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 $March\ 5,\ 2018$

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Introduction

In the course of a software development, I needed a buffering medium which was able to fulfil these requirements:

- Capable of handling millions of request per second
- Data persistence
- Easy to use

Where Redis clearly suits them.

My software being written in python, I evidently used one of the redis python client available; redis-py is the recommended one.

This report focuses only on throughput, so data size was fixed and was fitting my needs.

It contains benchmarks of redis-py, where we can see that it is particularly slow compared to the c-written hiredis library or redis-benchmark. It also presents some features and techniques to boost performances, as well as comparing methods to rapidly create strings.

Chapter 1

Aggregated results

1.1 Summary

These tables and plots show a summary for each tested different configuration.

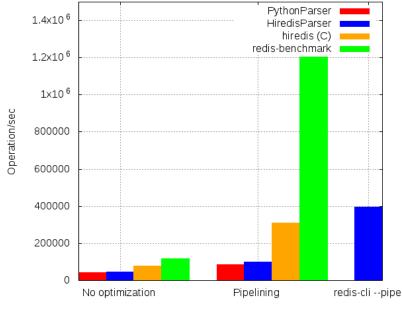
1.1.1 Pushing - LPUSH

	Tool/feature	Parser	Total	time	taken (s)	O	peration/s	ec
			worst	best	average	worst	best	average
2.2	Naive	PythonParser	2.86	2.20	2.37	34,942	45,464	42,158
	Naive	HiredisParser	2.20	2.09	2.17	$45,\!388$	47,745	$46,\!150$
	Naive	hiredis $(C)^1$	1.33	1.23	1.25	$75,\!363$	$81,\!365$	79,977
2.3	Pipelining -P 1000	PythonParser	1.18	1.16	1.17	85,078	85,979	85,642
	Pipelining -P 1000	HiredisParser	1.01	0.96	0.98	$98,\!863$	$104,\!368$	$101,\!535$
	Pipelining -P 1000	hiredis (C)	0.35	0.31	0.32	281,466	$320,\!858$	$308,\!386$
2.5, 2.6.1	redis-cli, format	HiredisParser	0.28	0.23	0.25	358,529	437,094	394,046
2.6.2	redis-cli, concatenation +	HiredisParser	0.25	0.21	0.22	400,002	$487,\!566$	$454,\!178$
2.6.3	redis-cli, substitution %	HiredisParser	0.16	0.15	0.16	$610,\!237$	687,978	638,709
_	redis-cli, (no generation)	HiredisParser	0.08	0.07	0.07	1,287,382	1,523,811	$1,\!400,\!926$
	redis-benchmark -c 1 -d 129			-			-	117,508
	redis-benchmark -c 1 -d 129 -F	1000		-		-	-	1,204,819

1.1.2 Popping - RPOP or LRANGE/LTRIM

	Tool/feature	parser	Total	time	taken (s)	0	peration/s	ec
			worst	best	average	worst	best	average
2.2	Naive	PythonParser	2.68	2.11	2.27	37,301	47,323	44,021
	Naive	HiredisParser	1.96	1.85	1.91	51,050	$53,\!950$	$52,\!383$
	Naive	hiredis (C)	1.10	1.05	1.07	$90,\!656$	$95,\!411$	93,724
2.3	Pipelining	PythonParser	1.10	1.09	1.09	90,862	91,730	91,402
	Pipelining	HiredisParser	0.83	0.77	0.80	$121,\!207$	$129,\!375$	124,924
	Pipelining	hiredis (C)	0.15	0.13	0.14	$659,\!078$	$751,\!359$	$720,\!310$
2.4	LRANGE trick	PythonParser	0.34	0.32	0.33	292,466	311,161	304,744
	LRANGE trick	HiredisParser	0.08	0.07	0.07	$1,\!207,\!644$	1,511,085	1,416,695
	LRANGE trick	hiredis (C)	0.05	0.04	0.04	$2,\!152,\!203$	2,427,007	$2,\!318,\!716$
	redis-benchma	rk -c 1 -d 129 (RPOP)		-		-	-	116,822
	redis-benchma	rk -c 1 -d 129 -P 1000 (RPOP)		-		-	-	1,136,363
	redis-benchma	rk -c 1 -d 129 (LRANGE_100)		-		-	=	4,048,000

 $^{^1{\}rm Official}$ C client



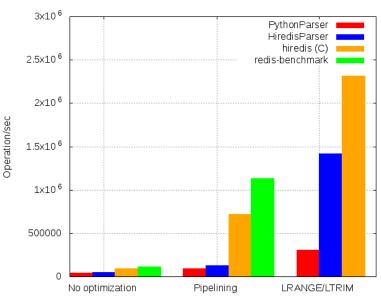
LPUSH:

