Structured Output

- The output from an LLM is unstructured (Text)
- Structured output is output in well-defined data format (eg. JSON)

```
{
    "name": "John Doe",
    "email": "john@example.com"
}
```

 LLMs can have a conversation with other systems with the help of structured output



with_structured_outp ut () → Use this to generate structured output

LLMs who cannot generate structured output → Use **Output parsers**

⅓1. Structured Output

3 Ways:

- 1. Typed Dictionary
- 2. Pydantic
- 3. JSON Schema

1. Typed Dictionary

- You define the key, values and their type
 - Name: str

Age: int



It does not give you any error if you pass str instead of int.

• This is just a way to define a dictionary

from typing import TypedDict

class Person(TypedDict):

name: str age: int

• You define a dictionary

Problem Statement:

- We have reviews
- We want columns: summary & sentiments

from langchain_groq import ChatGroq from dotenv import load_dotenv from typing import TypedDict load_dotenv()

model= ChatGroq(model="gemma2-9b-it")

Create a class:

Create a class

```
class Review (TypedDict):
summary: str
sentiment: str
```

• The class inherits **TypedDict**

Pass the dictionary into the model:

```
structured_model = model.with_structured_output(Review)
```

Invoke the model:

""")

```
structured_model.invoke(""""

I recently upgraded to the Samsung Galaxy S24 Ultra, and I must s

The S-Pen integration is a great touch for note-taking and quick sketches, thoug

However, the weight and size make it a bit uncomfortable for one-handed use. Al

Pros:
Insanely powerful processor (great for gaming and productivity)

Stunning 200MP camera with incredible zoom capabilities

Long battery life with fast charging
```

```
{'sentiment': 'mixed',
'summary': "The Samsung Galaxy S24 Ultra is a powerful phone with a great camera, but it's also heavy, expensive, and has bloatware."}
```

You can add a rating column as well:

S-Pen support is unique and useful

```
# Create a class

class Review (TypedDict):
  summary: str
```

```
sentiment: str
rating: int

structured_model = model.with_structured_output(Review)

structured_model.invoke(""""

Good mobile but facing heating issue 

""")
```

```
{'rating': 3,
  'sentiment': 'negative',
  'summary': 'Good mobile but facing heating issue'}
```



This predicts the exact ratings.

Annotation, Literal, Optional:

from typing import Annotated, Literal, Optional

 The model could not know what to do, in this case, we give it specific instructions.

```
class Review (TypedDict):
    key_themes: Annotated[list[str], "Write down all the key themes discussed in the summary: Annotated[str, "A brief summary of the review"]
    sentiment: Annotated[Literal["positive", "negative", "mixed"], "Return sentiment pros: Annotated[Optional[list[str]], "Write down all the pros inside a list"]
    cons: Annotated[Optional[list[str]], "Write down all the cons inside a list"]
```

```
rating: int

structured_model = model.with_structured_output(Review)

structured_model.invoke('''

Very ok quality product. Very tasty and crispy. Good packing. Please keep it up.

'''
)
```

```
{'cons': [],
  'pros': ['Very ok quality product', 'Very tasty and crispy', 'Good packing'],
  'rating': 5,
  'sentiment': 'positive',
  'summary': "A positive review praising the product's quality, taste, and packaging."}
```

Literal["positive", "negative", "mixed"]: You can specify the values for a specific key

Optional [list[str] → You can set an optional key

Pass Multiple reviews:

```
from typing import Annotated, Literal, Optional

class Review (TypedDict):
    key_themes: Annotated[list[str], "Write down all the key themes discussed i
n the review in a list"]
    summary: Annotated[str, "A brief summary of the review"]
    sentiment: Annotated[Literal["positive", "negative", "mixed"], "Return sentim
ent of the review either negative, positive or neutral"]
    pros: Annotated[Optional[list[str]], "Write down all the pros inside a list"]
    cons: Annotated[Optional[list[str]], "Write down all the cons inside a list"]
    rating: int

Reviews = ["Completely waste , fully disappointed about quality , strongly non
reffer , please don't purchase anyone , please visit local offline stores and pur
```

chase , 1299 and 4 days weiting time all waste", "These pistachios are of top quality—crunchy, fresh, and packed with flavor. Perfect as a healthy snack or to add to desserts and dishes. The packaging keeps them fresh for longer, an d the taste is absolutely delightful. A great value for the price! Highly recomm end!"]

```
structured_model = model.with_structured_output(Review)
for r in Reviews:
    data1= structured_model.invoke(r)
    print(data1)
```

```
{'rating': 1, 'sentiment': 'negative', 'summary': 'The product is completely disapp
{'cons': [], 'key_themes': ['quality', 'taste', 'freshness', 'versatility', 'value'
```



Con: You cannot do data validation

2. Pydantic

pip install pydantic

What is Pydantic?

