## ChronicleDB on a Raft: Performance Tests

```
In [1]:
          import pandas as pd
          import numpy as np
          import altair as alt
          #pip install vega
          #jupyter nbextension install --sys-prefix --py vega
          #alt.renderers.enable('notebook')
          alt.renderers.enable('html')
         RendererRegistry.enable('html')
Out[1]:
In [58]:
          import requests
          #host = 'http://localhost:8080/api'
          host = 'http://3.121.183.166:8080/api'
          def run test(count, target='event-store'):
              query = { "batchSize": count }
              response = requests.request('GET', f'{host}/sys-info/performance/measure/{target}/insert-events/{count}', params=
              return response.json()
          def clear_stream(stream='demo-event-store', target='event-store'):
              response = requests.request('DELETE', f'{host}/{target}/streams/{stream}/events')
              token = response.ison()
              print(token)
              response = requests.request('POST', f'{host}/{target}/clear-request-confirmation', json=token)
              return response
In [59]:
          clear_stream('demo_event_store', target='event-store/embedded')
         {'streamName': 'demo event store', 'token': '8 lqIecvKQimY9cpEha522qhH0UwDYSn'}
         <Response [200]>
In [60]:
          # TODO test on non-replicated chronicle db and compare
          # To create streams in the embedded db
          def create schema for embedded db(schema):
              response = requests.request('POST', f'{host}/event-store/embedded/streams', json=schema)
              return response
In [61]:
          create_schema_for_embedded_db({
               "streamName": "demo_event_store",
              "schema": [
                  {
                       "name": "SYMBOL",
"type": "STRING",
                       "properties": {}
                       "name": "SECURITYTYPE",
"type": "INTEGER",
                       "properties": {}
                      "name": "LASTTRADEPRICE",
"type": "FLOAT",
                       "properties": {}
                  }
              ]
          })
         <Response [500]>
Out[61]:
In [62]:
          import datetime
          import pandas as pd
          def run_tests(counts=[1,100], trials_per_count=1, env_info={}), target='event-store', stream='demo-event-store', csv_d
              now = datetime.datetime.now().replace(microsecond=0).isoformat().replace(':', '-')
              csv_name = f'results-{now}.csv
              env info columns = list(env info.keys())
              df = pd.DataFrame(columns = [*env_info_columns, 'buffer_size_in_bytes', 'event_count', 'trial', 'duration_in_ms',
              for count in counts:
                  for i in range(0, trials_per_count):
                      print(f'Trial {i+1} for event count {count}')
```

```
results = run_test(count, target)
                       print(results['message'])
                       df = df.append({
                           **env_info,
                           'buffer_size_in_bytes': results['bufferSize'],
                           'trial': i+1,
                           'event_count': count,
                           'duration_in_ms': results['timeElapsed'],
                           'measured_on': now
                       }, ignore_index=True)
                       df.to csv(f'measurements/{csv dir}/{csv name}', index=False)
                       # clean service after each trial. Must delete all events to avoid 000
                       stream_cleared = clear_stream(stream, target)
                      print(stream_cleared)
              return df
In [14]: | clear_stream()
          {'streamName': 'demo-event-store', 'token': 'wBINNQQtadtotOFZKgo-jathuDzaJlNc'}
         <Response [200]>
Out[14]:
In [72]:
          env info = {
              #'cluster_type': 'localMacBookProIntel19',
'cluster_type': 'awsLightsail2GB',
              'node_count': 3,
              'event_type': 'randomized',
              'buffer_type': 'blocking'
          run tests(counts=[1, 100, 10000, 1000000], trials per count=10, env info=env info, target='event-store', csv dir='aws
          # run_tests(counts=[10000000], trials_per_count=10, env_info=env_info)
          # an event of our examples has ~21 bytes
```

In [64]:	# measure performance of non-replicated, embedded store (the original one)
	<pre>env_info = {     'cluster_type': 'standalone-embedded-aws',     'node_count': 1,     'event_type': 'randomized',     'buffer_type': 'none-embedded' }</pre>
	run_tests(counts=[1, 100, 10000, 1000000], trials_per_count=10, env_info=env_info, target='event-store/embedded', str

