# Using Container Migration for HPC Workloads Resilience

Mohamad Sindi & John R. Williams

Massachusetts Institute of Technology Center for Computational Engineering (CCE)

HPEC'19, Sept 26 2019

# Agenda

The issue

Proposed mitigation method

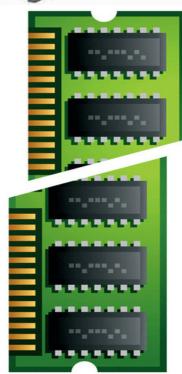
Demo

Main contributions & summary

#### The Issue

- Today's top HPC supercomputers are running in Petascale computing power (thousands of nodes, several millions of cores).
- Mean Time Between Failures (MTBF) for some of today's top HPC Petaflop systems is reported to be several days.
- Exascale computing is expected by 2020-2021 (billion cores).
- Some studies estimate MTBF for Exascale systems to be less than 60 minutes.
- Running sustainable workloads on such systems is becoming more challenging as the size of the HPC system grows.





#### Current Methods to Tolerate Failures

- Checkpoint-restart (CR) mechanism is commonly used (application periodically saves its state, it can restart from last checkpoint incase of failure).
- Popular tool for this is Berkeley Lab Checkpoint/Restart (BLCR).





#### Limitations of CR

- High overhead (performance, storage space, etc.)
- Studies estimate that future Exascale systems could have a MTBF smaller than the time required to complete a CR process.
- CR is a reactive method, it will remedy the fault after the fact that your workload had failed.

#### **Proposed Solution**

Proactively predict failures, then remedy the situation before failure occurs, without impacting performance.







#### **Proposed Solution**

Design a container-based proactive fault tolerance framework to improve the sustainability of running workloads on Linux HPC clusters.

- ► The framework mainly serves 2 objectives:
  - Predict potential compute node hardware failures.
     (not the scope of this presentation, but detailed in PhD thesis)
  - Remedy the situation once faults predicted, with minimal overhead on the running HPC workloads.
     (The focus of this presentation)





#### Remedy Environment

#### **Container Technology:**



- We propose using the Linux container technology to perform workload migrations once failures are predicted
- Allows us to self-contain HPC application and its required libraries
- Reduces the coupling of the workload from physical hardware
- Proven its success as a scalable and lightweight technology for micro services used in large scale data centers
   (e.g. Google's data centers run most of their micro services on containers)
- We adapt potential resilience capabilities of containers towards HPC workloads

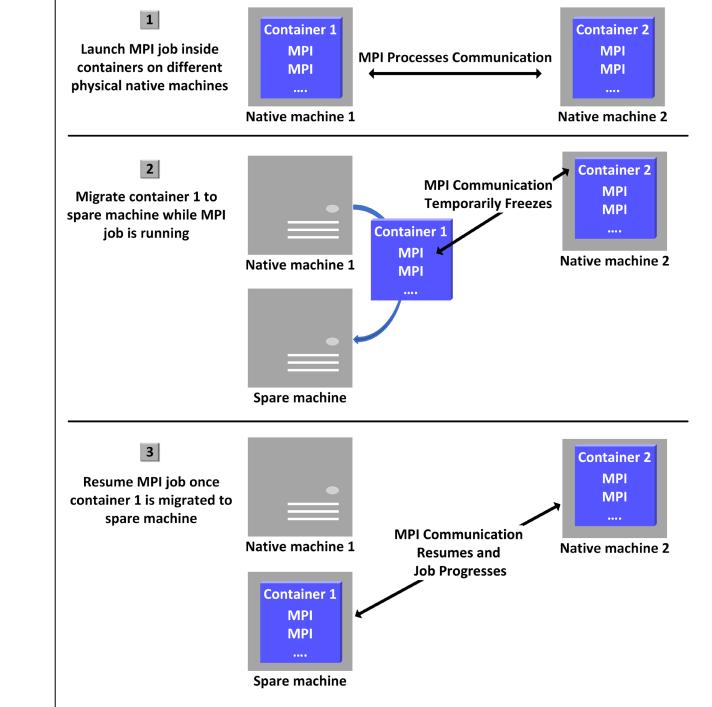
## **Work Summary**

#### Objective - Remedy Once Faults Predicted:



- ► In summary:
  - We setup a complete HPC environment that is container-based.
  - Tested it with 6 real HPC applications.
  - Applications use Message Passing Interface (MPI) de facto standard for HPC.
  - We were able to successfully do container migration for all HPC applications (after resolving numerous technical challenges).
  - Performed comprehensive performance benchmarks comparing container vs. native. Container performance was almost native.

