Scalable and distributed applications in Python

Julien Danjou – 23th September 2017 – PyCon FR

Hello!

I am Julien Danjou

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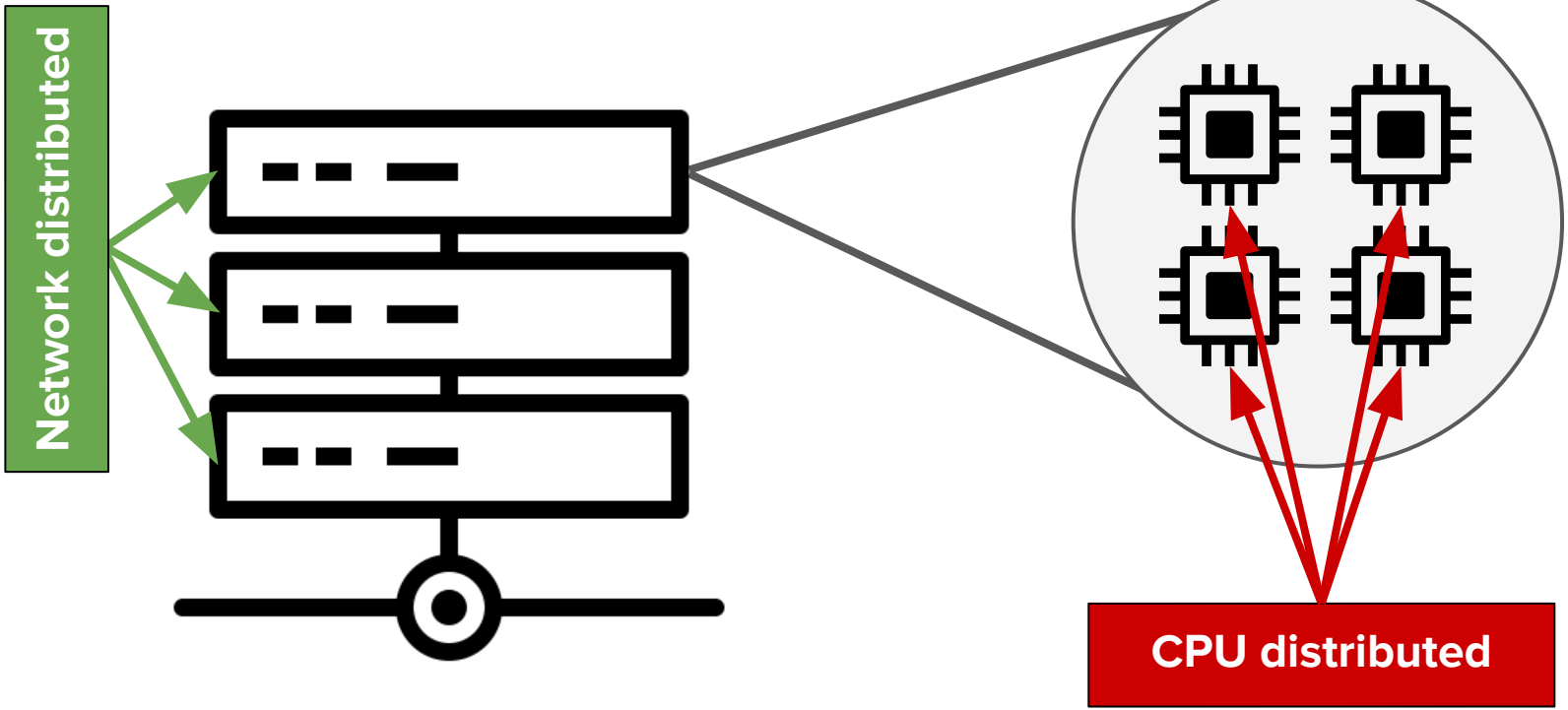
I **hack**, **create** and **contribute** to FOSS projects:

“the capability of a system, network, or process to handle a growing amount of work”

https://en.wikipedia.org/wiki/Scalability


FAN STATUS
FAN STATUS
UCS 2104XP
FAN STATUS
200-240V~
15.5 A Max 50/60 Hz
200-240V~
15.5 A Max 50/60 Hz
200-240V~
15.5 A Max 50/60 Hz
200-240V~
15.5 A Max 50/60 Hz
Use more resources
(efficiently)