64]:		cluster_type	node_count	event_type	buffer_type	buffer_size_in_bytes	event_count	trial	duration_in_ms	measured_on
_	0	standalone-embedded-aws	1	randomized	none-embedded	0	1	1	6	2022-03-16T15-44-04
	1	standalone-embedded-aws	1	randomized	none-embedded	0	1	2	6	2022-03-16T15-44-04
	2	standalone-embedded-aws	1	randomized	none-embedded	0	1	3	147	2022-03-16T15-44-04
	3	standalone-embedded-aws	1	randomized	none-embedded	0	1	4	9	2022-03-16T15-44-04
	4	standalone-embedded-aws	1	randomized	none-embedded	0	1	5	4	2022-03-16T15-44-04
	5	standalone-embedded-aws	1	randomized	none-embedded	0	1	6	12	2022-03-16T15-44-04
	6	standalone-embedded-aws	1	randomized	none-embedded	0	1	7	7	2022-03-16T15-44-04
	7	standalone-embedded-aws	1	randomized	none-embedded	0	1	8	9	2022-03-16T15-44-04
	8	standalone-embedded-aws	1	randomized	none-embedded	0	1	9	2	2022-03-16T15-44-04
	9	standalone-embedded-aws	1	randomized	none-embedded	0	1	10	0	2022-03-16T15-44-04
	10	standalone-embedded-aws	1	randomized	none-embedded	0	100	1	86	2022-03-16T15-44-04
	11	standalone-embedded-aws	1	randomized	none-embedded	0	100	2	3	2022-03-16T15-44-04
	12	standalone-embedded-aws	1	randomized	none-embedded	0	100	3	3	2022-03-16T15-44-04
	13	standalone-embedded-aws	1	randomized	none-embedded	0	100	4	22	2022-03-16T15-44-04
	14	standalone-embedded-aws	1	randomized	none-embedded	0	100	5	15	2022-03-16T15-44-04
	15	standalone-embedded-aws	1	randomized	none-embedded	0	100	6	3	2022-03-16T15-44-04
	16	standalone-embedded-aws	1	randomized	none-embedded	0	100	7	3	2022-03-16T15-44-04
	17	standalone-embedded-aws	1	randomized	none-embedded	0	100	8	9	2022-03-16T15-44-04
	18	standalone-embedded-aws	1	randomized	none-embedded	0	100	9	6	2022-03-16T15-44-04
	19	standalone-embedded-aws	1	randomized	none-embedded	0	100	10	26	2022-03-16T15-44-04
	20	standalone-embedded-aws	1	randomized	none-embedded	0	10000	1	60	2022-03-16T15-44-04

measured	duration_in_ms	trial	event_count	$buffer\_size\_in\_bytes$	buffer_type	event_type	node_count	cluster_type	
2022-03-16T15-4	34	2	10000	0	none-embedded	randomized	1	standalone-embedded-aws	21
2022-03-16T15-4	23	3	10000	0	none-embedded	randomized	1	standalone-embedded-aws	22
2022-03-16T15-4	28	4	10000	0	none-embedded	randomized	1	standalone-embedded-aws	23
2022-03-16T15-4	9	5	10000	0	none-embedded	randomized	1	standalone-embedded-aws	24
2022-03-16T15-4	15	6	10000	0	none-embedded	randomized	1	standalone-embedded-aws	25
2022-03-16T15-4	19	7	10000	0	none-embedded	randomized	1	standalone-embedded-aws	26
2022-03-16T15-4	141	8	10000	0	none-embedded	randomized	1	standalone-embedded-aws	27
2022-03-16T15-4	15	9	10000	0	none-embedded	randomized	1	standalone-embedded-aws	28
2022-03-16T15-4	19	10	10000	0	none-embedded	randomized	1	standalone-embedded-aws	29
2022-03-16T15-4	580	1	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	30
2022-03-16T15-4	422	2	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	31
2022-03-16T15-4	554	3	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	32
2022-03-16T15-4	460	4	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	33
2022-03-16T15-4	440	5	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	34
2022-03-16T15-4	398	6	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	35
2022-03-16T15-4	507	7	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	36
2022-03-16T15-4	388	8	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	37
2022-03-16T15-4	440	9	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	38
2022-03-16T15-4	582	10	1000000	0	none-embedded	randomized	1	standalone-embedded-aws	39

```
In [23]:
         # TODO we actually don't know how long the insert is in total as we are async as soon as events are offered to the bu
         \# So, as we are waiting for buffer flushes, but not the last flush, we miss the time it tooks to apply the last flush
          # TODO therefore need some callback in event store to know when the store is done, not just the buffer...
          # example: Inserted events 10000 times asynchronously [45ms] <-- this is just the time it takes to insert them into the
          \# -> at least 4.5 sec for 1mio events to put into the buffer
          # -> raft log entry size is limited! -> # Log entry size 13383315 exceeds the max buffer limit of 4194304
          # raft.server.log.appender.buffer.byte-limit = 4MB
In [ ]:
         # TODO test with maximum size buffer
```

```
In [ ]:
         # TODO test with raft log entries not persisted
```

