



GPT-3

GPT + DesignSafe: AI-driven metadata extraction

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Using AI for metadata extraction

- Access DesignSafe Data Depot using TAPIS API
- Read different file formats (csv / xlsx) and extract metadata:
 - Column names
 - Value ranges
 - File association
- Push the extracted metadata to a semantic search engine like Neo4j for context-specific searchers
- Example datasets considered:
 - LEAP liquefaction datasets (csv)
 - Liquefaction CPT data from Maurer (xlsx)

LEAP dataset

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PRJ-3484 | LEAP-2022 - Effects of overburden stress, relative density, and static shear stress on cyclic strength of Ottawa F65 sand

Download Dataset

PI

Manzari, Majid

Project Type

Experimental

Natural Hazard Type

Earthquake

Awards

Liquefaction Experiments and Analysis Projects | CMMI 1635524

Related Work

PRJ-1783: LEAP-2017 GWU Laboratory Tests

PRJ-1780: LEAP-GWU-2015 Laboratory Tests

PRJ-2136: LEAP-Asia-2018: Stress-strain response of Ottawa sand in Cyclic Torsional Shear Tests

Stress-strain behavior and liquefaction strength characteristics of Ottawa F65 sand

Stress-strain behavior of Ottawa sand in cyclic direct simple shear and modeling of cyclic strength using Artificial Neural Networks

Keywords

Confining Stress, Cyclic Direct Simple Shear Test, Liquefaction, Relative Density, Sand, Static Shear Stress, Stress-Strain Response

DOIs in Project

10.17603/ds2-q3fz-kq91

10.17603/ds2-z7zb-bq89

10.17603/ds2-52dm-s780

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Description

The stress-strain-strength behavior of Ottawa F-65 sand is investigated through a series of stress-controlled Cyclic Direct Simple Shear (CDSS) tests conducted as part of the Liquefaction Experiments and Analysis Project (LEAP-2022). The tests were performed with two different overburden stresses on specimens prepared with different relative densities and subjected to different static (initial) shear stresses prior to cyclic shearing. The report included here presents the setup and experimental procedure, as well as the liquefaction strength curves and the results obtained...
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PRJ-3484

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Experiment Type

Cyclic Direct Simple Shear Tests

Author(s)

Lbibb, Sarra; Manzari, Majid

Experimental Facility

Geotechnical Engineering Laboratory - The George Washington University

Equipment Type

SGI-Type Direct Simple Shear Equipment

Date of Experiment

01-05-2020 — 03-04-2021

Date of Publication

11-05-2022

DOI

Citation

10.17603/ds2-q3fz-kq91

License(s)

Open Data Commons Attribution

Data Reuse

View Experiment Metrics

The stress-strain behavior and liquefaction strength of Ottawa F65 sand was investigated through an extensive series of cyclic direct simple shear (CDSS) tests. The study quantified the effects of static shear stress on the cyclic strength of Ottawa F65 sand. The database of CDSS tests was used to develop an Artificial Neural Networks model to predict Ottawa F65 cyclic strength.

Report | Summary Report on Ottawa F65 Sand CDSS Tests

Model Configuration | Direct Simple Shear Experimental Setup

➔ Sensor Information | Direct Simple Shear Test Sensors

➔ Event | CDSS tests with Sig_v = 100 kPa, Dr = 67%, Alpha = 0.3

<input checked="" type="checkbox"/> Name	Size
<input type="checkbox"/> eo_0_601_sigv_100_CSR_0_150_Tau_30_csv	-
<input type="checkbox"/> eo_0_601_sigv_100_CSR_0_200_Tau_30_csv	-
<input type="checkbox"/> eo_0_601_sigv_100_CSR_0_250_Tau_30_csv	- 3
<input type="checkbox"/> eo_0_601_sigv_100_CSR_0_300_Tau_30_csv	-

Trial 0: GPT-3 (DaVinci 003)

Prompt:

Summarize the following table with minimum and maximum values

"Cycle", "Shear --

Strain", "Shear

Stress", "Axial

Strain", "Axial

Stress", "Excess

Pressure"

