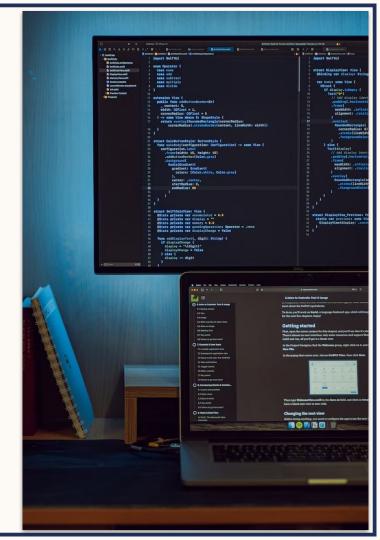
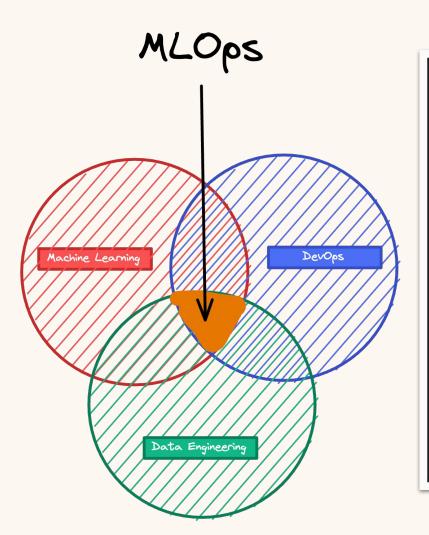


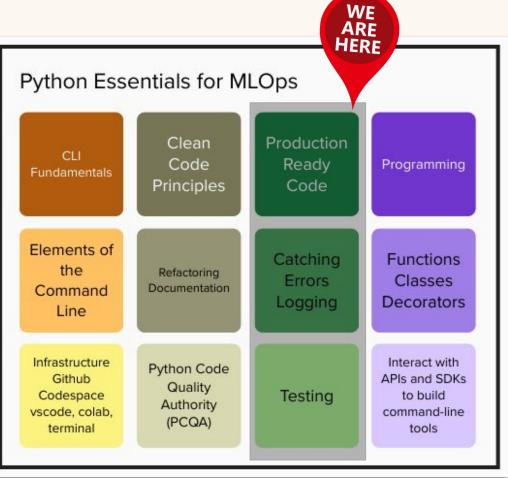
Production Ready Code

Handling Errors, Writing Tests and Logs

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Failed to load resource: net::ERR_BLOCKED_BY_AD Failed to load resource: net::ERR_BLOCKED_BY_AD Failed to load resource: net::ERR_BLOCKED_BY_AD Failed to load resource: net::ERR BLOCKED BY AD Failed to load resource: net::ERR BLOCKED BY AD

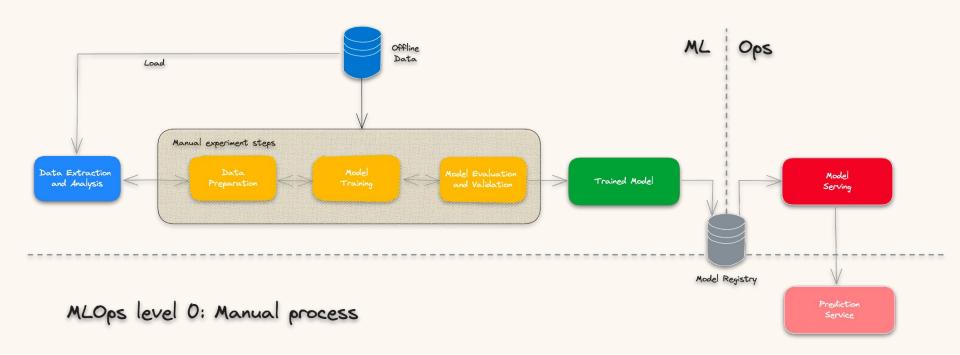
Failed to load resource: net::ERR BLOCKED BY AD

w Failed to load resource: Net::Enn_vtvtntv_r'_-'