## Evaluating the benchmark results

All of this is currently under conditions without out-of-order events

```
In [2]:
                                            # load all measurements
                                            from os import walk
                                            resource_folder = 'measurements'
                                            resources = next(walk(resource_folder), (None, None, []))[2]
                                            print(f'{len(resources)} files in total')
                                        24 files in total
In [3]:
                                            datasets = {filename: pd.read csv(f'{resource folder}/{filename}', sep=',', encoding='utf-8', error bad lines=False)
                                         /var/folders/mb/g5kpvlh145j5qpm9s7fmb08c0000gn/T/ipykernel\_85187/3005144025.py: 1: FutureWarning: The error\_bad\_lines for the error\_bad_lines for th
                                        argument has been deprecated and will be removed in a future version.
```

datasets = {filename: pd.read\_csv(f'{resource\_folder}/{filename}', sep=',', encoding='utf-8', error\_bad\_lines=Fals e) for filename in resources if not filename.startswith(".")}

```
In [4]:
         all measurements df = pd.concat(datasets, ignore index=True)
         all\_measurements\_df
```

Out[4]:		cluster_type	node_count	event_type	buffer_type	buffer_size_in_bytes	event_count	trial	duration_in_ms	measured_on
	0	localI7	3	randomized	blocking	1048576	10000000	1	124276	2022-01-29T15-16-22
	1	localI7	3	randomized	blocking	1048576	10000000	2	122494	2022-01-29T15-16-22
	2	localI7	3	randomized	blocking	1048576	10000000	3	117724	2022-01-29T15-16-22

	cluster_type	node_count	event_type	buffer_type	buffer_size_in_bytes	event_count	trial	duration_in_ms	measured_on
3	localI7	3	randomized	blocking	1048576	10000000	4	119596	2022-01-29T15-16-22
4	localI7	3	randomized	blocking	1048576	10000000	5	114629	2022-01-29T15-16-22
830	localI7	3	randomized	blocking	10240	1000000	6	20389	2022-01-29T12-48-43
831	localI7	3	randomized	blocking	10240	1000000	7	20635	2022-01-29T12-48-43
832	localI7	3	randomized	blocking	10240	1000000	8	21030	2022-01-29T12-48-43
833	localI7	3	randomized	blocking	10240	1000000	9	21108	2022-01-29T12-48-43
834	localI7	3	randomized	blocking	10240	1000000	10	21375	2022-01-29T12-48-43

835 rows × 9 columns

```
In [5]: # supress warnings
pd.options.mode.chained_assignment = None

# remove rows with duration = 0 to avoid divide by zero
all_measurements_df = all_measurements_df[all_measurements_df['duration_in_ms'] > 0]

# replace buffer_size 0 with 1 to allow log scale plotting
all_measurements_df.loc[all_measurements_df['buffer_size_in_bytes'] == 0, 'buffer_size_in_bytes'] = 1

all_measurements_df['duration_per_event'] = all_measurements_df['duration_in_ms'] / all_measurements_df['event_count'
all_measurements_df['duration_per_mio_events'] = all_measurements_df['duration_per_event'] * 1000000
all_measurements_df['duration_per_mio_events_in_sec'] = all_measurements_df['duration_per_mio_events'] / 1000
all_measurements_df['events_per_second'] = 1000 / all_measurements_df['duration_per_event']
pd.options.mode.chained_assignment = 'warn'
all_measurements_df.describe()
```