Distributed systems

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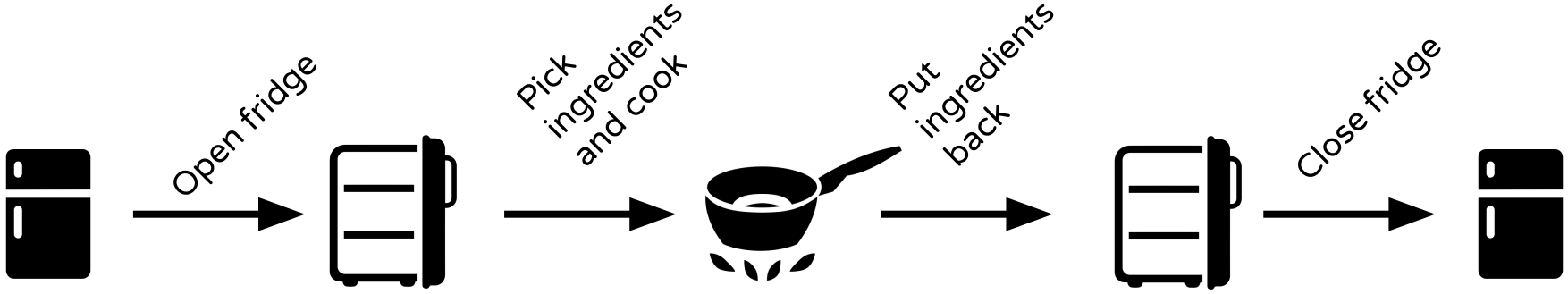
