



UNMANNED SYSTEMS LAB



TEXAS A&M UNIVERSITY

J. Mike Walker '66 Department of
Mechanical Engineering



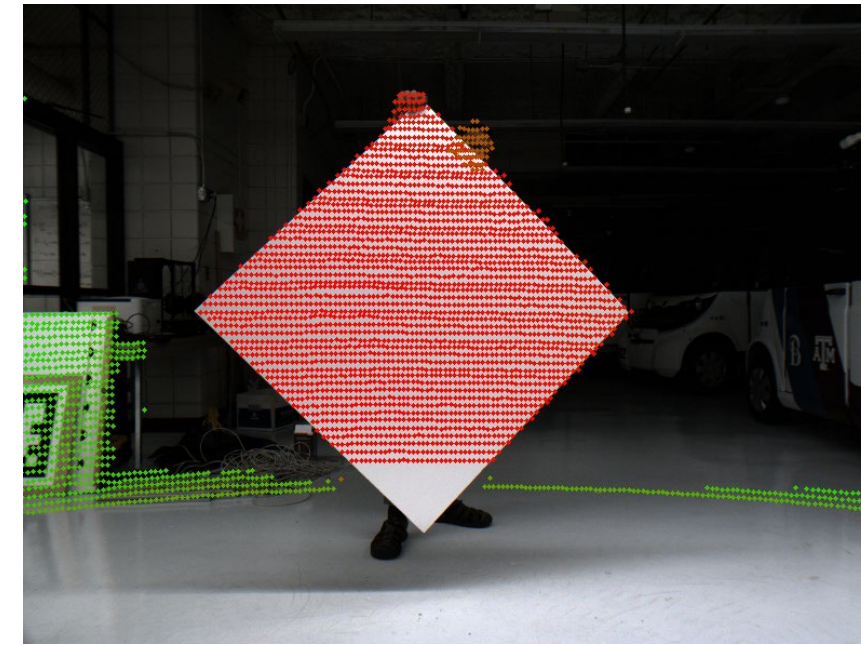
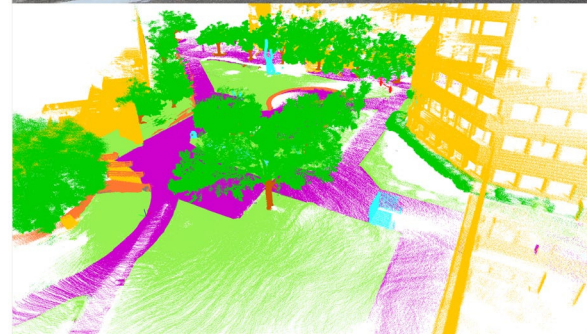
An Integrated Modeling and Testing Architecture for ROS Nodes