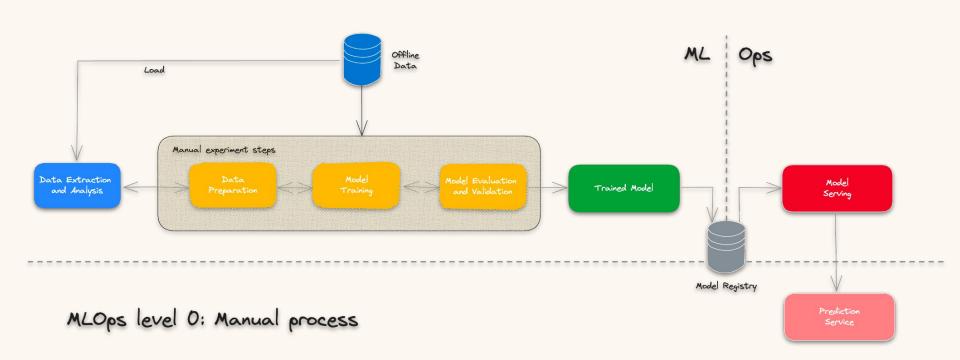
* Failed to load resource: net::ERR_BLOCKED_BY_AD

Why is Error Handling Essential in MLOps?

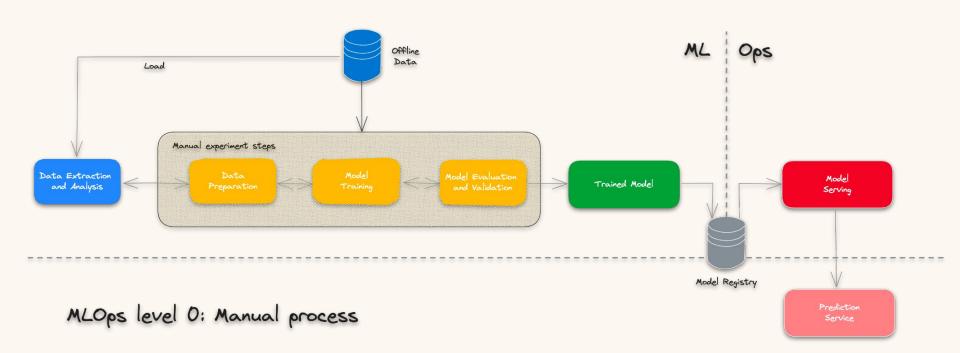
Data Loading Failures: Imagine an MLOps pipeline that trains a new model every day based on newly ingested data. One day, the data source changes its format without notice, and the pipeline fails to handle this change gracefully. As a result, no model is trained for that day, impacting various downstream applications.



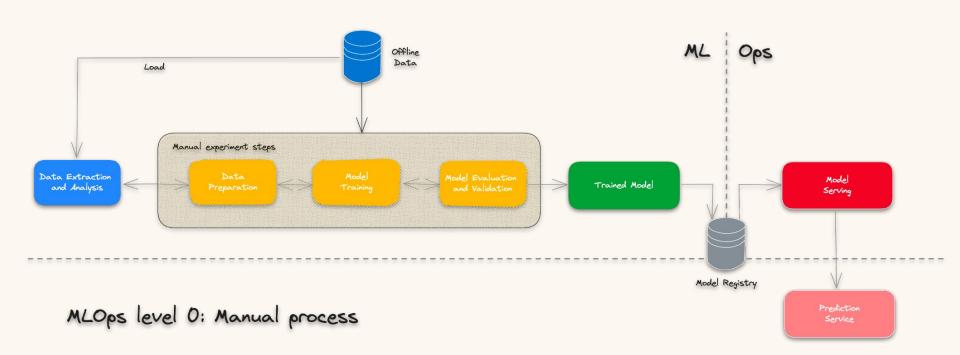
API Rate Limiting: Suppose your MLOps pipeline pulls data from an external API. If the API imposes rate limiting and your system doesn't handle this exception, the pipeline might crash or get stuck in an infinite loop, causing a series of failures downstream.



Resource Exhaustion: Imagine a machine learning model that's part of a recommendation engine for an e-commerce website. Poor error handling results in memory leaks, eventually crashing the recommendation service during peak sales hours, leading to significant loss of revenue.