0.008,-0.003181503,30.578,0,99.963,0

0.016,0.009387466,31.669,0,100.12,-0.15581

0.023,0.02851412,32.408,0,100.07,-0.1055

0.031,0.03817823,32.766,0,100.12,-0.15775

GPT-3 does a good job of extracting min and max values from the dataset presented. The summary also shows an accurate description of the behavior observed in the “limited” dataset shared with GPT-3

Response:

This table shows the relationship between shear strain, shear stress, axial strain, axial stress, and excess pressure for a cycle. It shows that as the shear strain increases, the shear stress, axial stress, and excess pressure increase, while the axial strain decreases. The minimum values for shear strain, shear stress, axial strain, axial stress, and excess pressure are 0.008, -0.003181503, 0, 99.963, and 0 respectively. The maximum values for shear strain, shear stress, axial strain, axial stress, and excess pressure are 0.133, 38.88, -0.0018094, 100.59, and -0.42067 respectively.

Trial 1: GPT-3 (DaVinci 003) - Computation

Prompt:

Summarize the following table with summary statistics as a JSON file

"Cycle","Shear --

Strain","Shear

Stress","Axial

Strain","Axial

Stress","Excess

Pressure"

0.008,-0.003181503,30.578,0,99.963,0

0.016,0.009387466,31.669,0,100.12,-0.15581

0.023,0.02851412,32.408,0,100.07,-0.1055

0.031,0.03817823,32.766,0,100.12,-0.15775

Response:

```
{
  "Shear Strain": {
    "mean": 0.077,
    "min": -0.003181503,
    "max": 0.2036224
  },
  "Shear Stress": {
    "mean": 35.072,
    "min": 30.578,
    "max": 38.88
  },
  "Axial Strain": {
    "mean": -0.001787,
    "min": 31.669,
    "max": 36.415
  },
  "Axial Stress": {
    "mean": -0.15581,
    "min": 30.578,
    "max": 37.7
  }
}
```

- GPT-3 got the column name wrong! (missing "Cycle" and "Excess Pressures" - due to badly formatted CSV file)
- GPT-3 does a good job of extracting min and max values correctly
- **GPT-3 mean calculation was wrong!**
 - **Axial stress column is inconsistent**

Mean values	GPT	Actual	Rel error (%)
Cycles	0.077	0.07035	9.45
Shear stress	35.072	35.708	1.78
Axial strain	-0.0018	-0.0005	237.17
Axial stress	-0.1558	100.303	100.16

Trial 2: ChatGPT - Generating code

Since GPT-3 is not so good at computations, why not generate code instead of the output directly. ChatGPT identified file as CSV and generated a code rather than the actual output. (we can also force it to generate a code using prompt engineering)

Prompt:

K

Generate a summary statistics for the following table as a JSON file:

"Cycle","Shear Strain","Shear Stress","Axial Strain","Axial Stress","Excess Pressure"

0.008	-0.003181503	30.578	0	99.963	0
0.016	0.009387466	31.669	0	100.12	-0.15581
0.023	0.02851412	32.408	0	100.07	-0.1055
0.031	0.03817823	32.766	0	100.12	-0.15775
0.039	0.05090463	33.442	-0.0018094	100.23	-0.26292
0.047	0.06355149	34.476	-1.12E-05	100.23	-0.26292
0.055	0.07934018	35.16	0	100.33	-0.36678
0.063	0.0986296	35.576	0	100.33	-0.36809
0.07	0.1206195	36.604	-0.001787	100.38	-0.42003
0.078	0.1303195	36.415	-2.26E-05	100.33	-0.36875
0.086	0.1462469	37.372	0	100.49	-0.52485
0.094	0.1654432	37.7	0	100.44	-0.47326
0.102	0.1877144	38.537	0	100.59	-0.63101
0.109	0.1877144	38.322	-0.0018094	100.38	-0.42067
0.117	0.1971913	38.631	-1.12E-05	100.44	-0.47293
0.125	0.1972592	38.505	-0.0017981	100.33	-0.36875
0.133	0.2036224	38.88	-0.0018094	100.38	-0.4206

Response:



Here's a Python script that generates a JSON file containing summary statistics for the table you provided:

```
import json
import pandas as pd

# Read the table into a Pandas DataFrame
df = pd.read_csv("table.csv")

# Compute summary statistics for each column
stats = df.describe().to_dict()

# Save the summary statistics to a JSON file
with open("summary_stats.json", "w") as f:
    json.dump(stats, f)
```

Note that in this script you need to have your table in a .csv file format with the name table.csv and run the script on your machine.