#### Concept of Migration

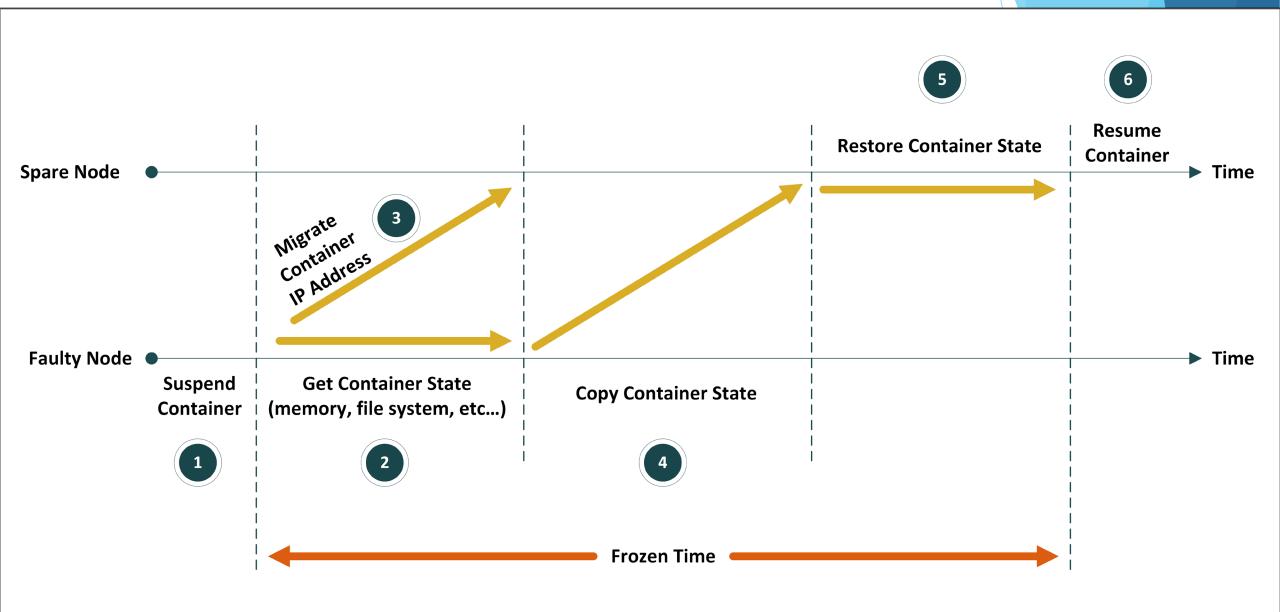


#### Migrating Containers

- CRIU Open-source Library:
- A tool that can be used to freeze/unfreeze processes running on Linux in user space
- May be applied to freeze/unfreeze containers
- At time of testing, was still beta with Linux RedHat 7 (no official support, buggy)
- Had to debug and modify some of the library's source code to work in our HPC environment (code modification to fix issue with NFS mounts inside containers).



# Migration Steps



## Testing Real HPC Applications in Containers

#### Applications use MPI, no need to modify code or binary executable

Application	Main Developers	Language	Domain	Visualization
OSU Micro-Benchmarks	Ohio State University	С	MPI benchmark for network bandwidth and latency	NA
Palabos	- Academic: University of Geneva - Industry: FlowKit CFD	C++	CFD/Complex Physics using lattice Boltzmann method	Post-process In-situ
Flow	Open Porous Media Initiative (Oil companies + Academia)	C++	Reservoir simulation	Post-process
Fluidity	Imperial College London	Fortran, C++	<ul> <li>CFD solving Navier-Stokes</li> <li>Geophysical fluid dynamics</li> <li>Ocean Modelling</li> <li>Adaptive unstructured mesh</li> </ul>	Post-process
GalaxSee	<ul> <li>Shodor Education Foundation</li> <li>National Center for</li> <li>Supercomputing Applications</li> <li>George Mason University</li> </ul>	C++	N-body galaxies simulation	In-situ
ECLIPSE	Schlumberger (commercial)	Fortran	Industry-reference reservoir simulator	Post-process

# Testing Real HPC Applications in Containers

- Test using various AWS hardware platforms (# cores, memory, network):
  - Low spec nodes: (4 physical cores, 32 GB RAM, 1 Gig network)
  - Med spec nodes: (18 physical cores, 72 GB RAM, 10 Gig network)
  - High spec nodes: (32 physical cores, 256 GB RAM, 25 Gig network) (36 physical cores, 512 GB RAM, 25 Gig network)
- Test migration with various MPI libraries: MPICH, Open MPI, Intel MPI
- MPI job sizes ranged from 4 to 144 processes.