Timeout Errors: In a complex MLOps setting, various services might depend on each other. Poor handling of timeout errors can cascade, causing a failure in multiple dependent systems.



```
try:
    x = int(input("Enter a number: "))
    y = int(input("Enter another number: "))
    result = x / y
except ZeroDivisionError:
    print("You can't divide by zero!")
except ValueError:
    print("Invalid input. Please enter a number.")
finally:
    print("This will execute no matter what.")
```

What is the output to?

$$x = 4$$
 and $y = 2$ $x = 4$ and $y = 0$ $x = "a"$ and $y = 2$

and to:

and?

```
def divide numbers(x, y):
    try:
        result = x / y
    except ZeroDivisionError:
        print("You can't divide by zero!")
    except TypeError:
        print("Both arguments must be numbers!")
    else:
        print(f"The result is: {result}")
    finally:
        print("This will execute no matter what.")
# Test the function
divide numbers(10, 2)
divide numbers(10, 0)
divide numbers(10, "a")
```

```
import pandas as pd
def filter and save csv(input file path, output file path):
    temp df = None # Initialize variable to hold the temporary DataFrame
    try:
        # Attempt to read the CSV into a DataFrame
        temp df = pd.read csv(input file path)
        # Filter rows where the value in the 'age' column is greater than 21
        filtered df = temp df[temp df['age'] > 21]
        # Save the filtered DataFrame to a new CSV file
        filtered df.to csv(output file path, index=False)
        print(f"Filtered data saved to {output file path}")
    except FileNotFoundError:
        print("The input file was not found.")
    except KeyError:
        print("The required column does not exist.")
    finally:
        # Always run this block to delete the temporary DataFrame and free up memory
        print("Cleaning up resources.")
        if temp df is not None:
            del temp df
```

Custom Exception

```
# Define a custom exception
class InvalidAgeException(Exception):
    pass
# Using the custom exception
try:
    age = int(input("Enter your age: "))
    if age < 0:
        raise InvalidAgeException("Age cannot be negative")
except InvalidAgeException as e:
    print(e)
```

Custom Exception

```
import pandas as pd
# Custom Exception for DataFrame Validation
class DataFrameValidationException(Exception):
   def init (self, message, missing columns=None, insufficient data columns=None):
       super(). init (message)
       self.missing columns = missing columns
       self.insufficient data columns = insufficient data columns
# Function to Validate DataFrame
def validate dataframe(df, required columns, min non null entries):
   missing columns = [col for col in required columns if col not in df.columns]
   insufficient data columns = [col for col in required columns if df[col].count() < min non null entries]
   if missing columns or insufficient data columns:
       raise DataFrameValidationException(
            "DataFrame validation failed.",
           missing columns=missing columns,
           insufficient data columns=insufficient data columns
```

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8. Errors and Exceptions

Until now error messages haven't been more than mentioned, but if you have tried out the examples you have probably seen some. There are (at least) two distinguishable kinds of errors: *syntax errors* and *exceptions*.

8.1. Syntax Errors

Syntax errors, also known as parsing errors, are perhaps the most common kind of complaint you get while you are still learning Python:

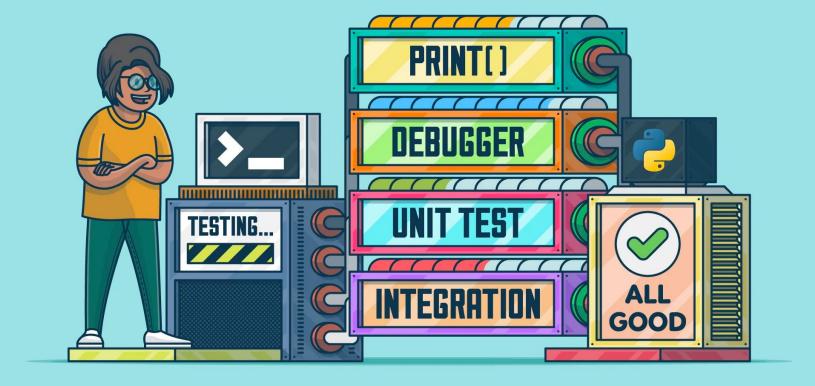
```
>>> while True print('Hello world')
File "<stdin>", line 1
while True print('Hello world')

SyntaxError: invalid syntax
```

The parser repeats the offending line and displays a little 'arrow' pointing at the earliest point in the line where the error was detected. The error is caused by (or at least detected at) the token *preceding* the arrow: in the example, the error is detected at the function print(), since a colon (':') is missing before it. File name and line number are printed so you know where to look in case the input came from a script.

