



# DAY 02 — Iterators & Generators

**Goal :** Understand *lazy evaluation* and write memory-efficient Python code

**Why it matters for ML:** datasets can be **millions of rows** — you cannot load everything into memory.

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## 1 What is Iteration

when you write:

```
for x in data:
```

```
...
```

Python internally does:

```
iterator = iter(data)
value = next(iterator)
```

iteration is pulling values one by one, not all at once.

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## 2 Iterator Basics

example:

```
nums = [1,2,3]
it = iter(nums)

print(next(it)) # 1
print(next(it)) # 2
print(next(it)) # 3
```

If you call "**next()**" again:

```
StopIteration
```

that's how Python knows the loop is finished.

### 3 Custom Iterator (Interview-Level Knowledge)

To create own iterator, implement:

- `__iter__()`
- `__next__()`

Example: Custom counter Iterator

```
class Counter:  
    def __init__(self, max_value):  
        self.current = 0  
        self.max_value = max_value  
  
    def __iter__(self):  
        return self  
  
    def __next__(self):  
        if self.current >= self.max_value:  
            raise StopIteration  
        self.current += 1  
        return self.current
```

Usage:

```
for num in Counter(s):  
    print(num)  
  
# output  
1 2 3 4 5
```

### ML Insight

Custom iterators are used for:

- Batch loading
- Streaming data
- Reading huge files line-by-line

## 4 Why Iterators Matter (Memory)

✗ Bad (loads everything):

```
data = [x * x for x in range(10_000_000)]
```

✓ Good (lazy):

```
data = (x * x for x in range(10_000_000))
```

second one use **almost zero extra memory**.

## 5 Generators

Generators are **simpler iterators**.

### Generator Function

```
def count_up_to(n):
    for i in range(1, n+1):
        yield i
```

Usage:

```
gen = count_up_to(5)

print(next(gen)) # 1
print(next(gen)) # 2
```

or:

```
for x in count_up_to(5):
    print(x)
```

### Key Rule:

- `return` → stops function
- `yield` → pauses function and remembers state

## 6 Generator Expressions (Pythonic & Powerful)

Like list comprehensions, but lazy.

```
squares = (x * x for x in range(10))
```

Instead of :

```
squares = [x * x for x in range(10)]
```

## 7 ML-Style Generator Example (Important)

Imagine a dataset too large to load.

```
def data_loader(file_path):
    with open(file_path) as f:
        for line in f:
            yield line.strip()
```

Usage:

```
for record in data_loader("data.txt"):
    process(record)
```

This is **exactly how ML pipelines work internally.**