Jacob Hartzler

ROSCON 2023

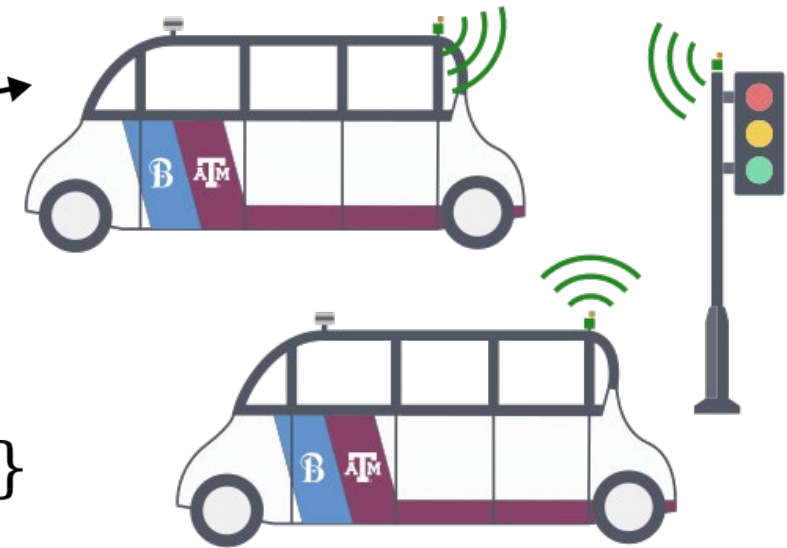
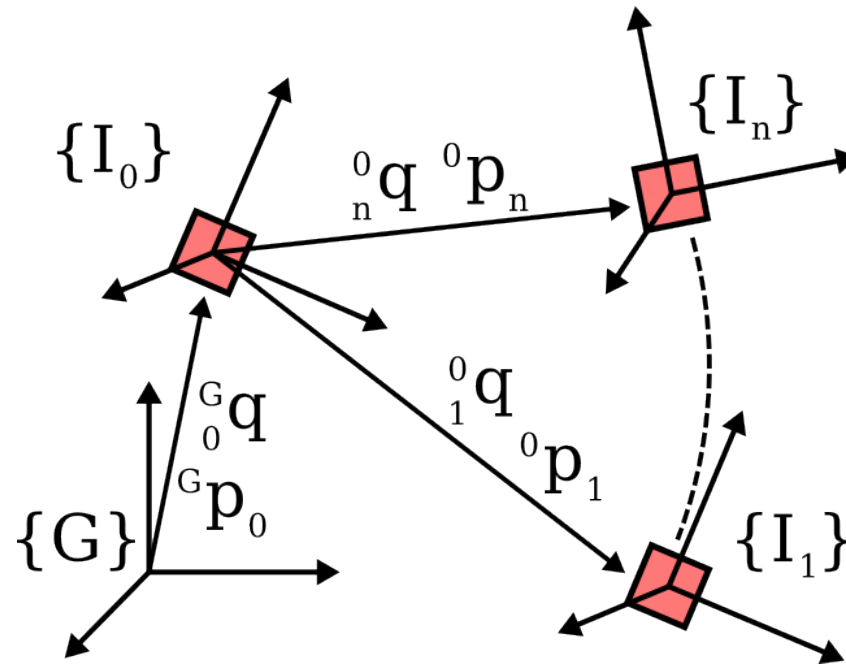
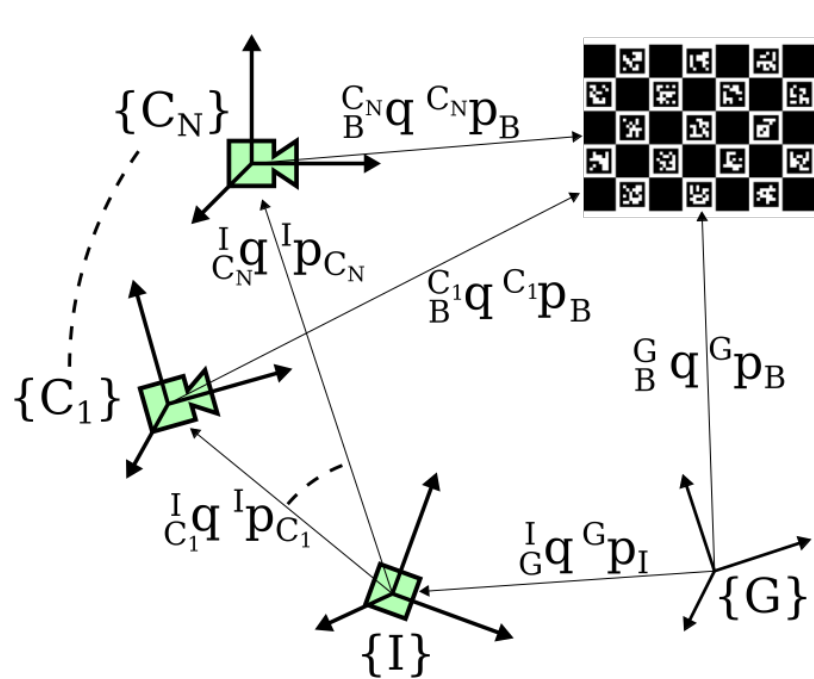
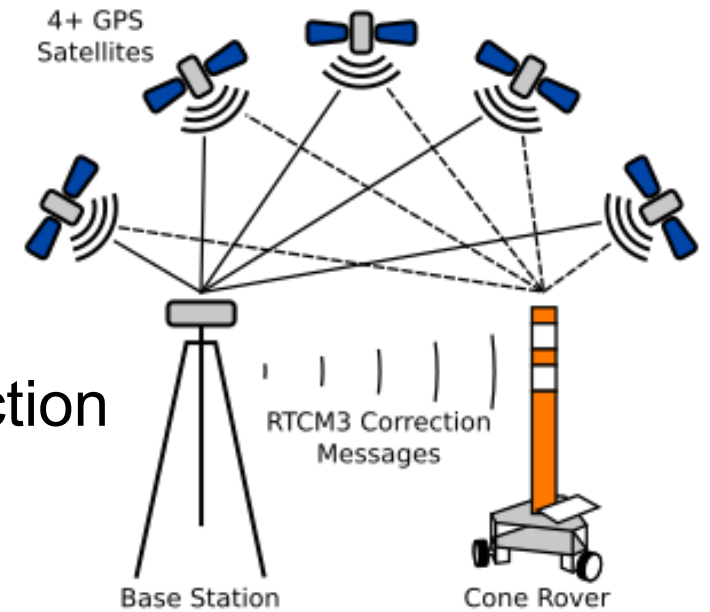
Unmanned Systems Lab