It will generate a file summary_stats.json in your local directory which will have all the summary statistics like count, mean, std, min, max and quartiles for each column.

Extracting data from XLSX file

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PRJ-3758 | Select CPT-Based Liquefaction Case Histories from the 2001 Nisqually, Washington, Earthquake

Download Dataset

Author(s) [Maurer, Brett](#); [Geyin, Mertcan](#); [Rasanen, Ryan](#)

Data Type **Database**

Natural Hazard Type **Earthquake**

Date of Publication **11-25-2022**

Awards **CAREER: Decoding the Enigmas of U.S. Seismic Hazard Via Multi-Scale, Multi-Physics Approaches to Paleoliquefaction Analysis | CMMI-1751216**

Nisqually, Earthquake, Liquefaction, Case History, Cone Penetration Test

DOI [Citation](#) **10.17603/ds2-nsf8-7944**

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Description | While soil liquefaction is common in earthquakes, the case history data required to train and test state-of-practice prediction models remains comparatively scarce, owing to the breadth and expense of data that comprise a single case history. The 2001 Nisqually, Washington, earthquake, for example, occurred in a metropolitan region and induced damaging liquefaction in the urban cores of Seattle and Olympia, yet case history data has not previously been published. Accordingly, we compile 24 cone-penetration-test (CPT) case histories from free-field locations. The...

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PRJ-3758

<input checked="" type="checkbox"/> Name	Size
<input type="checkbox"/> Case History Data	--
Data	

File Preview: Blair Waterway-01.xlsx

File Metadata

File name	Blair Waterway-01.xlsx
File path	/PRJ-3758/Case History Data/Blair Waterway-01.xlsx
File size	68.8 kB
Last modified	11/25/22 12:34 PM

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Excel		Blair Waterway-01 - Saved		Search (Option + Q)															
File		Home		Insert		Draw		Page Layout		Formulas		Data		Review		View		Help	
12		B				A				General		Σ		Z		V			
1		X		Y		fx													
A		B		C		D		E		F		G		H					
1	depth_m	q_MPa	fs_MPa									Site Name	Blair Waterway-01						
2		0.974283717	0.026134834									Latitude	47.26486794						
3		0.01.974283717	0.026134834									Longitude	-122.3875596						
4		0.02.974283717	0.026134834									Geotech Report	Hart-Crowser & Associates, Inc. (15						
5		0.03.974283717	0.026134834									Test Date	2/9/1981						
6		0.04.974283717	0.026134834									Manifestation	No						
7		0.05.974283717	0.026134834									Water Table Depth (m)	3.05						
8		0.06.974283717	0.026134834									pre-drill (m)	0.18						
9		0.07.974283717	0.026134834									Conditional Median PGA (g)	0.104						
10		0.08.974283717	0.026134834									Conditional Standard Deviation	0.236						
11		0.09.974283717	0.026134834																
12		0.1.974283717	0.026134834																
13		0.11.974283717	0.026134834																
14		0.12.974283717	0.026134834																
15		0.13.974283717	0.026134834																

Trial 0: Extract metadata directly from text

K

Extract metadata from this file sample:

depth_m	q_MPa	fs_MPa	Site Name	Blair Waterway-01
0	0.974283717	0.026134834	Latitude	47.26486794
0.01	0.974283717	0.026134834	Longitude	-122.3875596
0.02	0.974283717	0.026134834	Geotech Report	Hart-Crowser & Associates, Inc. (1981)
0.03	0.974283717	0.026134834	Test Date	2/9/1981
0.04	0.974283717	0.026134834	Manifestation	No
0.05	0.974283717	0.026134834	Water Table Depth (m)	3.05
0.06	0.974283717	0.026134834	pre-drill (m)	0.18
0.07	0.974283717	0.026134834	Conditional Median PGA (g)	0.104
0.08	0.974283717	0.026134834	Conditional Standard Deviation	0.236
0.09	0.974283717	0.026134834		
0.1	0.974283717	0.026134834		
0.11	0.974283717	0.026134834		
0.12	0.974283717	0.026134834		