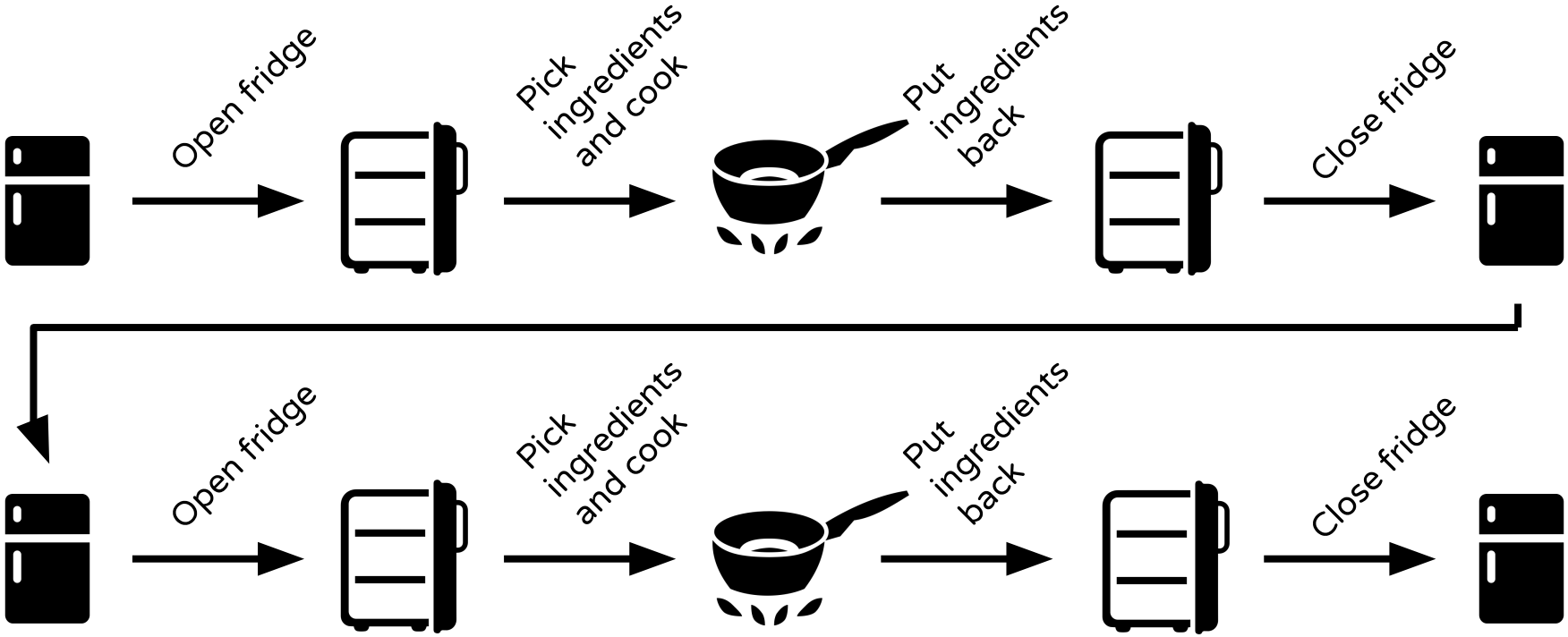
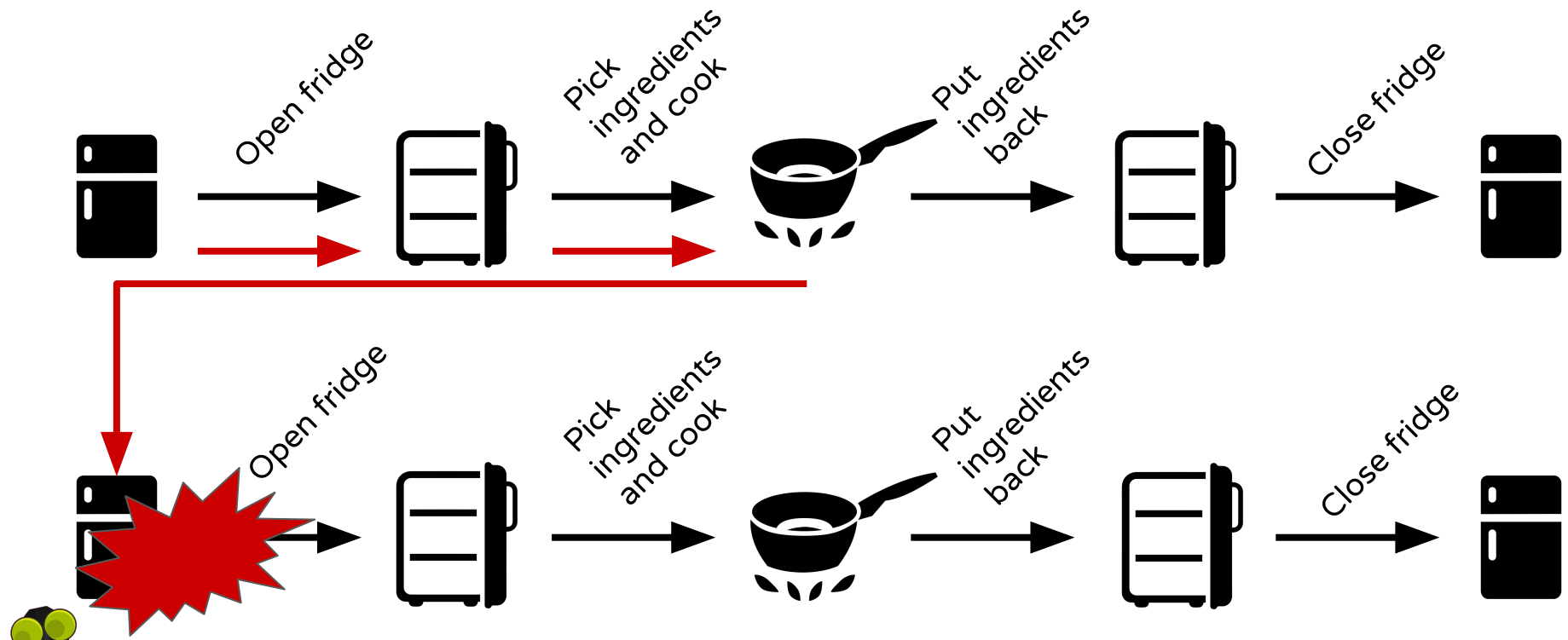
Concurrency