- On/Off Road Autonomy
- LIDAR and RADAR Odometry and Perception
- Multi-Sensor Fusion and Calibration



Our Work - Calibration

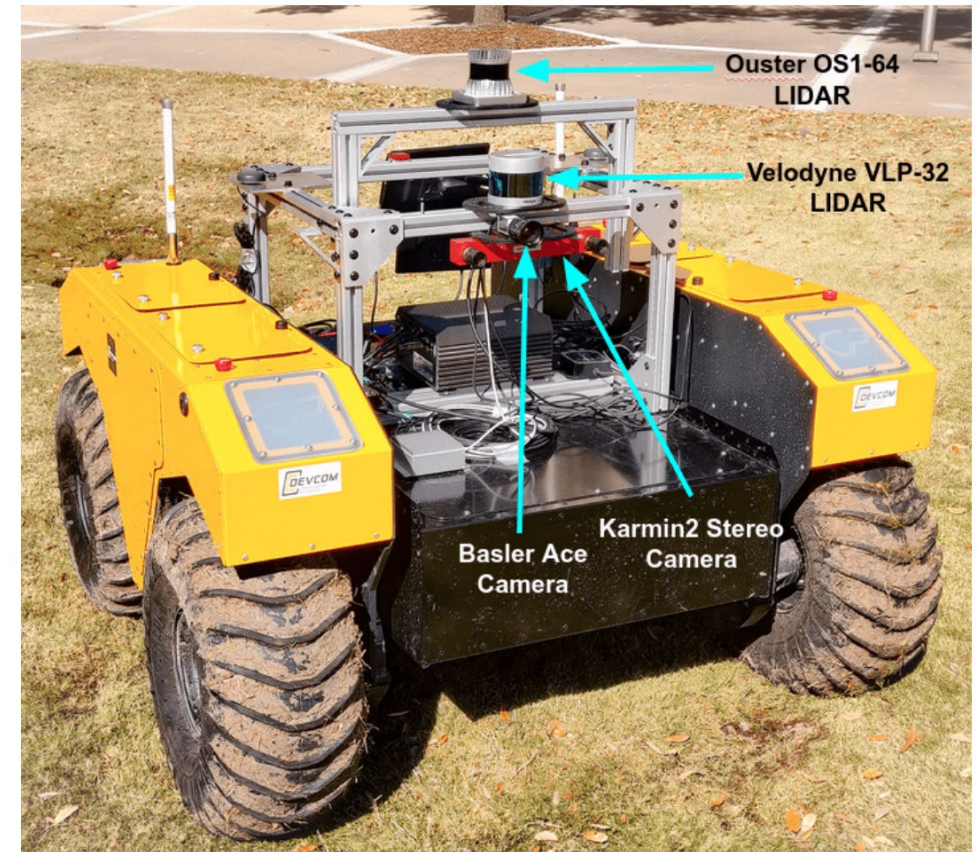
- Utilize a wide variety of sensors and environments
- RTK GPS in highway applications
- Ultra-Wideband for cooperative localization and detection
- Multi-IMU, Multi-Camera fusion