Out[5]:		node_count	buffer_size_in_bytes	event_count	trial	duration_in_ms	duration_per_event	$duration\_per\_mio\_events$	duration_per_mio_eve
	count	743.000000	7.430000e+02	7.430000e+02	743.000000	743.000000	743.000000	7.430000e+02	7.
	mean	3.189771	6.178415e+05	3.998901e+05	5.425303	14820.282638	2.483687	2.483687e+06	24
	std	1.155231	5.059331e+05	1.204331e+06	2.891013	61681.256728	12.500844	1.250084e+07	125
	min	1.000000	1.000000e+00	1.000000e+00	1.000000	1.000000	0.000388	3.880000e+02	
	25%	3.000000	1.024000e+04	1.000000e+02	3.000000	4.000000	0.008908	8.908000e+03	
	50%	3.000000	1.048576e+06	1.000000e+04	5.000000	37.000000	0.020000	2.000000e+04	1
	75%	3.000000	1.048576e+06	1.000000e+06	8.000000	7308.000000	0.225000	2.250000e+05	2:
	max	6.000000	1.048576e+06	1.000000e+07	10.000000	485271.000000	246.000000	2.460000e+08	2460

```
#replicated_measurements_df = all_measurements_df[all_measurements_df['cluster_type'] != 'standalone-embedded']
replicated_measurements_df = all_measurements_df[all_measurements_df['cluster_type'] == 'localIT']
replicated_measurements_3_nodes_df = replicated_measurements_df[replicated_measurements_df['node_count'] == 3]
embedded_measurements_df = all_measurements_df[all_measurements_df['cluster_type'] == 'standalone-embedded']
remote_replicated_measurements_df = all_measurements_df[all_measurements_df['cluster_type'] == 'awsLightsail2GB']
local_and_remote_replicated_measurements_3_nodes_df = all_measurements_df[all_measurements_df['node_count'] == 3]
```

```
In [7]:
         import altair as alt
         def plot_event_rate(df, title="Event Throughput Rates", show_benchmark=False, show_benchmark_at=10000000, aggregate="""
             buffer_size_label_expr = "datum.label && datum.label[0] == '1' ? (datum.value >= 1000000 ? datum.value / 1000000
                 buffer_size_label_expr = f'datum.value == {show_benchmark_at} ? "Standalone" : ({buffer_size_label_expr})'
             blocking_queue_event_rate_plot = alt.Chart(df).mark_bar(clip=True, width=15).encode(
                 x=alt.X('buffer_size_in_bytes:Q', scale=alt.Scale(type='log'), title="Buffer Size (in Bytes, Log Scale)", axi
                     labelExpr=buffer size label expr
                 y=alt.Y(f'{aggregate}(events_per_second):Q', title=f"{aggregate.capitalize()} of Events/sec", scale=alt.Scale
                 color=alt.condition(
                     alt.datum.buffer_size_in_bytes == show_benchmark_at,
                     alt.value('#919eac'),
                     alt.value('#4c78a8')
             ).properties(width=300, height=300, title=title)
             return blocking queue event rate plot + blocking queue event rate plot.mark text(
                 align='center',
                 color='black',
                 dx=0,
                 dy=-8
             ) .encode(
                 text=alt.Text(f'{aggregate}(events_per_second):Q', format=',.0f'),
```

```
color=alt.value('black')
              ).transform_calculate(label='datum.y + " inches"')
Tn [8]:
         buffer_size_label_expr = "datum.label && datum.label[0] == '1' ? (datum.value >= 1000000 ? datum.value / 1000000 +
         duration_plot = alt.Chart(replicated_measurements_3_nodes_df[replicated_measurements_3_nodes_df['buffer_size_in_bytes
              x=alt.X('median(duration per mio events in sec):Q', title="Median of Duration per 1 Mio. Events (in Seconds)"),
              y=alt.Y('buffer size in bytes:Q', scale=alt.Scale(type='log'), title="Buffer Size (in Bytes, Log Scale)", axis=al
                  labelExpr=buffer_size_label_expr
         ).properties(width=300, title="Time to ingest 1 Mio Events (3 Nodes)")
         duration_plot = duration_plot + duration_plot.mark_text(
             align='right',
              color='white',
              dx=-4.
              #dy=-18
         ) .encode(
              text=alt.Text('median(duration_per_mio_events_in_sec):Q', format=',.0f'),
         duration plot | plot event rate(replicated measurements 3 nodes df[replicated measurements 3 nodes df['buffer size in
                     Time to ingest 1 Mio Events (3 Nodes)
                                                                           Event Throughput Rates (3 Nodes)
Out[8]:
           10 MB
                                                               110.000
                                                                                                    99.880
                                                               100,000
                                                                90,000
                                                                                            80.818
         Bytes, Log Scale)
            1 MB
                                                                80,000
                                                                70,000
                                                                60.000
                                                               50,000
         Ξ
         Size
                                                                40,000
```