With no optimization, python performs close to 60% of hiredis (C) performance:

$$\frac{\text{python perf}}{\text{hiredis (C) perf}} = \frac{46,150}{79,977} = 57.70\%$$

With pipelining, python performs close to only 30% of hiredis (C) performance:

$$\frac{\text{python perf}}{\text{hiredis (C) perf}} = \frac{101,535}{308,386} = 30.91\%$$



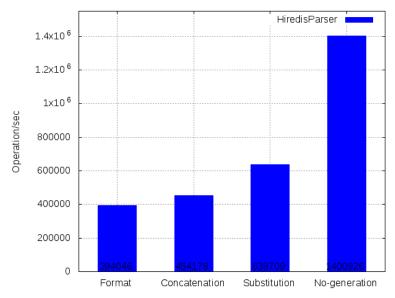
POP:

With no optimization, python performs close to 60% of hiredis (C) performance::

$$\frac{\text{python perf}}{\text{hiredis (C) perf}} = \frac{52,383}{93,724} = 55.89\%$$

With pipelining, python performs close to only 13% of hiredis (C) performance:

$$\frac{\text{python perf}}{\text{hiredis (C) perf}} = \frac{91,402}{720,310} = 12.69\%$$



String Generation:

Method	Cummulative gain
format	0%
Concatenation	15.26%
Substitution	40.63%

Chapter 2

Detailed implentations and tests

2.1 Benchmark parameters

All benchmarks have been performed using these parameters

- Usage of redis unix_socket_path
- Payload: size = 129 bytes

```
'{"origin": null, "channel": 0, "content": "redis@tshark_save:53619abd-a27c-432c-8f8d-1d059aab5f24", "size": 54, "redirect": true}'
```

- Operations:
 - -100000 LPUSH
 - Number of POP may differ if we use RPOP or LRANGE/LTRIM
- For each tests, we are using two different parsers for responses: PythonParser and HiredisParser¹
- In order to simulate a working software popping and adding elements to a buffer, the logic of the benchmark is the folloging:

- For each benchmark, the processing has been done 10 times, then averaged
- For each $100 \left(\frac{100,000}{1,000}\right)$ iterations, we are pushing and popping 1,000 elements

¹Hiredis is a C library that is available with Python bindings, redis-py will attempt to use the HiredisParser if you have the hiredis module installed and will fallback to the PythonParser otherwise.

2.2 Naive implementation

For each payload to be buffered, we push them immediatly. Then, we retrieve them one at a time with a simple POP command.

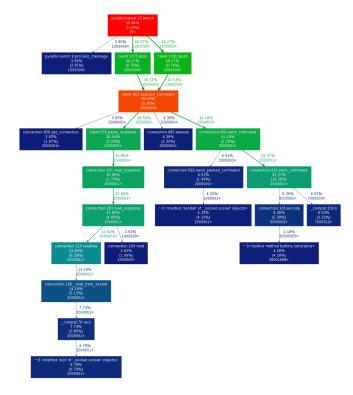
Source code

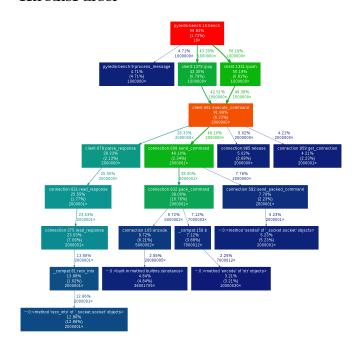
```
1 # push
 t1=time.time()
2
3 for j in range(d):
      redis.lpush('k', payload)
      cpush+=1
6 time_push += time.time()-t1
7
 # pop
8 t1=time.time()
9 for j in range(d):
      msg = redis.rpop('k')
      process_message(msg)
11
12
      cpop+=1
time_pop += time.time()-t1
```

Profiling summary

Cumulative CPU %	PythonParser	HiredisParser	
Processing	3.95%	4.71%	
Receiving response	38.54%	28.33%	
Sending command	41.18%	48.10%	
Parser total gain: 19.24%			

PythonParser





2.3 Pipeline feature

Here, we use the redis pipeline² feature. For each payload to be buffered, we push them in a pipeline wich will execute pending commands every 1000 pushing operations. Then, we retrieve them by sending the POP command in a pipeline which will also execute pending commands every 1000 popping operations.