The file appears to be a tab-separated data file that contains various metadata about a site. The metadata includes:

- Site Name: Blair Waterway-01
- Latitude: 47.26486794
- Longitude: -122.3875596
- Geotech Report: Hart-Crowser & Associates, Inc. (1981)
- Test Date: 2/9/1981
- Manifestation: No
- Water Table Depth (m): 3.05

K

convert metadata to a JSON file



Here is an example of how the metadata from the file could be converted to a JSON file:

```
{
  "Site Name": "Blair Waterway-01",
  "Latitude": 47.26486794,
  "Longitude": -122.3875596,
  "Geotech Report": "Hart-Crowser & Associates, Inc. (1981)",
  "Test Date": "2/9/1981",
  "Manifestation": "No",
  "Water Table Depth (m)": 3.05,
  "pre-drill (m)": 0.18,
  "Conditional Median PGA (g)": 0.104,
  "Conditional Standard Deviation": 0.236
}
```

It's important to notice that JSON does not accept tab as separator and the data in the file may need to be preprocessed accordingly and also, JSON file format use double quotes instead of single quotes.

Trial 1: ChatGPT to write code

K

write a Python code to read an XLSX file to a text stream and using OpenAI text-davinci API to summarize the variables and the values in the xlsx file:



```
import openai_secret_manager
import pandas as pd
import requests

# Authenticate with OpenAI API
secrets = openai_secret_manager.get_secrets("openai")
api_key = secrets["api_key"]

# Read XLSX file to a Pandas dataframe
file_path = "path/to/file.xlsx"
df = pd.read_excel(file_path)

# Convert dataframe to a text stream
text_stream = df.to_string()

# Use OpenAI text-davinci API to summarize the variables and values in
the xlsx file
response = requests.post(
    "https://api.openai.com/v1/engines/davinci-codex/completions",
    json={
        "prompt": f"Please summarize the variables and values in the
following text:\n{text_stream}",
        "max_tokens": 128,
```