#### Testing Real HPC Applications in Containers

Important questions to answer during container testing:

- 1. Will application performance be impacted? (container vs. native)
- 2. Can we actually migrate containers with MPI processes without affecting HPC job?
- 3. Will produced results be intact?(no data corruption due to migration)

#### **Application Performance Summary**

- More than 130 test runs were performed for the study using the various HPC applications and hardware platforms.
- OSU benchmarks (point-to-point and collective):
  - Network latency overhead was ~6.8% on average with containers.
  - Network bandwidth overhead was ~3.9% on average with containers.
- However, performance overhead of the real HPC applications was very negligible and close to native performance (%0.034 on average).
- Worst case application performance overhead was %0.9
- Overall, the performance on containers was acceptable for all HPC applications tested (almost native).



# Migration!

## Migration Behavior

- Average container migration time was 34 seconds (using standard SSD disk)
- Migration time mainly influenced by size of app binaries stored inside container.
- In the case of Palabos, GalaxSee, and ECLIPSE, application binaries were stored on shared NFS storage and not inside container.
- When testing with 10G/25G network, migration time was still the same!
- Bottleneck is not network, but speed of local hard disk storing container data.
- Testing with faster SSD hard disk reduced avg migration time to 22 seconds.

<b>Container Application</b>	Migration Time (seconds)		
Fluidity	50		
Flow	35		
Palabos	30		
GalaxSee	29		
ECLIPSE	26		

#### Migration Behavior

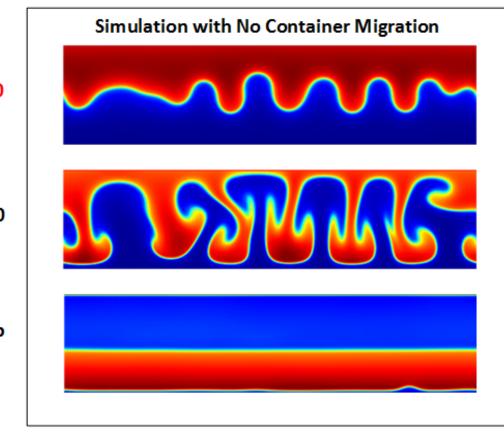
#### Example of checking results integrity:

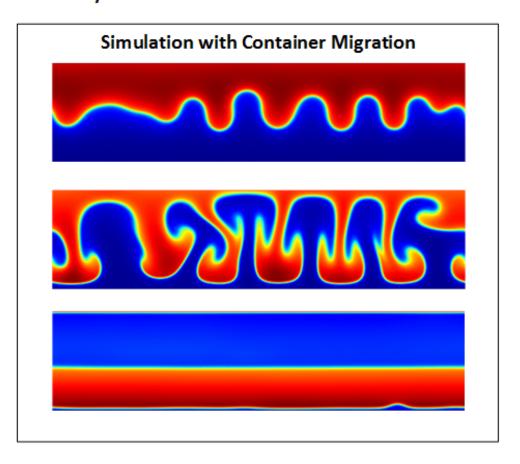
Results Integrity Check After Container Migration at Timestep 2400 (Palabos HPC Simulator)

Timestep 2400

Timestep 3000

Final Timestep





#### Demos (available on YouTube)

Palabos: Migrate container while MPI/visualization job is running.

More YouTube demo scenarios for the various applications tested are available in paper and PhD thesis.

Application	Demo Video Link
Palabos	https://youtu.be/1v73E2Ao3Mk

## Main Contributions & Summary

- 1. To the best of our knowledge, this work is the first in the HPC domain to demonstrate successful migration of MPI-based real HPC workloads using containers and CRIU.
- 2. Performed comprehensive performance benchmarks on containers using real HPC workloads on multiple computing platforms.
- 3. Using containers in HPC is a young topic, the challenges we faced and the solutions adopted are valuable experiences to share with the HPC community.

# Thank You!

# Questions?