The Challenge of Calibration

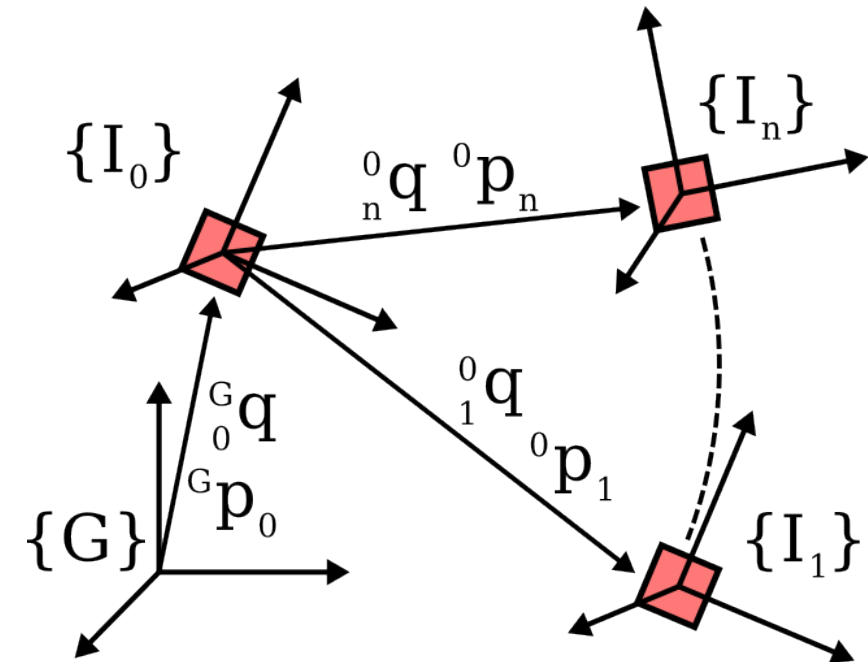
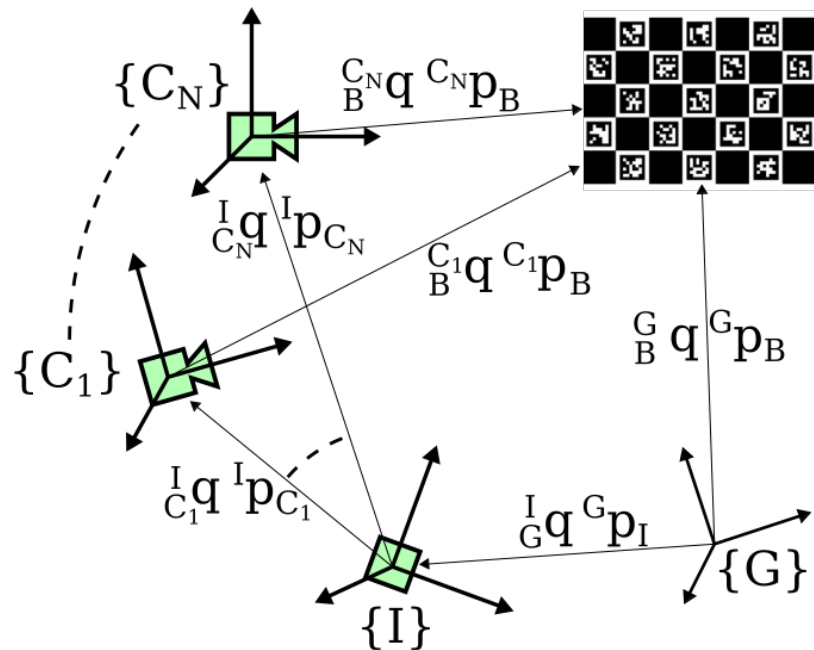
- Sensors do not work well when uncalibrated
- Ideally would like to calibrate sensors *online*
- Re-calibrating on the fly is more robust
- Proving stability and consistency is hard
- Sensor flexibility makes it even harder

**Must test many, many
datasets and environments**



EKF-CAL Package

- EKF-CAL is flexible MSCKF-based sensor calibration package
- Inputs are YAML based and compatible with ROS 2 parameter declarations
- Inherently multi-sensor (IMU, Camera, and GPS soon)
- Developed with integrated testing and Monte Carlo simulation in mind
- Open Source!