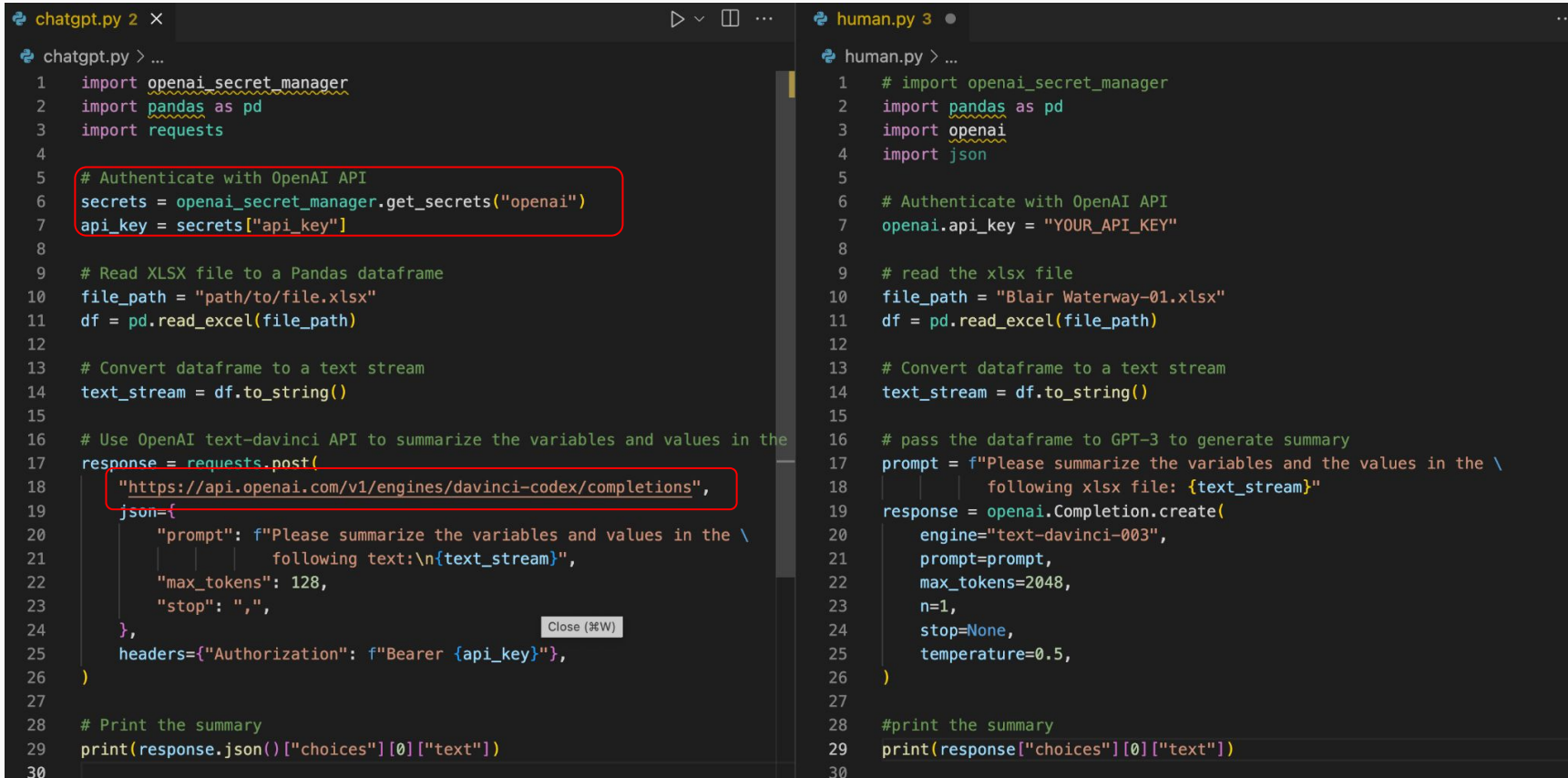
Copy code



Ask ChatGPT to write a code using OpenAI API to summarize XLSX file.

- ChatGPT only knows about Text-davinci-002 model as it was trained with data before 2021.
- It uses the package ``openai_secret_manager`` that didn't work.

Trial 2: human intervention in the code



```
chatgpt.py 2 x
chatgpt.py > ...
1 import openai_secret_manager
2 import pandas as pd
3 import requests
4
5 # Authenticate with OpenAI API
6 secrets = openai_secret_manager.get_secrets("openai")
7 api_key = secrets["api_key"]
8
9 # Read XLSX file to a Pandas dataframe
10 file_path = "path/to/file.xlsx"
11 df = pd.read_excel(file_path)
12
13 # Convert dataframe to a text stream
14 text_stream = df.to_string()
15
16 # Use OpenAI text-davinci API to summarize the variables and values in the
17 response = requests.post(
18     "https://api.openai.com/v1/engines/davinci-codex/completions",
19     json={
20         "prompt": f"Please summarize the variables and values in the \
21             following text:\n{text_stream}",
22         "max_tokens": 128,
23         "stop": ",",
24     },
25     headers={"Authorization": f"Bearer {api_key}"},
26 )
27
28 # Print the summary
29 print(response.json()["choices"][0]["text"])
30
```

```
human.py 3
human.py > ...
1 # import openai_secret_manager
2 import pandas as pd
3 import openai
4 import json
5
6 # Authenticate with OpenAI API
7 openai.api_key = "YOUR_API_KEY"
8
9 # read the xlsx file
10 file_path = "Blair Waterway-01.xlsx"
11 df = pd.read_excel(file_path)
12
13 # Convert dataframe to a text stream
14 text_stream = df.to_string()
15
16 # pass the dataframe to GPT-3 to generate summary
17 prompt = f"Please summarize the variables and the values in the \
18     following xlsx file: {text_stream}"
19 response = openai.Completion.create(
20     engine="text-davinci-003",
21     prompt=prompt,
22     max_tokens=2048,
23     n=1,
24     stop=None,
25     temperature=0.5,
26 )
27
28 #print the summary
29 print(response["choices"][0]["text"])
30
```

Trial 2: Human + AI code -> Output

```
(xai) krishna@kks32 metadata-extraction % python3 metadata.py
```