In [9]:
 blocking\_queue\_df = replicated\_measurements\_3\_nodes\_df[replicated\_measurements\_3\_nodes\_df['buffer\_type'] == 'blocking\_non\_blocking\_queue\_df = replicated\_measurements\_3\_nodes\_df[replicated\_measurements\_3\_nodes\_df['buffer\_type'] == 'non-plot\_event\_rate(blocking\_queue\_df, title="Event Throughput Rates using Blocking Queue") | plot\_event\_rate(non\_blocking\_queue\_df) | plot\_event\_rate(non\_blo

100 B

30,000 20,000 10,000

1 KB

10 12 14 16 18 20 22

lian of Duration per 1 Mio. Events (in Seconds)

4.960

1 KB

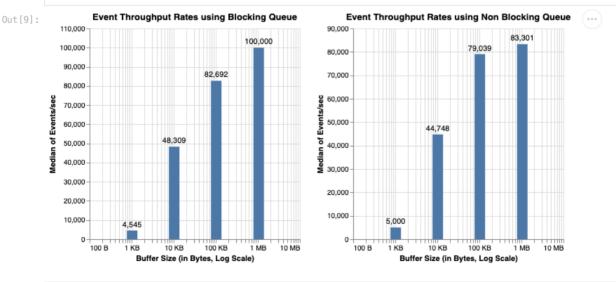
10 KB

100 KB

Buffer Size (in Bytes, Log Scale)

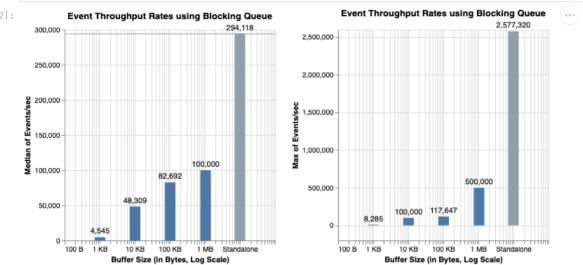
1 MB

10 MB

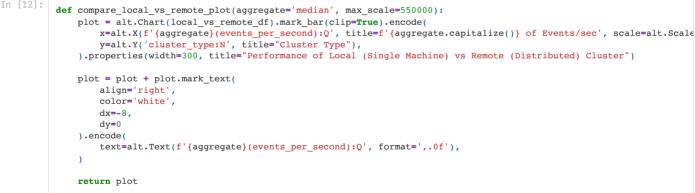


```
In [22]: blocking_queue_vs_standalone_df = all_measurements_df[((all_measurements_df['buffer_type'] == 'blocking') & (all_meas show_benchmark_at = 10000000
    blocking_queue_vs_standalone_df.loc[(blocking_queue_vs_standalone_df['buffer_type'] == 'none-embedded'), 'buffer_size median_plot = plot_event_rate(blocking_queue_vs_standalone_df, title="Event Throughput Rates using Blocking Queue", s max_plot = plot_event_rate(blocking_queue_vs_standalone_df, title="Event Throughput Rates using Blocking Queue", show benchmark_score = all_measurements_df[all_measurements_df['buffer_type'] == 'none-embedded']['events_per_second'].med median_plot + alt.Chart(pd.DataFrame({'y': [benchmark_score]})).mark_rule(color="#919eac").encode(y='y') | max_plot
```