Pydantic is a Python library that helps you **define data shapes** (schemas) using simple Python classes.

You tell Pydantic what kind of data you expect, and it will:

- Validate the data
- V Enforce the types
- * Throw errors if the data is wrong

```
from pydantic import BaseModel

class Student(BaseModel):
    name: str

new_student= {'name': 'John'}

student1= Student(**new_student)

print(student1)
```

name='John'

**new_student Unpacks {'name': 'John'} into name='John'.

Why Use *?

The ** operator allows you to dynamically pass fields to a class or function when the values are already stored in a dictionary. It's especially useful when:

- You don't know the keys ahead of time.
- You're working with dynamic data, like parsing JSON.

If we replace John with a number → It will throw an error

```
from pydantic import BaseModel

class Student(BaseModel):
    name: str

new_student= {'name': 25}

student1= Student(**new_student)
```

```
print(student1)
```

```
ValidationError: 1 validation error for Student

name

Input should be a valid string [type=string_type, input_value=25, input_type=int]

For further information visit <a href="https://errors.pydantic.dev/2.11/v/string_type">https://errors.pydantic.dev/2.11/v/string_type</a>
```

· It's expecting a string

Set default value:

```
from pydantic import BaseModel

class Student(BaseModel):
    name: str = 'John'

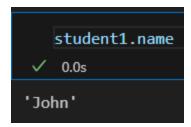
new_student= {}

student1= Student(**new_student)

print(student1)
```

name='John'

You can fetch the object with .name as well:



Optional Fields:



If no value → value=None

```
from pydantic import BaseModel
from typing import Optional

class Student(BaseModel):
    name: str = 'John'
    age: Optional[int]=None

new_student= {}

student1= Student(**new_student)

print(student1)
```

name='John' age=None

```
new_student= {'age': 4}
student1= Student(**new_student)
print(student1)
```

name='John' age=4

Type Coercing:

Even if you pass number like 45', Pydantic will convert it into Int

```
new_student= {'age': '45'}
student1= Student(**new_student)
print(student1)
```

name='John' age=45

Validate Emails

```
pip install pydantic[email]
```

from pydantic import BaseModel, EmailStr

EmailStr

• It's a built-in datatype inside Pydantic that validates email

```
from pydantic import BaseModel, EmailStr
from typing import Optional

class Student(BaseModel):
    name: str = 'John'
    age: Optional[int]=None
    email: EmailStr

new_student= {'age': 45, 'email': "abc@gmail.com"}

student1= Student(**new_student)
student1
```

Student(name='John', age=45, email='abc@gmail.com')

XWrong email:

```
from pydantic import BaseModel, EmailStr
from typing import Optional

class Student(BaseModel):
    name: str = 'John'
    age: Optional[int]=None
    email: EmailStr

new_student= {'age': 45, 'email': "abc"}

student1= Student(**new_student)
student1
```

```
ValidationError: 1 validation error for Student
email
value is not a valid email address: An email address must have an @-sign. [type=value_error, input_value='abc', input_type=str]
```

Field() Function

from pydantic import Field

Purpose	Example
Set default values	Field(default ="India")
Add description/help text	Field(, description = "User name")
Set constraints	Field(gt =0) (greater than 0)
Make things optional/required	Field(default=None) or

Field Validations (Constraints)

Constraint	Meaning Example		
gt=0	Greater than 0	Field(gt=0)	
It=100	Less than 100	Field(It=100)	

Constraint	Meaning Example		
ge=1	Greater than or equal to 1	Field(ge=1)	
le=10	Less than or equal to 10	Field(le=10)	
min_length=5	Minimum length for a string	Field(min_length=5)	
max_length=50	Maximum length for a string	Field(max_length=50)	
regex=	Pattern matching using Regex	Field(regex="^[a-z]+\$")	

So basically:

Field() gives more control and meaning to your fields.

```
from pydantic import BaseModel, EmailStr, Field
from typing import Optional

class Student(BaseModel):

name: str = 'john'
age: Optional[int] = None
email: EmailStr
cgpa: float = Field(gt=0, lt=10, default=5, description='A decimal value repr
esenting the cgpa of the student')

new_student = {'age':'32', 'email':'abc@gmail.com'}

student1 = Student(**new_student)
student1
```

```
Student(name='john', age=32, email='abc@gmail.com', cgpa=5)
```

• This is Pydantic object

Convert this into python dictionary:

```
student_dict = dict(student1)
student_dict
```

```
{'name': 'john', 'age': 32, 'email': 'abc@gmail.com', 'cgpa': 6.3}
```