The xlsx file contains the following variables and values:

depth_m: 0.00-0.22

q_MPa: 0.974284-0.911027

fs_MPa: 0.026135-0.026977

Unnamed: 3: No values

Unnamed: 4: No values

Unnamed: 5: No values

Site Name: Blair Waterway-01

Latitude: 47.264868

Longitude: -122.38756

Geotech Report: Hart-Crowser & Associates, Inc. (1981)

Test Date: 1981-02-09 00:00:00

Manifestation: No

Water Table Depth (m): 3.05

pre-drill (m): 0.18

Conditional Median PGA (g): 0.104

Conditional Standard Deviation: 0.236

Trial 2b: Extend GPT to extract it as a JSON

K

Extract variables and values from the following text output as a JSON file:


depth_m: 0.00 - 0.22 (increments of 0.01)
q_MPa: 0.974284 - 0.911027 (increments of 0.05 - 0.06)
fs_MPa: 0.026135 - 0.026977 (increments of 0.001 - 0.002)
Unnamed: 3: Empty
Unnamed: 4: Empty
Unnamed: 5: Empty
Site Name: Blair Waterway-01
Latitude: 47.264868
Longitude: -122.38756
Geotech Report: Hart-Crowser & Associates, Inc. (1981)
Test Date: 1981-02-09 00:00:00
Manifestation: No
Water Table Depth (m): 3.05
pre-drill (m): 0.18
Conditional Median PGA (g): 0.104
Conditional Standard Deviation: 0.236

Generate output as JSON

```
1 {  
2   "depth_m": {  
3     "min": 0,  
4     "max": 0.22,  
5     "increment": 0.01  
6   },  
7   "q_MPa": {  
8     "min": 0.974284,  
9     "max": 0.911027,  
10    "increment": {  
11      "min": 0.05,  
12      "max": 0.06  
13    }  
14  },  
15  "fs_MPa": {  
16    "min": 0.026135,  
17    "max": 0.026977,  
18    "increment": {  
19      "min": 0.001,  
20      "max": 0.002  
21    }  
22  },  
23  "Unnamed: 3": "Empty",  
24  "Unnamed: 4": "Empty",  
25  "Unnamed: 5": "Empty",  
26  "Site Name": "Blair Waterway-01",  
27  "Latitude": 47.264868,  
28  "Longitude": -122.38756,  
29  "Geotech Report": "Hart-Crowser & Associates, Inc. (1981)",  
30  "Test Date": "1981-02-09 00:00:00",  
31  "Manifestation": "No",  
32  "Water Table Depth (m)": 3.05,  
33  "pre-drill (m)": 0.18,  
34  "Conditional Median PGA (g)": 0.104,  
35  "Conditional Standard Deviation": 0.236  
36 }
```

Trial 2c: Can we write a code to convert to JSON?

K

Write a generic python code to extract variables and values and output as a JSON file using the following text stream only as an example: 

The xlsx file contains the following variables and values:

depth_m: 0.00-0.22

q_MPa: 0.974284-0.911027

fs_MPa: 0.026135-0.026977

Unnamed: 3: No values

Unnamed: 4: No values

Unnamed: 5: No values

Site Name: Blair Waterway-01

Latitude: 47.264868

Longitude: -122.38756

Geotech Report: Hart-Crowser & Associates, Inc. (1981)

Test Date: 1981-02-09 00:00:00

Manifestation: No

Water Table Depth (m): 3.05

pre-drill (m): 0.18

Conditional Median PGA (g): 0.104

Conditional Standard Deviation: 0.236

Trial 2c: Text to JSON output

```
chatgpt.py > ...
1  import json
2
3  data = {}
4
5  text = ""The.xlsx file contains the following variables and values: \n depth_m: 0.00-0.
6  q_MPa: 0.974284-0.911027 \n fs_MPa: 0.026135-0.026977 \n Unnamed: 3: No values \n Unr
7  Site Name: Blair Waterway-01 \n Latitude: 47.264868 \n Longitude: -122.38756
8  Geotech Report: Hart-Crowser & Associates, Inc. (1981) \n Test Date: 1981-02-09 00:00:0
9  Manifestation: No \n Water Table Depth (m): 3.05 \n pre-drill (m): 0.18
10 Conditional Median PGA (g): 0.104 \n Conditional Standard Deviation: 0.236""
11
12 lines = text.split("\n")
13 for line in lines:
14     parts = line.split(":")
15     variable = parts[0].strip()
16     value = parts[1].strip()
17     data[variable] = value
18
19 with open("data.json", "w") as f:
20     json.dump(data, f)
21
```