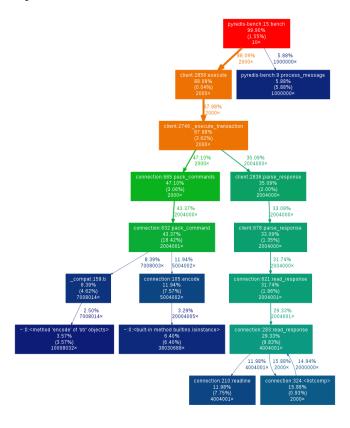
Source code

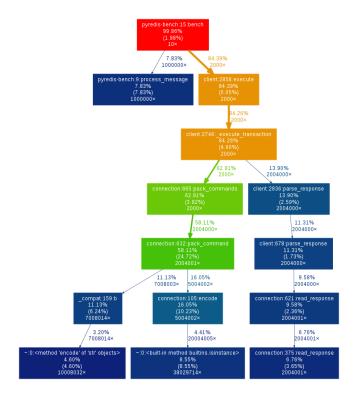
```
# push
  t1=time.time()
  for j in range(d):
3
      pipeline.lpush('k', payload)
5
      cpush+=1
6
  pipeline.execute()
7
  tpush += time.time()-t1
8
  # pop
  t1=time.time()
  for j in range(d):
10
      pipeline.rpop('k')
11
  resp = pipeline.execute()
12
  for msg in resp:
13
      process_message(msg)
14
15
      cpop+=1
  tpop += time.time()-t1
16
```

Profiling summary

Cumulative CPU %	PythonParser	HiredisParser
Processing	5.88%	7.83%
Receiving response	35.09%	13.90%
Sending command	47.10%	62.91%
Parser total gain:	33.1	6%
Gain compare to 2.2:	48.86%	66.24%

PythonParser





²Pipelining allows to send multiple commands to the server without waiting for the replies, and finally read the replies in a single step

2.4 Replacing RPOP by LRANGE/LTRIM

In this implementation, we are still using the redis pipeline feature for pushing data into the buffer, but we modify the popping behavior. Instead of sending one POP command at a time to the pipeline, we fetch a range (100 items) of buffered data with LRANGE, then we trim the buffer so that it mirror the effect of a POP command.

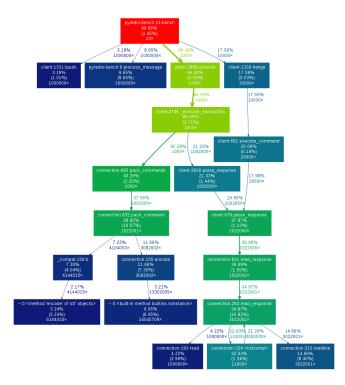
Source code

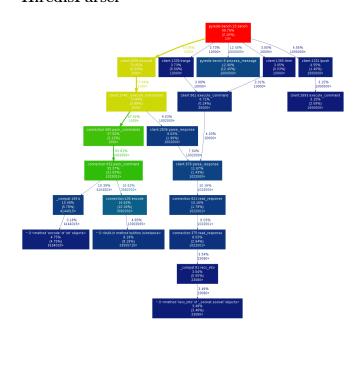
```
1 # lrange_count = 100
2 # push
3 t1=time.time()
  for j in range(d):
4
      pipeline.lpush('k', payload)
6
      cpush+=1
7
  pipeline.execute()
  tpush += time.time()-t1
  # pop
10 t1=time.time()
  for j in range(int(d/lrange_count)+1):
11
      msg_list = redis.lrange('k', -
12
      lrange_count, -1)
      redis.ltrim('k', 0, -len(msg_list)-1)
13
      for msg in msg_list:
14
           process_message(msg)
15
          cpop+=1
16
  tpop += time.time()-t1
```

Profiling summary

Cumulative CPU %	PythonParser	HiredisParser
Processing	8.65%	12.40%
Receiving response	37.87%	11.67%
Sending command	40.29%	57.52%
Parser total gain:	43.3	85%
Gain compare to 2.3:	47.11%	58.37%

PythonParser





2.5 Using redis-cli --pipe

In this implementation, we are still using the redis pipeline feature and the LRANGE/LTRIM replacement of POP. However, we will use the redis-cli binary (provided by redis) with the --pipe options³.

In order to use this binary, we have to generate the valid redis protocol for the wanted command, which will be written to the STDIN of redis-cli.

Source code

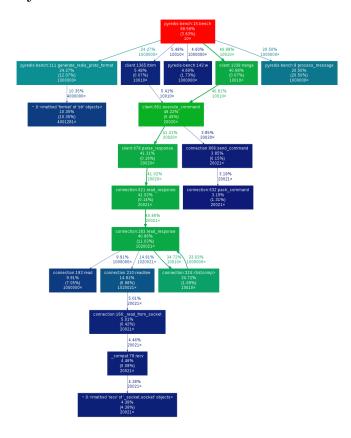
```
# push with redis-cli
2 t1=time.time()
3 for j in range(d):
      write_to_stdin(generate_redis_protocol
      ('lpush', 'k', payload))
      cpush+=1
  tpush += time.time()-t1
  # pop with pipeline
8 t1=time.time()
  for j in range(int(d/lrange_count)+1):
9
      msg_list = redis.lrange('k', -
10
      lrange_count, -1)
      redis.ltrim('k', 0, -len(msg_list)-1)
11
      for msg in msg_list:
12
          process_message(msg)
13
14
          cpop+=1
15
  tpop += time.time()-t1
16
  # flush and close stdin
    pop newly flushed items
17
```