Convert to JSON:

```
#Convert to JSON:
student1_json= student1.model_dump_json()
student1_json
```

```
'{"name":"john","age":32,"email":"abc@gmail.com","cgpa":6.3}'
```

Pydantic + with_structured_output (Mostly Used)

from langchain_groq import ChatGroq from dotenv import load_dotenv from typing import TypedDict, Annotated, Optional, Literal from pydantic import BaseModel, Field

Class will inherit the BaseModel from Pydantic

```
class Review(BaseModel):
```

```
key_themes : list[str] = Field(description= "Write down all the key themes discussion of the review") sentiment: Literal['Positive', 'Negative', "Mixed"] = Field(description= "Return of the review") pros: Optional[list[str]] = Field(default=None, description="Write down all the cons: O
```

```
name: Optional[str] = Field(default=None, description="Write the name of the
#default=None required for Optional value

structured_model = model.with_structured_output(Review)
```

name, pros, cons → Optional

Invoke:

```
result = structured_model.invoke("""I recently upgraded to the Samsung Galaxy:

The S-Pen integration is a great touch for note-taking and quick sketches, thoug

However, the weight and size make it a bit uncomfortable for one-handed use. A

Pros:
Insanely powerful processor (great for gaming and productivity)

Stunning 200MP camera with incredible zoom capabilities

Long battery life with fast charging

S-Pen support is unique and useful

Review by Randy Orton

""")

print(result)
```

Output:

key_themes=['Performance', 'Camera Quality', 'Battery Life', 'Size and Weight', 'Bloatware', 'Price']

summary='The Samsung Galaxy S24 Ultra is a powerful smartphone with a fa ntastic camera and long battery life, but its size, weight, bloatware, and high p

```
rice are drawbacks.'

sentiment='Mixed'

pros=['Insanely powerful processor (great for gaming and productivity)', 'Stun ning 200MP camera with incredible zoom capabilities', 'Long battery life with f ast charging', 'S-Pen support is unique and useful']

cons=['The weight and size make it a bit uncomfortable for one-handed use.', 'Samsung's One UI still comes with bloatware—why do I need five different Sa msung apps for things Google already provides?', 'The $1,300 price tag is also a hard pill to swallow.']

name='Randy Orton'
```

3. JSON Schema

• Used when project is made in multiple languages

JSON Schema is a way to describe the structure and rules for a JSON document.

It's like saying:

"This JSON must have these fields, and those fields must follow these rules."

```
"title": "student",
  "description": "schema about students",
  "type": "object",
  "properties":{
     "name":"string",
     "age":"integer"
```

```
},
  "required":["name"] #Name of the student must be present
}
```

- title → Title of schema
- description → Optional
- type → datatype of schema
- properties → You write all the attributes in this
- required → These fields must be present.

Pydantic Schema → JSON Schema

Pydantic Schema:

```
class Review(BaseModel):
    key_themes : list[str] = Field(description= "Write down all the key themes discussion of the review")
    summary: str = Field(description= "A brief summary of the review")
    sentiment: Literal['Positive', 'Negative', "Mixed"] = Field(description= "Return of the pros: Optional[list[str]] = Field(default=None, description="Write down all the name: Optional[str] = Field(default=None, description="Write the name of the name)
```

JSON Schema:

```
json_schema = {
  "title": "Review",
  "type": "object",
  "properties": {
    "key_themes": {
      "type": "array", #Array = List
      "items": {
      "type": "string" # Type of items inside the list= string
```

```
},
   "description": "Write down all the key themes discussed in the review in a list
  "summary": {
   "type": "string",
   "description": "A brief summary of the review"
  },
  "sentiment": {
   "type": "string",
   "enum": ["pos", "neg"], #"enum"= Literal
   "description": "Return sentiment of the review either negative, positive or neu
  },
  "pros": {
   "type": ["array", "null"],
   "items": {
    "type": "string"
   },
   "description": "Write down all the pros inside a list"
  },
  "cons": {
   "type": ["array", "null"],
   "items": {
    "type": "string"
   },
   "description": "Write down all the cons inside a list"
  },
  "name": {
   "type": ["string", "null"],
   "description": "Write the name of the reviewer"
  }
 },
 "required": ["key_themes", "summary", "sentiment"]
}
```

array = List

```
{
  "key_themes": {
    "type": "array", #Array = List
    "items": {
        "type": "string" # Type of items inside the list= string
    }
}
```

- "type": "array" → key_themes is a list
 - "type": "string" → Type of items inside the list= string
 - In short, list[string]
- "description" → Optional

```
enum" = Literal
```

