell 1: Import Libraries
import spacy

# Load the English spaCy model (ensure you've run 'python -m spacy download en_core_web_sm')
try:
    nlp = spacy.load("en_core_web_sm")
    print("spaCy model loaded successfully!")
except OSError:
    print("SpaCy model not found. Please run 'python -m spacy download en_core_web_sm'")
    exit() # Exit if model not loaded
```

spaCy model loaded successfully!

```
In [2]: # Cell 2: Define Sample Review Texts
review_texts = [
    "The new iPhone 15 Pro is an amazing device. Apple has outdone themselves.",
    "This Samsung Galaxy S24 has a terrible battery life. Very disappointed with the brand.",
    "Excellent Bose QuietComfort headphones! Sound quality is superb.",
    "I bought a cheap knockoff charger, it stopped working in a week. Don't waste your money.",
    "The Sony PlayStation 5 is fantastic for gaming, but it's often out of stock.",
    "My new Kindle Oasis arrived quickly. It's great for reading, a truly wonderful experience.",
    "Terrible experience with this Dell XPS laptop, constant crashes."
]

print("Sample review texts defined.")
```

Sample review texts defined.

```
In [3]: # Cell 3: Perform Named Entity Recognition (NER)
print("--- Named Entity Recognition (NER) ---")
extracted_entities = []

for i, text in enumerate(review_texts):
    doc = nlp(text)
    entities_in_review = []
    print(f"\nReview {i+1}: \"{text}\"")
    for ent in doc.ents:
        # We're primarily interested in products, organizations, and people
        if ent.label_ in ["ORG", "PRODUCT", "GPE", "PERSON", "NORP"]: # Add more labels as needed
            entities_in_review.append({"text": ent.text, "label": ent.label_})
    print(f" - Entity: '{ent.text}' (Type: {ent.label_})")
    extracted_entities.append(entities_in_review)
```

--- Named Entity Recognition (NER) ---

Review 1: "The new iPhone 15 Pro is an amazing device. Apple has outdone the mselves."

- Entity: 'Apple' (Type: ORG)

Review 2: "This Samsung Galaxy S24 has a terrible battery life. Very disappointed with the brand."

Review 3: "Excellent Bose QuietComfort headphones! Sound quality is superb."
- Entity: 'Bose QuietComfort' (Type: PERSON)

Review 4: "I bought a cheap knockoff charger, it stopped working in a week. Don't waste your money."

Don't waste your money.

Review 5: "The Sony PlayStation 5 is fantastic for gaming, but it's often out of stock."

- Entity: 'Sony' (Type: ORG)
- Entity: 'PlayStation 5' (Type: PRODUCT)

Review 6: "My new Kindle Oasis arrived quickly. It's great for reading, a truly portable library."

- Entity: 'Kindle Oasis' (Type: ORG)

Review 7: "Terrible experience with this Dell XPS laptop, constant crashes."

- Entity: 'Dell XPS' (Type: ORG)

In [4]:

```
# Cell 4: Analyze Sentiment (Rule-Based Approach)
print("\n--- Sentiment Analysis (Rule-Based) ---")

positive_words = ["amazing", "excellent", "superb", "fantastic", "great",
negative_words = ["terrible", "disappointed", "stopped working", "waste",

def analyze_sentiment_rule_based(text):
    text_lower = text.lower()
    pos_score = sum(1 for word in positive_words if word in text_lower)
    neg_score = sum(1 for word in negative_words if word in text_lower)

    if pos_score > neg_score:
        return "Positive"
    elif neg_score > pos_score:
        return "Negative"
    else:
        return "Neutral" # Or if pos_score == neg_score

for i, text in enumerate(review_texts):
    sentiment = analyze_sentiment_rule_based(text)
    print(f"\nReview {i+1}: \"{text}\"")
    print(f" - Sentiment: {sentiment}")
```

--- Sentiment Analysis (Rule-Based) ---

Review 1: "The new iPhone 15 Pro is an amazing device. Apple has outdone the mselves."

- Sentiment: Positive

Review 2: "This Samsung Galaxy S24 has a terrible battery life. Very disappointed with the brand."

- Sentiment: Negative

Review 3: "Excellent Bose QuietComfort headphones! Sound quality is superb."

- Sentiment: Positive

Review 4: "I bought a cheap knockoff charger, it stopped working in a week. Don't waste your money."

- Sentiment: Negative

Review 5: "The Sony PlayStation 5 is fantastic for gaming, but it's often out of stock."

- Sentiment: Positive

Review 6: "My new Kindle Oasis arrived quickly. It's great for reading, a truly portable library."

- Sentiment: Positive

Review 7: "Terrible experience with this Dell XPS laptop, constant crashes."

- Sentiment: Negative