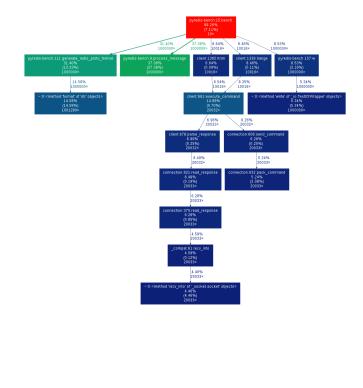
Profiling summary

Cumulative CPU %	PythonParser	HiredisParser
Processing	20.50%	37.08%
Receiving response	41.51%	6.96%
Sending command ^a	$8.45\%^{b}$	$14.79\%^{c}$
Parser total gain:	80.8	37 %
Gain compare to 2.4:	136.99%	199.03%

 $[^]a$ popping + mass insertion with redis-cli

PythonParser





 $^{^{}b}3.85\% + 4.6\%$

 $^{^{}c}6.26\% + 8.53\%$

³redis-cli --pipe consists in writing new commands while you read replies at the same time, ignoring the round trip time for every command

2.6 Generating redis protocol

We saw in section 2.5 that we need to generate the redis protocol ourself. We will now see differents implentation of doing so, each one with drastically performance improvement.

For each algorythm, we use the same source code as in 2.5 with the HiredisParser.

2.6.1 Generating redis protocol with string. Formatter

```
def generate_redis_proto_format(cmd, key, value=''):
2
      cmd_split = cmd.split()
      if value == '':
3
          proto = '*{argNum}\r\n${argLen1}\r\n{arg1}\r\n${argLen2}\r\n'.format(
4
                  argNum=3 if value != '' else 2,
                   argLen1=len(cmd), arg1=cmd,
                   argLen2=len(key), arg2=key
      else:
9
          proto = '*{argNum}\r\n${argLen1}\r\n{arg1}\r\n${argLen2}\r\n${arg2}\r\n${argLen3}\r\n{
10
      arg3}\r\n'.format(
                   argNum=3 if value != '' else 2,
11
                   argLen1=len(cmd), arg1=cmd,
12
13
                   argLen2=len(key), arg2=key,
                   argLen3=len(value), arg3=value)
14
15
      return proto
```

Profiling summary

It should be noted that the cumulative CPU time of process_message entirely depends on the function implentation. Still, it is given as a starting point for comparision.

Cumulative CPU %	HiredisParser
process_message	37.08%
generate_redis_protocol_format	31.40%
Generation loss ⁴	46.87%

2.6.2 Generating redis protocol with String + concatenation operator

```
def generate_redis_proto_concat(cmd, key, value=''):
    cmd_split = cmd.split()
    proto = '*'+(str(3) if value != '' else str(2))+'\r\n'
    proto += '$'+str(len(cmd))+'\r\n'+cmd+'\r\n'
    proto += '$'+str(len(key))+'\r\n'+key+'\r\n'
    if value != '':
        proto += '$'+str(len(value))+'\r\n'+value+'\r\n'
    return proto
```

Profiling summary

Cumulative CPU %	HiredisParser
processing	40.53%
generate_redis_protocol_string	26.34%
Gain compare to 2.6.1:	9.30%
Generation loss	34.26%

⁴Cost of generating the protocol compated to not generating it a all.

2.6.3 Generating redis protocol with String % substitution operator

```
def generate_redis_proto_subst(cmd, key, value=''):
    cmd_split = cmd.split()

if value != '':
    proto = '*%s\r\n$%s\r\n$%s\r\n$%s\r\n$%s\r\n\s\r\n' % ((str(3) if value != '' else str(2)), len(cmd), cmd, len(key), key, len(value), value)
else:
    proto = '*%s\r\n$%s\r\n\s\r\n\s\r\n' % ((str(3) if value != '' else str(2)), len (cmd), cmd, len(key), key)
return proto
```

Profiling summary

Cumulative CPU %	HiredisParser
processing	43.62%
generate_redis_protocol_subst	19.92%
Gain compare to 2.6.2:	7.62%
Generation loss	24.85%

Conclusion

We saw that it is not recommended to use redis-py if we need pure throughput. Even so, some features like pipelining or using a different parser increase the performance reasonably, we are still far from those achieved by hiredis (C).

In anycase, using the HiredisParser can increase performances up to 40% for our case. Don't forget: redis-py uses HiredisParser if it is installed, therefore:

```
pip install redis
pip install hiredis
```