8.2. Exceptions

Even if a statement or expression is syntactically correct, it may cause an error when an attempt is made to execute it. Errors detected during execution are called *exceptions* and are not unconditionally fatal: you will soon learn how to handle them in Python programs. Most exceptions are not handled by programs, however, and result in error messages as shown here:

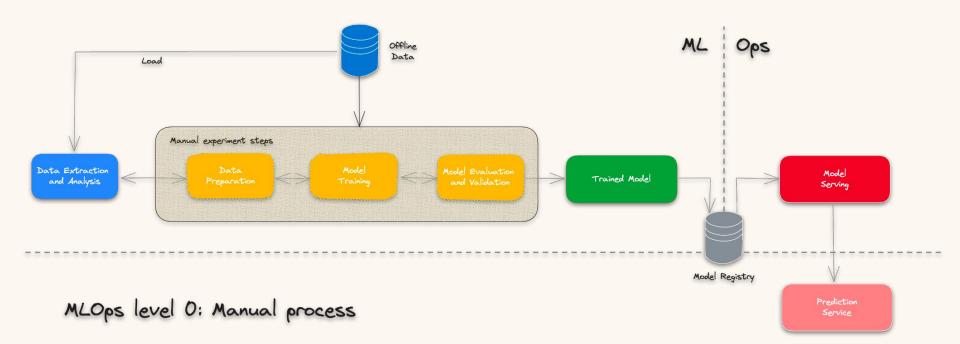


Real Rython

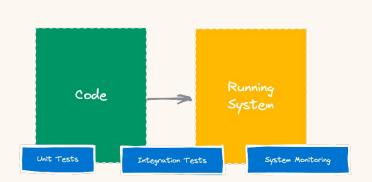
Debugging is hard, testing is easy (...)

- a) Rapid Application Development (RAD) tools
- b) Manual & Script-Driven
- c) Disconnection between ML and operation
- d) Infrequent release iterations
- e) No Continuous Integration (CI)
- f) No Continuous Delivery (CD)
- g) No Continuous Training (CT)

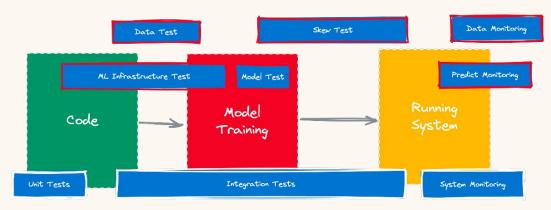
- h) Deployment refers to the prediction service (e.g a microservice with a REST API), rather than deploying the entire ML system
- i) Lack of active performance monitoring



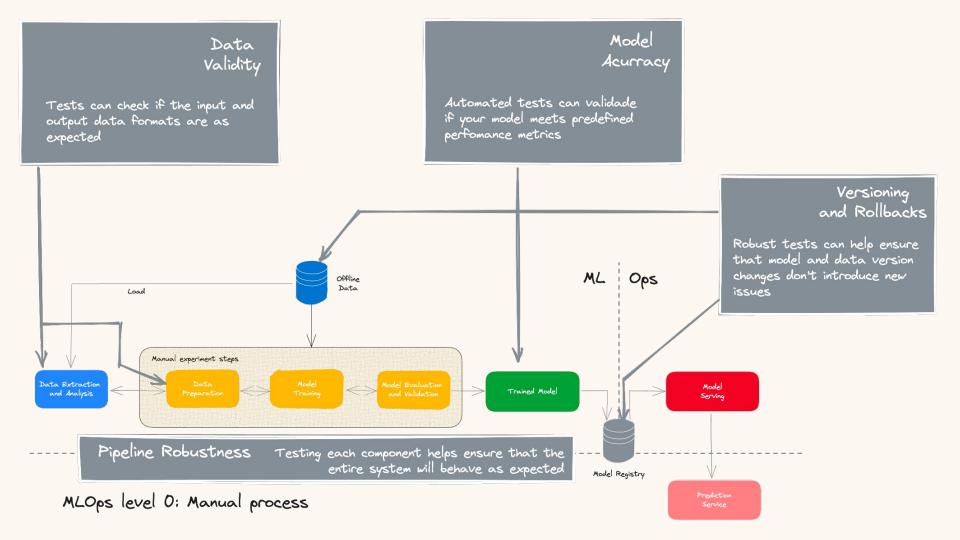
The complete development pipeline includes three essential components, data pipeline, ML model pipeline, and application pipeline. In accordance with this separation we distinguish three scopes for testing in ML systems: tests for features and data, tests for model development, and tests for ML infrastructure.