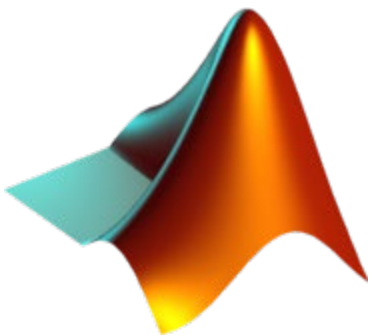


Typical Development

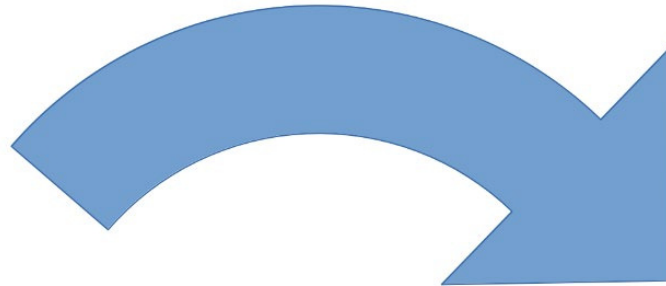


Logo for Julia, featuring the word "julia" in a bold, lowercase font with four colored dots (blue, green, red, purple) above the letters 'i', 'l', 'i', and 'a' respectively.

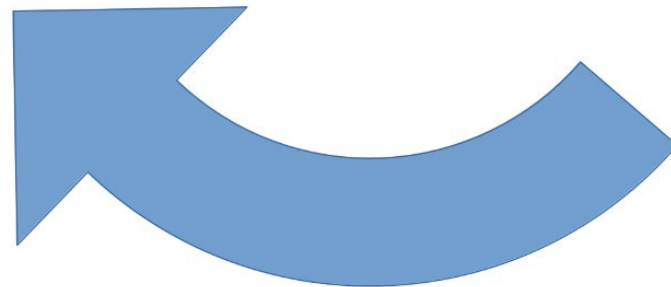
Develop Algorithm in
scripted language
(MatLab, Julia, etc.)



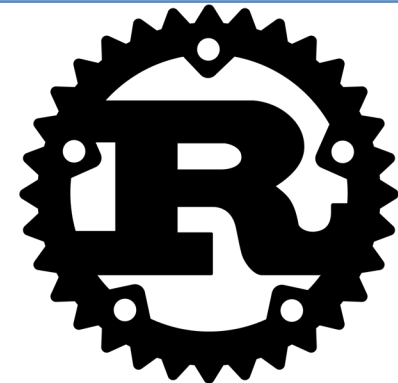
Need performance / Real-Time



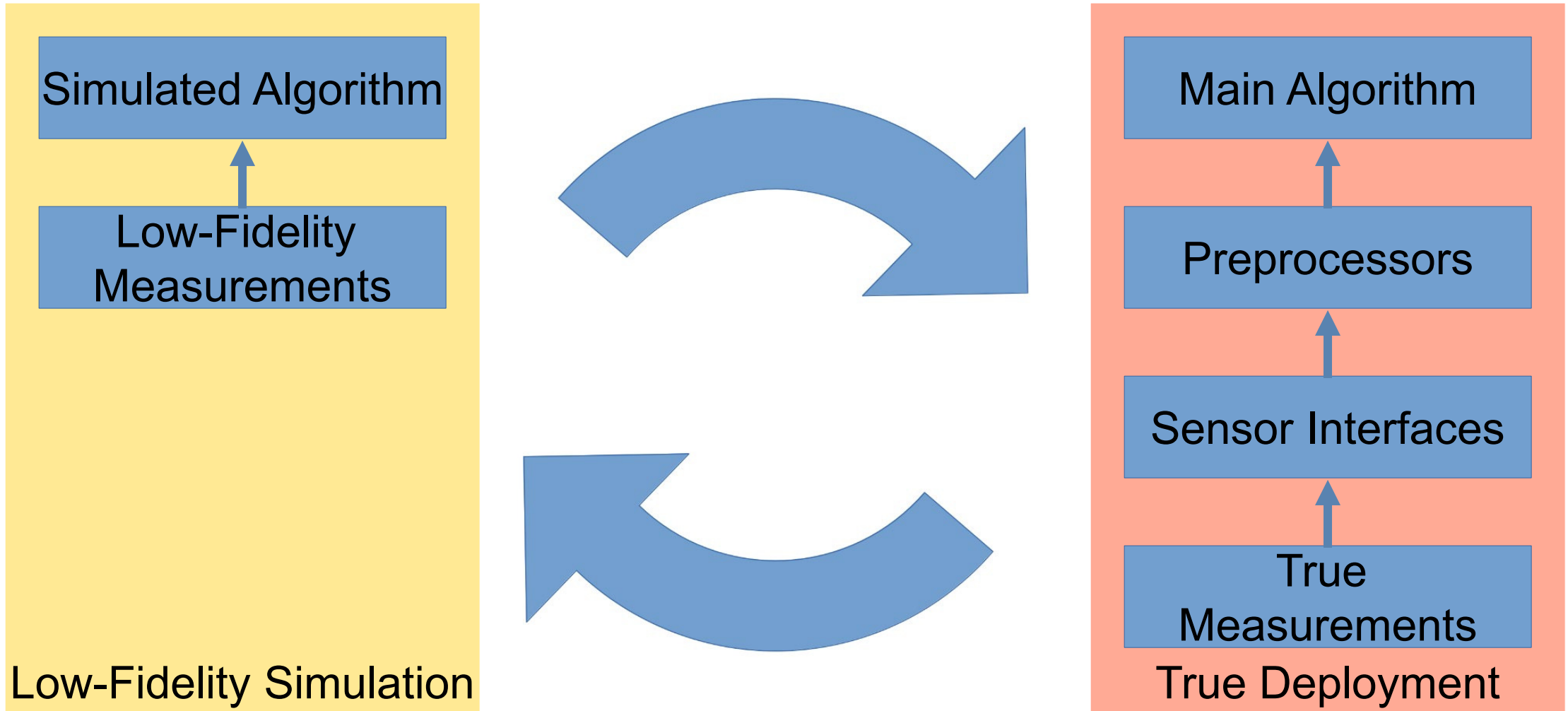
Deploy code in
compiled language
(C++, Rust, etc.)