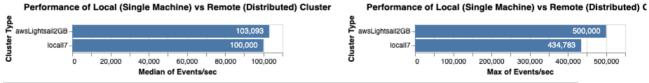
```
# a = plot_event_rate(all_measurements_df[(all_measurements_df['buffer_type'] == 'blocking')], title="Event Throughpu"
# b = plot_event_rate(all_measurements_df[(all_measurements_df['buffer_type'] == 'none-embedded')], title="Event Throughpu"
# a + b
# a + b
```



## Comparison of Local Cluster (Single Machine) vs. Remote Cluster on AWS







- [] multiple nodes on AWS
- [] benchmarking against standalone on aws node

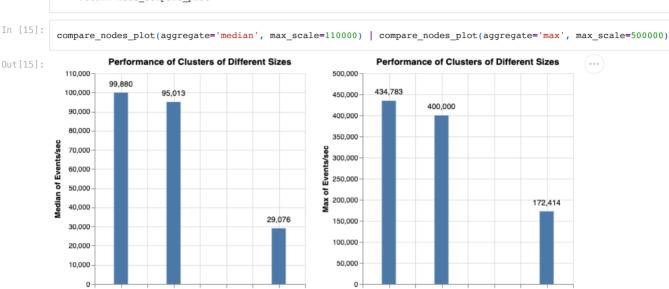
# Comparison of Different Cluster Sizes (# of Nodes)

Evaluation has been run on local machine

```
).properties(width=300, title="Performance of Clusters of Different Sizes")

node_compare_plot = node_compare_plot + node_compare_plot.mark_text(
    align='center',
    color='black',
    dx=0,
    dy=-12
).encode(
    text=alt.Text(f'{aggregate}(events_per_second):Q', format=',.0f'),
)

return node_compare_plot
```



## Comparison of Blocking vs Non-Blocking Buffer

```
In [16]:
          buffered measurements df = replicated measurements df[replicated measurements df['buffer type'] != 'none']
          time_per_buffer_plot = alt.Chart(buffered_measurements_df).mark_bar(clip=True).encode(
              x=alt.X('median(duration_per_mio_events_in_sec):Q', title="Median of Duration per 1 Mio. Events (in Seconds)"),
              y=alt.Y('buffer_type:N',title="Buffer Type"),
          ).properties(width=300, title="Median Time to ingest 1 Mio Events per Buffer Type")
          events_per_sec_plot = alt.Chart(replicated_measurements_df).mark_bar(clip=True).encode(
              x=alt.X('median(events_per_second):Q', title="Median of Events/sec", scale=alt.Scale(domain=[0, 80000])),
              y=alt.Y('buffer_type:N',title="Buffer Type"),
          ).properties(width=300, title="Median of Events per Second by Buffer Type")
          events_per_sec_plot = events_per_sec_plot + events_per_sec_plot.mark_text(
              align='left
              color='black',
              dx=4,
              #dy=-18
          ).encode(
              text=alt.Text('median(events_per_second):Q', format=',.0f'),
          time_per_buffer_plot | events_per_sec_plot
```

Out[16]: Median Time to ingest 1 Mio Events per Buffer Type Median of Events per Second by Buffer Type 61.031 blocking blockingnon-blocking non-blocking 33.333 95 10 15 20 25 30 Median of Duration per 1 Mio. Events (in Seconds) 80.000 40,000 60,000 20,000

### Comparison with Standalone/Embedded ChronicleDB

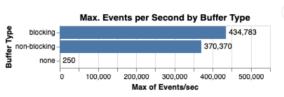
```
events_per_sec_plot = events_per_sec_plot + events_per_sec_plot.mark_text(
    align='left',
    color='black',
    dx=4,
    #dy=-18
).encode(
    text=alt.Text('max(events_per_second):Q', format=',.0f'),
)
time_per_buffer_plot | events_per_sec_plot
```