Traditional System Testing and Monitoring



ML-Based System Testing and Monitoring





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About pytest

pytest is a mature full-featured Python testing tool that helps you write better programs.

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Support pytest

pytest: helps you write better programs

The pytest framework makes it easy to write small, readable tests, and can scale to support complex functional testing for applications and libraries.

pytest requires: Python 3.7+ or PyPy3.

PyPI package name: pytest

A quick example

```
# content of test_sample.py
def inc(x):
    return x + 1

def test_answer():
    assert inc(3) == 5
```

Next Open Trainings

Germany / Remote

Also see previous talks and blogposts.

 Professional Testing with Python, via Python Academy, March 5th to 7th

2024 (3 day in-depth training), Leipzig,

```
To execute it:
$ pvtest
platform linux -- Python 3.x.y, pytest-7.x.y, pluggy-1.x.y
rootdir: /home/sweet/project
collected 1 item
test sample.py F
                                        [100%]
test_answer _____
  def test answer():
    assert inc(3) == 5
    assert 4 == 5
     + where 4 = inc(3)
test_sample.py:6: AssertionError
FAILED test_sample.py::test_answer - assert 4 == 5
```

```
# math_operations.py

def add(a, b):
    return a + b
```

```
# test_math_operations.py
from math_operations import add

def test_add():
    result = add(3, 5)
    assert result == 8
```

pytest test_math_operations.py

pip install pytest pytest-sugar

Pytest Fixtures

```
# ml_operations.py

def normalize(numbers):
    max_num = max(numbers)
    min_num = min(numbers)
    return [(x - min_num) / (max_num - min_num) for x in numbers]
```

```
# test_ml_operations.py

import pytest
from ml_operations import normalize

@pytest.fixture
def sample_data():
    return [5, 10, 15, 20, 25]

def test_normalize(sample_data):
    result = normalize(sample_data)
    assert result == [0.0, 0.25, 0.5, 0.75, 1.0]
```

pytest test_ml_operations.py

Pytest Fixtures: using a fixture across multiple test cases

```
# conftest.py
import pytest
@pytest.fixture
def sample_data():
    return [5, 10, 15, 20, 25]
```

```
# test_ml_operations2.py

from ml_operations import normalize

def test_normalize_again(sample_data):
    result = normalize(sample_data)
    assert result == [0.0, 0.25, 0.5, 0.75, 1.0]
```

pytest test_ml_operations2.py



Once the ML has been deployed, it need to be monitored

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logging — Logging facility for Python

Source code: Lib/logging/__init__.py

This module defines functions and classes which implement a flexible event logging system for applications and libraries.

The key benefit of having the logging API provided by a standard library module is that all Python modules can participate in logging, so your application log can include your own messages integrated with messages from third-party modules.

Important

This page contains the API reference information. For tutorial information and discussion of more advanced topics, see

- Basic Tutorial
- Advanced Tutorial
- Logging Cookbook

The simplest example:

```
>>> import logging
>>> logging.warning('Watch out!')
WARNING:root:Watch out!
```

The module provides a lot of functionality and flexibility. If you are unfamiliar with logging, the best way to get to grips with it is to view the tutorials (see the links above and on the right).

The basic classes defined by the module, together with their functions, are listed below.

- Loggers expose the interface that application code directly uses.
- Handlers send the log records (created by loggers) to the appropriate destination.
- Filters provide a finer grained facility for determining which log records to output.
- Formatters specify the layout of log records in the final output.