Find a bug / invalid assumptions



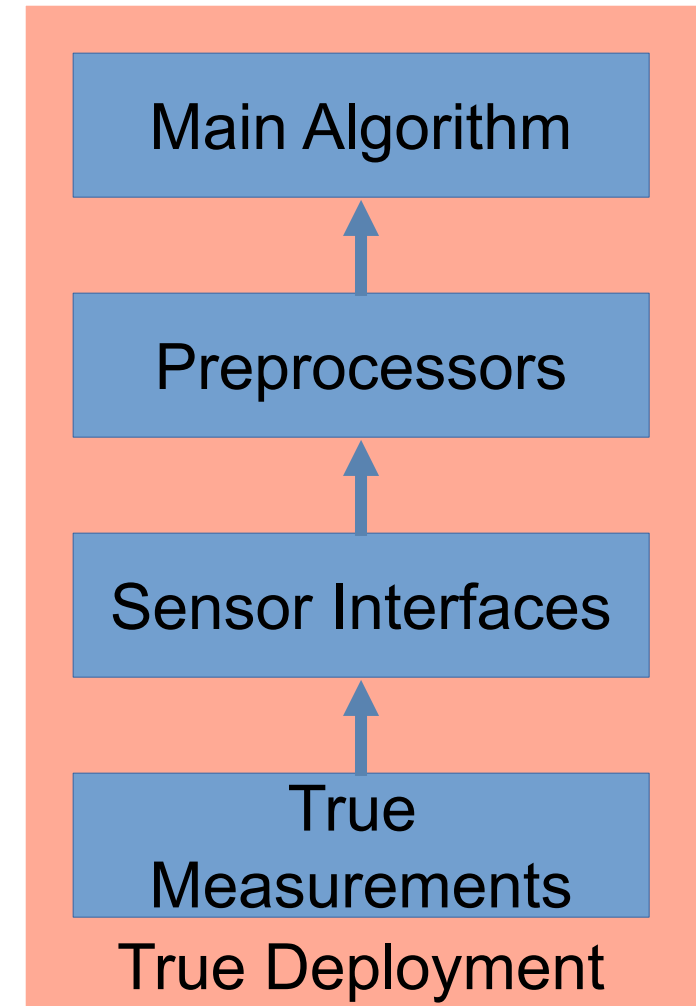
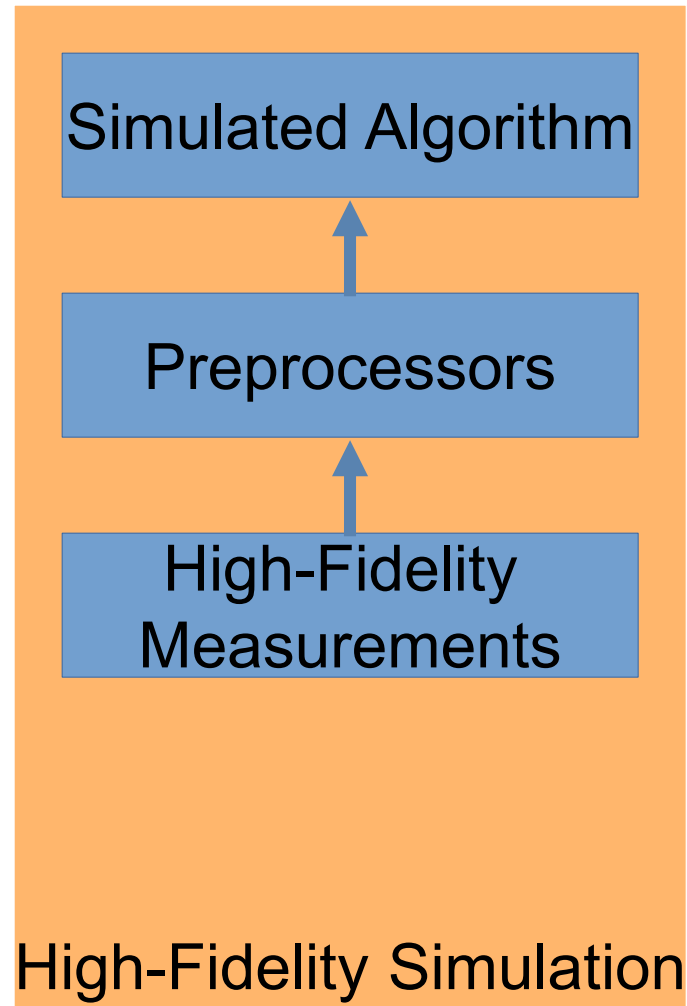