"type": ["array", "null"] → Null=Optional Field

Remaining steps are exactly same:

```
structured_model = model.with_structured_output(json_schema)

result = structured_model.invoke("""I recently upgraded to the Samsung Galaxy

The S-Pen integration is a great touch for note-taking and quick sketches, thoug
```

However, the weight and size make it a bit uncomfortable for one-handed use. Al

Pros:

Insanely powerful processor (great for gaming and productivity)
Stunning 200MP camera with incredible zoom capabilities
Long battery life with fast charging
S-Pen support is unique and useful

```
Review by Randy Orton """)
```

print(result)

Output:

{'cons': ['The weight and size make it a bit uncomfortable for one-handed us e.', 'Samsung's One UI still comes with bloatware—why do I need five different Samsung apps for things Google already provides?', 'The \$1,300 price tag is a Iso a hard pill to swallow.'],

'key_themes': ['Performance', 'Camera', 'Battery Life', 'S-Pen', 'Bloatware', 'Price'],

'name': 'Randy Orton',

'pros': ['Insanely powerful processor (great for gaming and productivity)', 'Stu nning 200MP camera with incredible zoom capabilities', 'Long battery life with fast charging', 'S-Pen support is unique and useful'],

'sentiment': 'pos',

'summary': 'The Samsung Galaxy S24 Ultra is a powerful and feature-rich pho ne with a stunning camera and long battery life, but its size, bloatware, and hi gh price are drawbacks.'}



The type you get is a dictionary.

JSON Schema Types You Can Use

Туре	Description	Example	
string	Text	"Alice"	
number	Any number (int or float)	10.5	
integer	Whole numbers	25	
boolean	True or False	true	
array	List of items	[1, 2, 3]	
object	JSON object (dictionary)	{ "key": "value" }	
null	Empty / null value	null	

Convert Pydantic to JSON Schema

Pydantic models automatically generate JSON Schema.

Class.model_json_schema()

Ex.

class Review(BaseModel):

key_themes : list[str] = Field(description= "Write down all the key themes di scussed in the review in a list")

summary: str = Field(description = "A brief summary of the review")

sentiment: Literal['Positive', 'Negative', "Mixed"] = Field(description= "Return sentiment of the review either negative, positive or Mixed")

pros: Optional[list[str]] = Field(default=None, description="Write down all th
e pros inside a list")

cons: Optional[list[str]] = Field(default=None, description="Write down all t
he cons inside a list")

name: Optional[str] = Field(default=None, description="Write the name of the reviewer")

Review.model_json_schema()

Output:

```
{'properties': {'key_themes': {'description': 'Write down all the key themes dis
cussed in the review in a list',
  'items': {'type': 'string'},
  'title': 'Key Themes',
  'type': 'array'},
 'summary': {'description': 'A brief summary of the review',
  'title': 'Summary',
  'type': 'string'},
 'sentiment': {'description': 'Return sentiment of the review either negative, p
ositive or Mixed',
  'enum': ['Positive', 'Negative', 'Mixed'],
  'title': 'Sentiment',
  'type': 'string'},
 'pros': {'anyOf': [{'items': {'type': 'string'}, 'type': 'array'},
  {'type': 'null'}],
  'default': None,
  'description': 'Write down all the pros inside a list',
  'title': 'Pros'},
 'cons': {'anyOf': [{'items': {'type': 'string'}, 'type': 'array'},
  {'type': 'null'}],
  'default': None,
 'description': 'Write down all the cons inside a list',
  'title': 'Cons'},
 'name': {'anyOf': [{'type': 'string'}, {'type': 'null'}],
  'default': None,
  'description': 'Write the name of the reviewer',
  'title': 'Name'}},
'required': ['key_themes', 'summary', 'sentiment'],
'title': 'Review',
'type': 'object'}
```

• Type = Dictionary

When to use what?

Use TypedDict if:

- You only need type hints (basic structure enforcement).
- You don't need validation (e.g., checking numbers are positive).
- You trust the LLM to return correct data.

Use Pydantic if:

- You need data validation (e.g., sentiment must be "positive", "neutral", or "negative").
- You need default values if the LLM misses fields.
- You want automatic type conversion (e.g., "100" → 100).

Use JSON Schema if:

- You don't want to import extra Python libraries (Pydantic).
- You need validation but don't need Python objects.
- You want to define structure in a standard JSON format.

Feature	TypedDict 🗸	Pydantic 🚀	JSON Schema 🜍
Basic structure	✓	~	~
Type enforcement	✓	~	✓
Data validation	×	~	✓
Default values	×	~	×
Automatic conversion	×	~	×
Cross-language compatibility	×	X	~