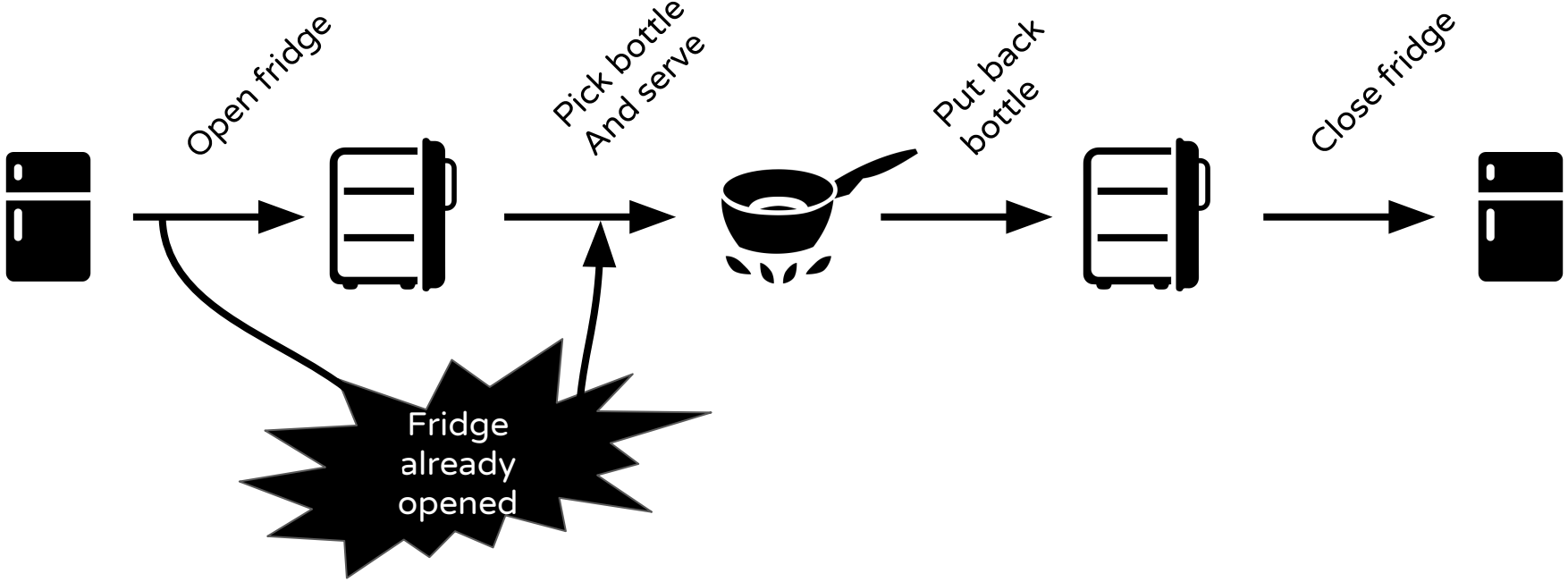
Cooking a dish



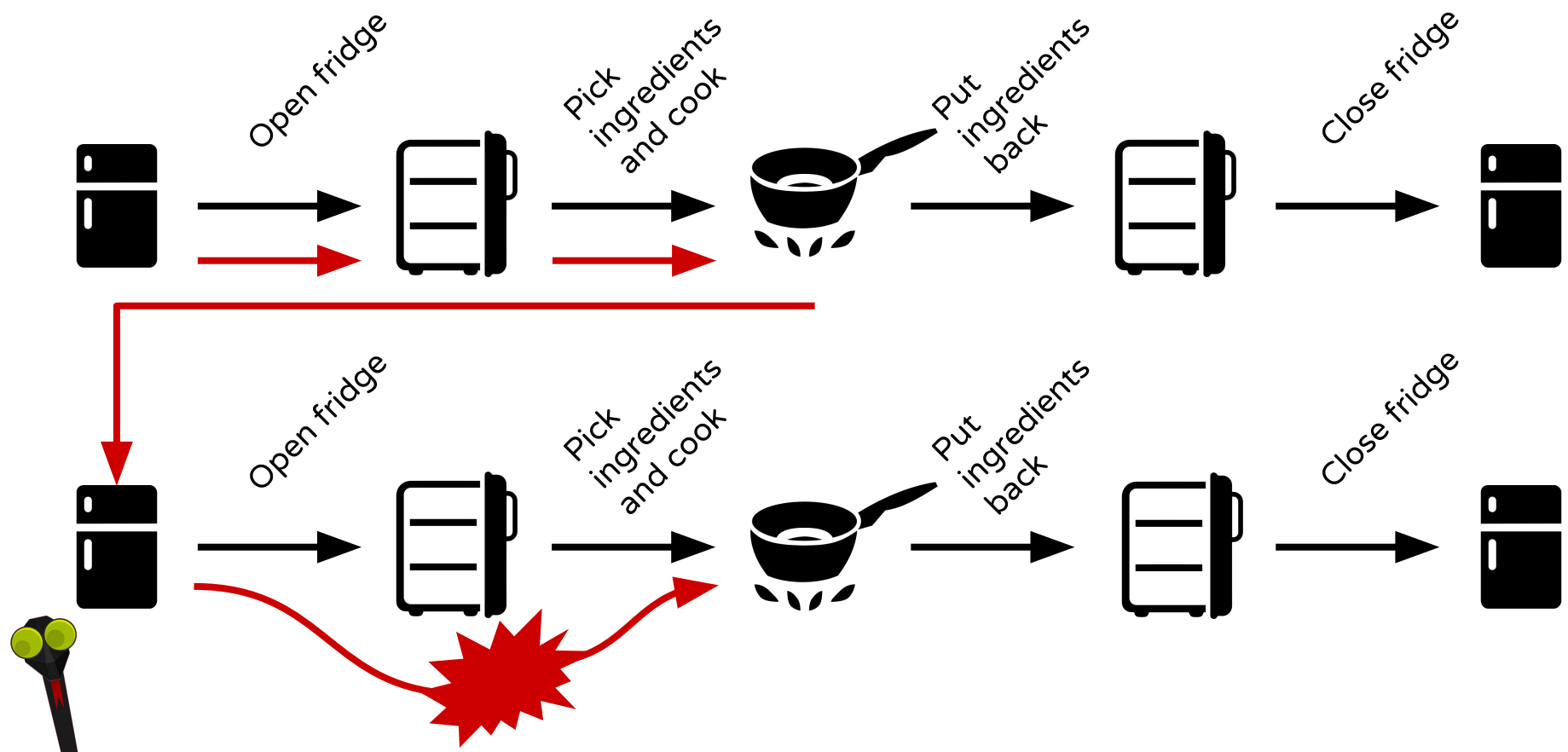
Cooking 2 dishes



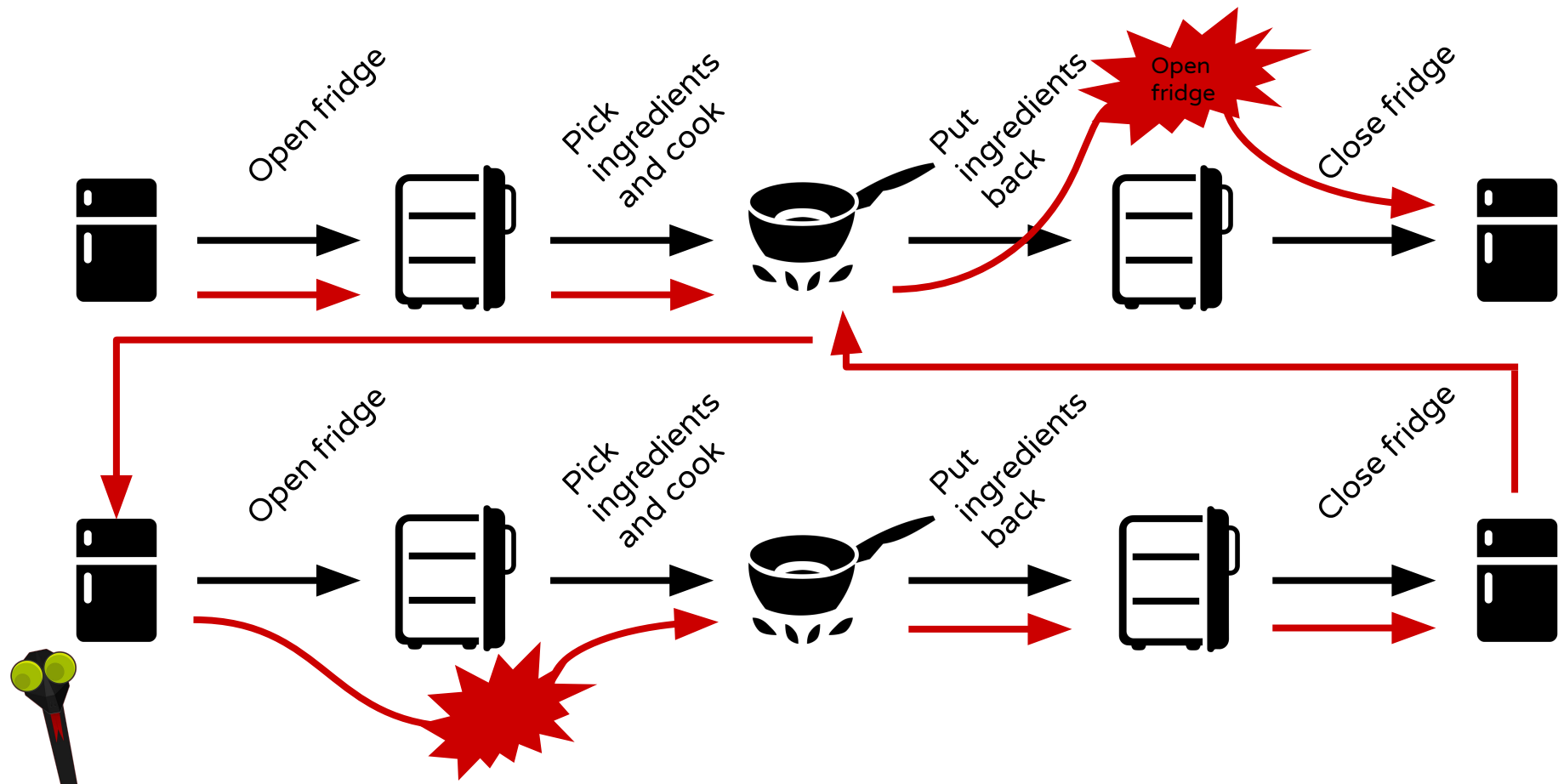
Cooking 2 dishes concurrently



Cooking one dish, handling exception



Cooking two dishes concurrently handling fridge opened



Cooking two dishes concurrently handling fridge opened then closed

Parallelism


Multi-threading

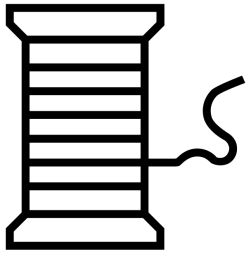
Multi-threading

How to

- Shared global state
- Modules
  - Threading
  - Concurrent.futures
    - threading + pool
  - Futurist
    - concurrent.futures with backlog control and statistics
- threading.Lock

Global Interpreter Lock

Use multi-threading for



- I/O intensive workload that can be parallelized and asynchronous
- Computing stuff without accessing Python data structures
- Running C extensions in parallel