The code above will take the text stream as input and extract the variables and values from it. It then stores them in a dictionary, and finally writes that dictionary to a JSON file named "data.json". The resulting JSON file will have the following format:

```
Copy code
{
  "depth_m": "0.00-0.22",
  "q_MPa": "0.974284-0.911027",
  "fs_MPa": "0.026135-0.026977",
  "Unnamed: 3": "No values",
  "Unnamed: 4": "No values",
  "Unnamed: 5": "No values",
  "Site Name": "Blair Waterway-01",
  "Latitude": "47.264868",
  "Longitude": "-122.38756",
  "Geotech Report": "Hart-Crowser & Associates, Inc. (1981)",
  "Test Date": "1981-02-09 00:00:00",
  "Manifestation": "No",
  "Water Table Depth (m)": "3.05",
  "pre-drill (m)": "0.18",
  "Conditional Median PGA (g)": "0.104",
  "Conditional Standard Deviation": "0.236"
}
```

Building a semantic search engine from JSON

K

write a python code to import JSON file to Neo4j



Here is an example of a Python script that imports a JSON file into a Neo4j database using the `py2neo` library:



```
from py2neo import Graph, Node, Relationship
import json

# Connect to Neo4j
graph = Graph("bolt://localhost:7687", auth=("neo4j", "password"))

# Load json file
with open("path/to/file.json") as json_file:
    data = json.load(json_file)

# Iterate through json object and create nodes and relationships
for key, value in data.items():
    node = Node("Entity", name=key)
    graph.create(node)
    if type(value) == dict:
        for subkey, subvalue in value.items():
            subnode = Node("Entity", name=subkey)
            graph.create(subnode)
            rel = Relationship(node, "HAS", subnode)
            graph.create(rel)
    else:
        subnode = Node("Entity", name=value)
```

Copy code

```
else:
    subnode = Node("Entity", name=value)
    graph.create(subnode)
    rel = Relationship(node, "HAS", subnode)
    graph.create(rel)

print("Data imported successfully!")
```

The script uses the `Graph` class from `py2neo` to connect to a Neo4j database running on the localhost at port 7687. It loads the JSON file using the `json` package, then it iterates through the objects in the JSON file, creating `Node` and `Relationship` objects for each key-value pair. It creates the nodes and relationships in the Neo4j database using the `create` method.

It's important to notice that this script assumes that you have Neo4j installed locally and running on the default port 7687 and also, you will need to have `py2neo` package installed in your environment, if not, you can install it with the following command:

```
pip install py2neo
```

Copy code

Also, you may need to adjust the credentials and the connection url to match your Neo4j setup.

Challenges: GPT is a probabilistic language model

Outputs will be different for the same input!

Output 1:

```
(xai) krishna@kks32 metadata-extraction % python3 metadata.py
```

The xlsx file contains six columns and 23 rows. The first three columns are 'depth_m', 'q_MPa', and 'fs_MPa', which contain values of depth in meters, q in MPa, and fs in MPa, respectively. The remaining columns are unnamed and contain values of site name, latitude, longitude, geotech report, test date, manifestation, water table depth in meters, pre-drill in meters, conditional median PGA in g, and conditional standard deviation, respectively.

Output 2:

```
(xai) krishna@kks32 metadata-extraction % python3 metadata.py
```

The xlsx file contains the following variables and values:

```
depth_m: 0.00-0.22
q_MPa: 0.974284-0.911027
fs_MPa: 0.026135-0.026977
Unnamed: 3: No values
Unnamed: 4: No values
Unnamed: 5: No values
Site Name: Blair Waterway-01
Latitude: 47.264868
Longitude: -122.38756
Geotech Report: Hart-Crowser & Associates, Inc. (1981)
Test Date: 1981-02-09 00:00:00
Manifestation: No
Water Table Depth (m): 3.05
pre-drill (m): 0.18
Conditional Median PGA (g): 0.104
Conditional Standard Deviation: 0.236
```