```
In [5]: # Cell 1: Import Libraries
import spacy
import pandas as pd
import random

print(f"spaCy Version: {spacy.__version__}")
print("Libraries imported successfully!")
```

spaCy Version: 3.8.7
Libraries imported successfully!

```
In [6]: # Cell 2: Load spaCy English Model
try:
    # Load the small English model
    nlp = spacy.load("en_core_web_sm")
    print("spaCy 'en_core_web_sm' model loaded successfully.")
except OSError:
    print("spaCy model 'en_core_web_sm' not found. Downloading...")
    spacy.cli.download("en_core_web_sm")
    nlp = spacy.load("en_core_web_sm")
    print("spaCy model 'en_core_web_sm' downloaded and loaded successfully")
```

spaCy 'en_core_web_sm' model loaded successfully.

```
In [7]: # Cell 3: Sample Text Data (Amazon Reviews style)
amazon_reviews = [
    "The product is excellent! Very happy with the purchase.",
    "Battery life is terrible, died after 2 hours. Very disappointed.",
    "Works as expected, good value for money. Highly recommended.",
    "This is the worst item I've ever bought. A complete waste of money.",
    "It's okay, not great, not bad. Just mediocre.",
    "Fantastic performance, totally exceeded my expectations!",
    "Wish it had more features, but it's decent for the price.",
    "The delivery was fast, but the item was damaged.",
    "Absolutely love this! The design is sleek and it's so easy to use.",
    "Received a broken one. Customer service was unhelpful."
]

print("Sample Amazon reviews loaded.")
```

Sample Amazon reviews loaded.

```
In [8]: # Cell 4: Tokenization, POS Tagging, and Lemmatization
print("--- Tokenization, POS Tagging, and Lemmatization ---")
for i, text in enumerate(amazon_reviews[:3]): # Process first 3 reviews for
    doc = nlp(text)
    print(f"\nReview {i+1}: '{text}'")
    print(f"{'Token':<15} {'Lemma':<15} {'POS':<10} {'Is Alpha?':<10} {'St
    print("-" * 70)
    for token in doc:
        print(f"{'str(token)':<15} {token.lemma_:<15} {token.pos_:<10} {str(
```

--- Tokenization, POS Tagging, and Lemmatization ---

Review 1: 'The product is excellent! Very happy with the purchase.'

Review 1: 'The product is excellent! Very happy with the purchase.'

| Token | Lemma | POS | Is Alpha? | Stopword? |
|-----------|-----------|-------|-----------|-----------|
| ----- | | | | |
| The | the | DET | True | True |
| product | product | NOUN | True | False |
| is | be | AUX | True | True |
| excellent | excellent | ADJ | True | False |
| ! | ! | PUNCT | False | False |
| Very | very | ADV | True | True |
| happy | happy | ADJ | True | False |
| with | with | ADP | True | True |
| the | the | DET | True | True |
| purchase | purchase | NOUN | True | False |
| . | . | PUNCT | False | False |

Review 2: 'Battery life is terrible, died after 2 hours. Very disappointed.'

| Token | Lemma | POS | Is Alpha? | Stopword? |
|--------------|--------------|-------|-----------|-----------|
| ----- | | | | |
| Battery | battery | NOUN | True | False |
| life | life | NOUN | True | False |
| is | be | AUX | True | True |
| terrible | terrible | ADJ | True | False |
| , | , | PUNCT | False | False |
| died | die | VERB | True | False |
| after | after | ADP | True | True |
| 2 | 2 | NUM | False | False |
| hours | hour | NOUN | True | False |
| . | . | PUNCT | False | False |
| Very | very | ADV | True | True |
| disappointed | disappointed | ADJ | True | False |
| . | . | PUNCT | False | False |

Review 3: 'Works as expected, good value for money. Highly recommended.'

| Token | Lemma | POS | Is Alpha? | Stopword? |
|-------------|-----------|-------|-----------|-----------|
| ----- | | | | |
| Works | work | NOUN | True | False |
| as | as | SCONJ | True | True |
| expected | expect | VERB | True | False |
| , | , | PUNCT | False | False |
| good | good | ADJ | True | False |
| value | value | NOUN | True | False |
| for | for | ADP | True | True |
| money | money | NOUN | True | False |
| . | . | PUNCT | False | False |
| Highly | highly | ADV | True | False |
| recommended | recommend | VERB | True | False |
| . | . | PUNCT | False | False |