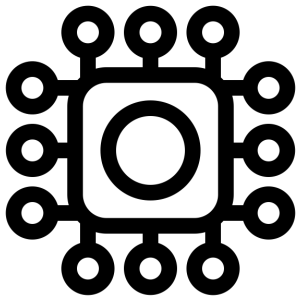

Multi-processes

Multi-processing

How to

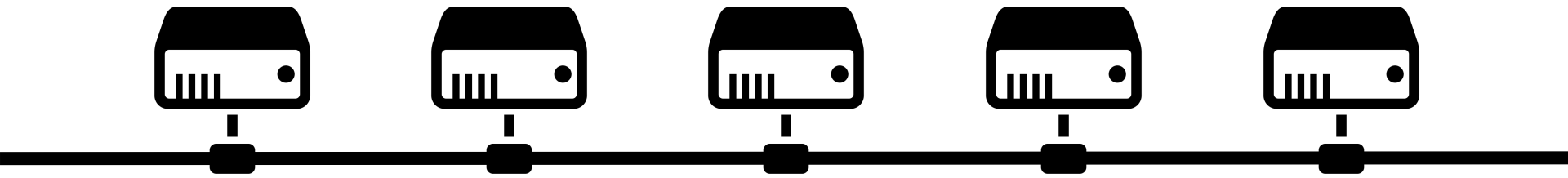
- No shared global state
- 100% CPUs usage!
  - Independent GILs
- Modules
  - Multiprocessing
  - Concurrent.futures
    - Multiprocessing + pool
  - Cotyledon
    - Daemons
- Locks
  - `multiprocessing.Lock`
  - POSIX/SysV IPC
  - File-based locks (fasteners)
- No need for **Go**
  - multiprocessing.Manager().Queue()

Use multi-processing for



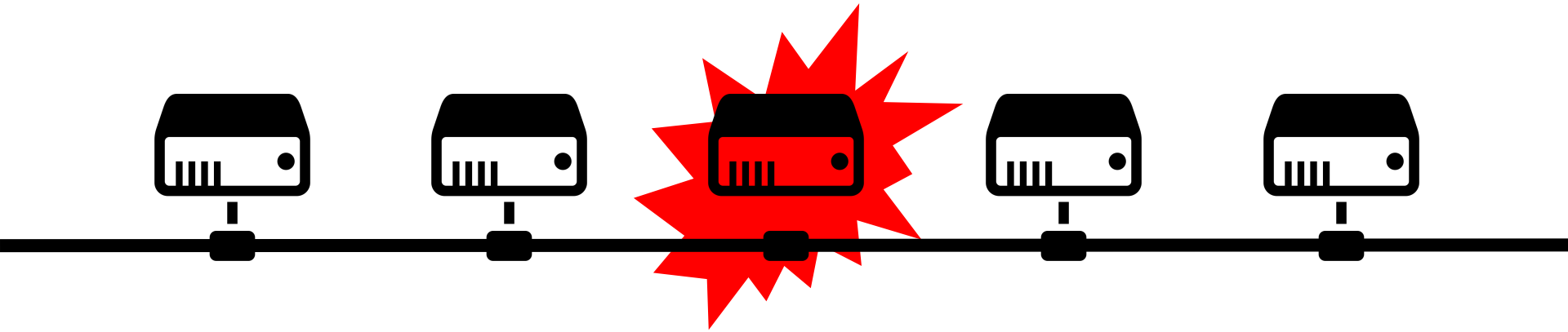
- Daemons, long-running tasks
- Stateless job processing
- First step toward a distributed system

Distributed system



Distributed across network

Multiple servers



What can fail will fail.

Multiple servers

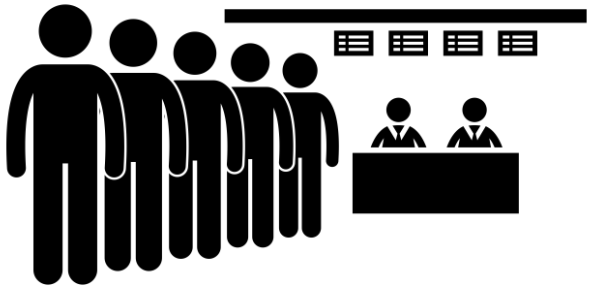
Pros

- No single-point-of-failure
- Horizontal scalability

Cons

- Failure of a node
  - Make failure a default scenario
- Failure of network
- Latency
  - Asynchronicity

Use queues based systems



- Easy to scale
- Easy to implement
- Easy to debug
- … because they are **functional (no-side effect)**

rq

- Based on Redis
- Low barrier of entry
- Easy scalability

Celery

- Multi-broker
- Widely used
- Language agnostic
- Advanced features:
  - Rate limits
  - Routing
  -

Distributed locks

Not all are equals.