Out [17]:

Min. Time to ingest 1 Mio Events per Buffer Type

blocking
non-blocking
0.0 0.5 1.0 1.5 2.0 2.5

Min of Duration per 1 Mio. Events (in Seconds)



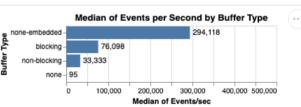
```
In [18]:
                                 buffered and embedded measurements df = all measurements df[all measurements df[buffer type'] != 'none']
                                 time per buffer plot = alt.Chart(buffered and embedded measurements df).mark bar(clip=True).encode(
                                             x=alt.X('median(duration_per_mio_events_in_sec):Q', title="Median of Duration per 1 Mio. Events (in Seconds)"),
y=alt.Y('buffer_type:N',title="Buffer Type", sort='-x'),
                                 ).properties(width=300, title="Median Time to ingest 1 Mio Events per Buffer Type")
                                 events\_per\_sec\_plot = alt.Chart(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).mark\_bar(clip=True).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_measurements\_df).encode(all\_meas
                                              x=alt.X('median(events_per_second):Q', title="Median of Events/sec", scale=alt.Scale(domain=[0, 500000])),
                                              y=alt.Y('buffer_type:N',title="Buffer Type", sort='-x'),
                                 ).properties(width=300, title="Median of Events per Second by Buffer Type")
                                 events_per_sec_plot = events_per_sec_plot + events_per_sec_plot.mark_text(
                                              align='left'
                                              color='black'
                                              dx=4,
                                              #dy=-18
                                 ) .encode(
                                              text=alt.Text('median(events_per_second):Q', format=',.0f'),
                                 time_per_buffer_plot | events_per_sec_plot
```

Out [18]:

Median Time to ingest 1 Mio Events per Buffer Type

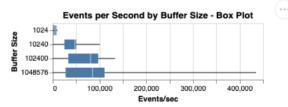
non-blockingblockingnone-embedded0 5 10 15 20 25 30

Median of Duration per 1 Mio. Events (in Seconds)



In [19]:
 alt.Chart(buffered\_measurements\_df).mark\_boxplot(extent='min-max').encode(
 y=alt.Y('buffer\_type:N',title="Buffer Type"),
 x=alt.X('events\_per\_second:Q', title="Events/sec"),
).properties(width=300, title="Events per Second by Buffer Type - Box Plot") | alt.Chart(buffered\_measurements\_df).ma
 y=alt.Y('buffer\_size\_in\_bytes:O',title="Buffer Size"),
 x=alt.X('events\_per\_second:Q', title="Events/sec"),
).properties(width=300, title="Events per Second by Buffer Size - Box Plot")

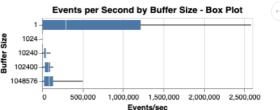
blocking non-blocking 100,000 200,000 300,000 400,000 Events/sec



```
alt.Chart(buffered_and_embedded_measurements_df).mark_boxplot(extent='min-max').encode(
    y=alt.Y('buffer_type:N',title="Buffer Type"),
        x=alt.X('events_per_second:Q', title="Events/sec"),
).properties(width=300, title="Events per Second by Buffer Type - Box Plot") | alt.Chart(buffered_and_embedded_measur
    y=alt.Y('buffer_size_in_bytes:O',title="Buffer Size"),
    x=alt.X('events_per_second:Q', title="Events/sec"),
).properties(width=300, title="Events per Second by Buffer Size - Box Plot")
```

Out[19]:

#### 



Running the cluster without a buffer leads to 100% utilization of the machines IO, as in the current naive implementation of the raft log and state machine, each event is sent to all nodes, needs to be committed by at least a quorum of notes and is also written into the raft log of each node.

{localHostName=ip-172-26-0-78.eu-central-1.compute.internal, javaVersion=11, localHostAddress=172.26.0.78, jdkVersion=11.0.14.1, storagePath=/home/ec2-user/chronicledb, osName=Linux, springVersion=5.3.9, osVersion=4.14.262-200.489.amzn2.x86\_64, remoteHostAddress=3.121.183.166, nodeId=n1}