```
In [9]: # Cell 5: Named Entity Recognition (NER)
print("\n--- Named Entity Recognition (NER) ---")
for i, text in enumerate(amazon_reviews):
    doc = nlp(text)
    if doc.ents:
        print(f"\nReview {i+1}: '{text}'")
        for ent in doc.ents:
            print(f"  Entity: {ent.text}, Type: {ent.label_}, SpaCy Expla")
    else:
        print(f"\nReview {i+1}: '{text}' - No entities found.")
```

--- Named Entity Recognition (NER) ---

Review 1: 'The product is excellent! Very happy with the purchase.' - No entities found.

Review 2: 'Battery life is terrible, died after 2 hours. Very disappointed.'

Entity: 2 hours, Type: TIME, SpaCy Explanation: Times smaller than a day

Review 3: 'Works as expected, good value for money. Highly recommended.' - No entities found.

Review 4: 'This is the worst item I've ever bought. A complete waste of money.' - No entities found.

Review 5: 'It's okay, not great, not bad. Just mediocre.' - No entities found.

Review 6: 'Fantastic performance, totally exceeded my expectations!'

Entity: Fantastic, Type: NORP, SpaCy Explanation: Nationalities or religious or political groups

Review 7: 'Wish it had more features, but it's decent for the price.' - No entities found.

Review 8: 'The delivery was fast, but the item was damaged.' - No entities found.

Review 9: 'Absolutely love this! The design is sleek and it's so easy to use.' - No entities found.

Review 10: 'Received a broken one. Customer service was unhelpful.' - No entities found.

In [10]:

```
# Cell 6: Basic Rule-Based Sentiment Analysis (Illustrative - very simple)
print("\n--- Basic Rule-Based Sentiment Analysis (Illustrative) ---")

positive_words = ["excellent", "happy", "good", "recommended", "fantastic"]
negative_words = ["terrible", "disappointed", "worst", "waste", "mediocre"]

def simple_sentiment(text):
    doc = nlp(text.lower()) # Process lowercase text
    sentiment_score = 0
    for token in doc:
        if token.text in positive_words:
            sentiment_score += 1
        elif token.text in negative_words:
            sentiment_score -= 1
    if sentiment_score > 0:
        return "Positive"
    elif sentiment_score < 0:
        return "Negative"
    else:
        return "Neutral"

for i, review in enumerate(amazon_reviews):
    sentiment = simple_sentiment(review)
    print(f"Review {i+1}: '{review}'\n Sentiment: {sentiment}\n")

print("\nNote: This is a very simplistic rule-based sentiment analysis.")
print("It lacks context understanding, sarcasm detection, and nuances. For")
print("Review: 'This is great, another broken item!' (Should be Negative)")
doc_sarcasm = nlp("This is great, another broken item!")
print(f" Simple rule-based analysis: {simple_sentiment(str(doc_sarcasm))}")
print("\nAdvanced NLP (like machine learning models or deep learning) is n
```

--- Basic Rule-Based Sentiment Analysis (Illustrative) ---

Review 1: 'The product is great! I love how easy it is to use.'

```
review 1: The product is excellent! very happy with the purchase.
```

```
Sentiment: Positive
```

```
Review 2: 'Battery life is terrible, died after 2 hours. Very disappointed.'
```

```
Sentiment: Negative
```

```
Review 3: 'Works as expected, good value for money. Highly recommended.'
```

```
Sentiment: Positive
```

```
Review 4: 'This is the worst item I've ever bought. A complete waste of money.'
```

```
Sentiment: Negative
```