- ZooKeeper
- etcd
- Redis
- Consul
- memcached
- PostgreSQL
- MySQL

Tooz implements all of them, give it a try!

etcd

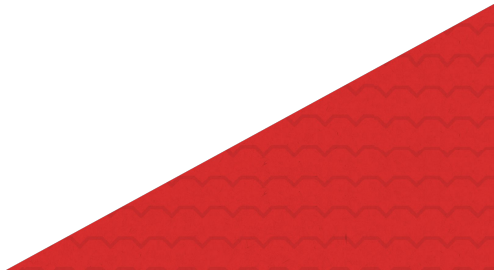
```
import etcd3


client = etcd3.client()
lock = client.lock("foobar")
with lock:
    print("do something")
```

Don’t forget

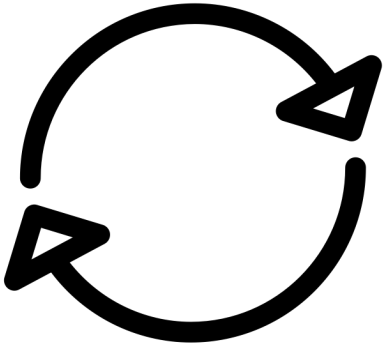
- Client disconnections, crashes, timeouts…
- Concurrency!
- Reliability
- Distributed lock manager crash (SPOF)
- Network latency
- Network outage

Designing for failure

```
while True:
    try:
        do_something()
    except:
        pass
    else:
        break
```



Need to handle:



- When to retry
- How often to retry
- What to do before retrying
- What to do after retrying

tenacity

pip install tenacity

- Define when to retry
- Define how often to retry
- Define what to do before retrying
- Define what to do after retrying

Retry on exceptions and stop at some point

```
import tenacity

@tenacity.retry(
    wait=wait.wait_fixed(1),
    stop=stop.stop_after_delay(60),
    retry=(retry.retry_if_exception_type(IOError) |
           retry.retry_if_result(lambda result: result == None))
def do_something_and_retry():
    do_something()
```

Group membership

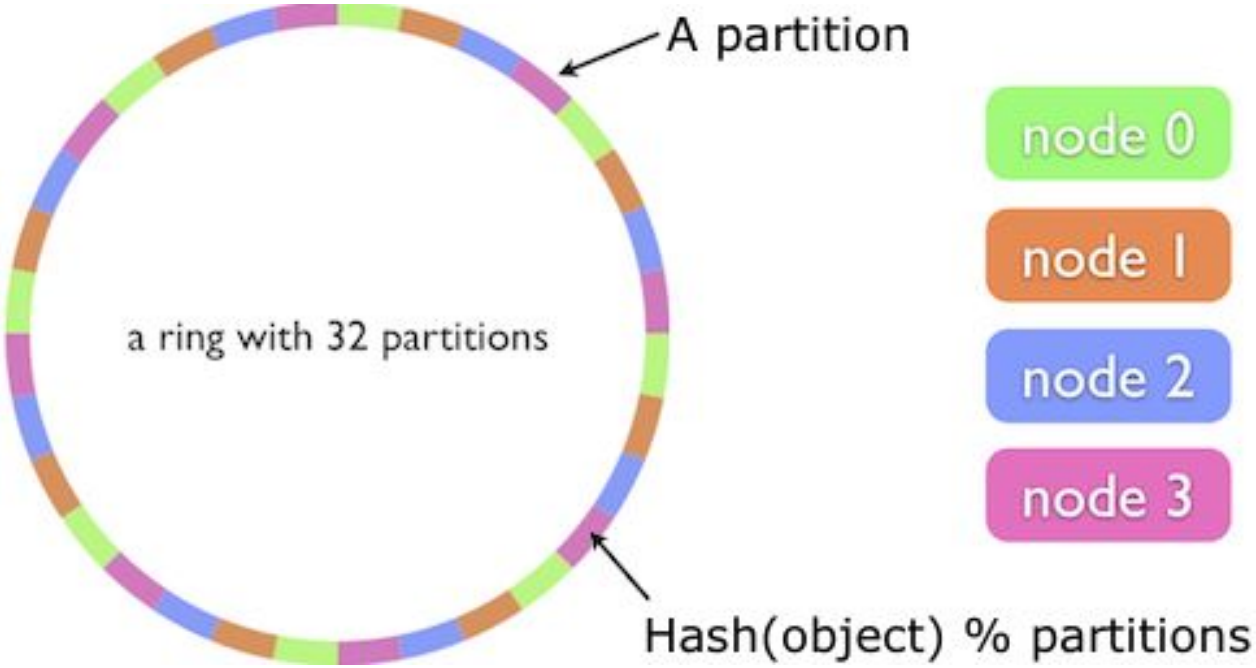
Pick a coordinator

- ZooKeeper
- etcd
- Redis
- Consul
- memcached

Tooz implements all of them, give it a try!

Use cases

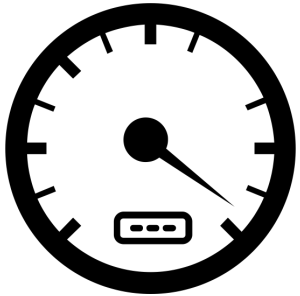
- Workload distribution
- Aliveness check
- Consistent hashring



Consistent hashring

Caching

Memoization or remote cache



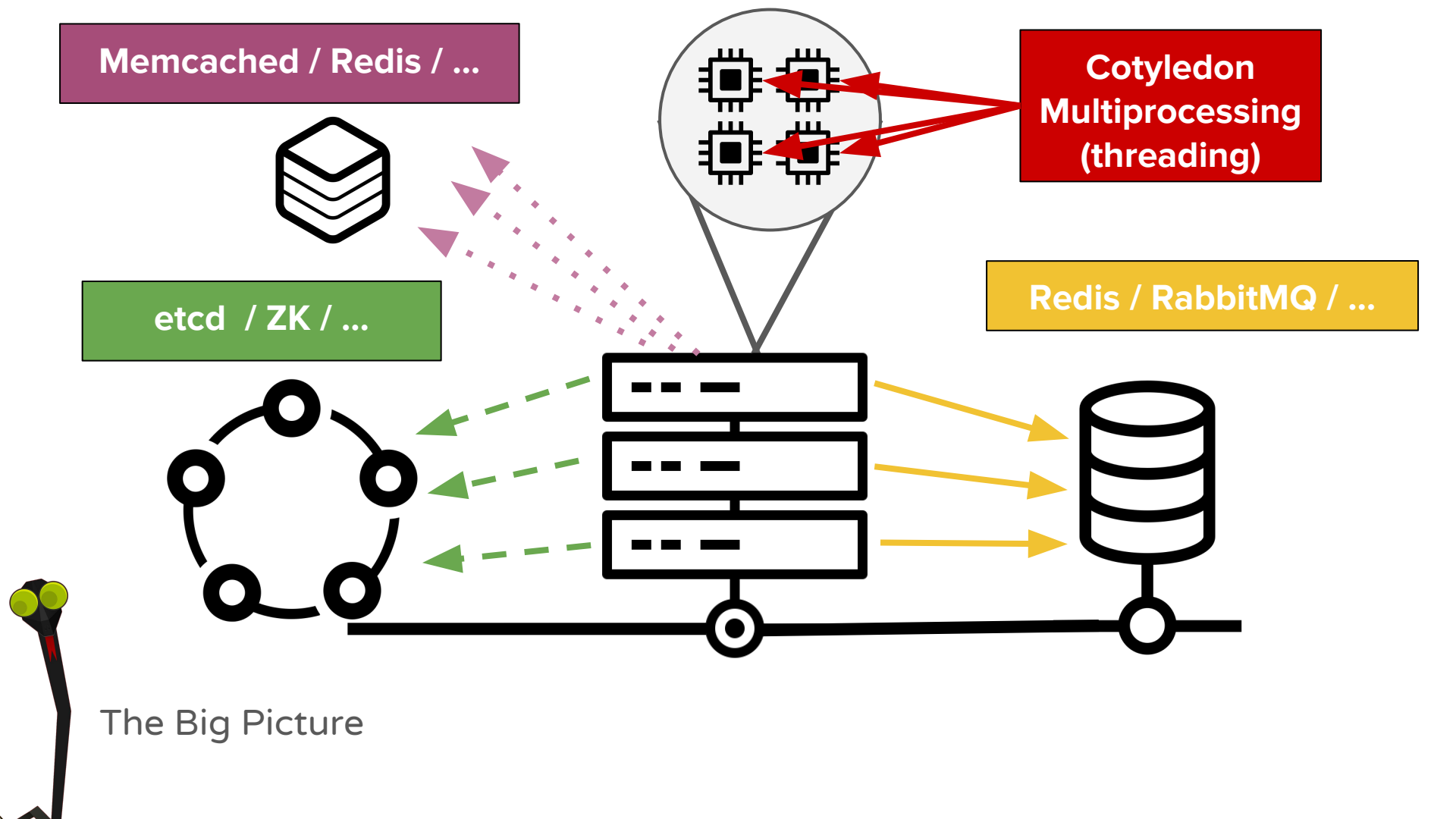
- High cost of computing data
- High latency of accessing data

- cachetools
- functools.lru_cache
- dogpile.cache

- **Cache invalidation**

cachetools

```
>>> import cachetools.func
>>> import math
>>> import time
>>> memoized_sin = cachetools.func.ttl_cache(ttl=5)(math.sin)
>>> memoized_sin(3)
0.1411200080598672
>>> memoized_sin.cache_info()
CacheInfo(hits=0, misses=1, maxsize=128, currsize=1)
>>> memoized_sin(3)
0.1411200080598672
>>> memoized_sin.cache_info()
CacheInfo(hits=1, misses=1, maxsize=128, currsize=0)
>>> time.sleep(5)
>>> memoized_sin.cache_info()
>>> CacheInfo(hits=1, misses=1, maxsize=128, currsize=0)
```



The Big Picture

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Talk inspired by http://scaling-python.com