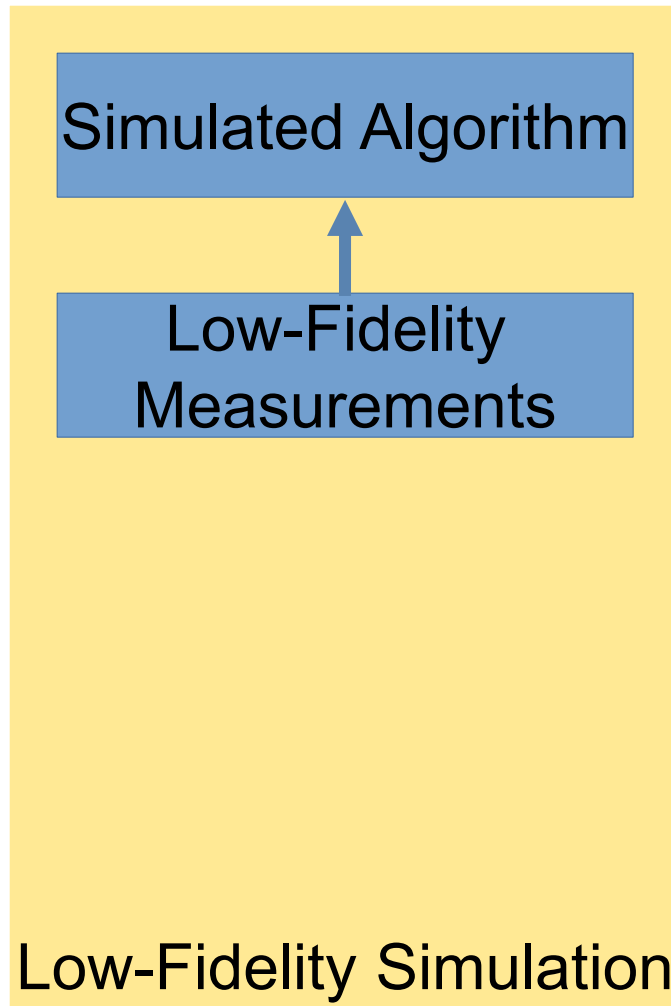
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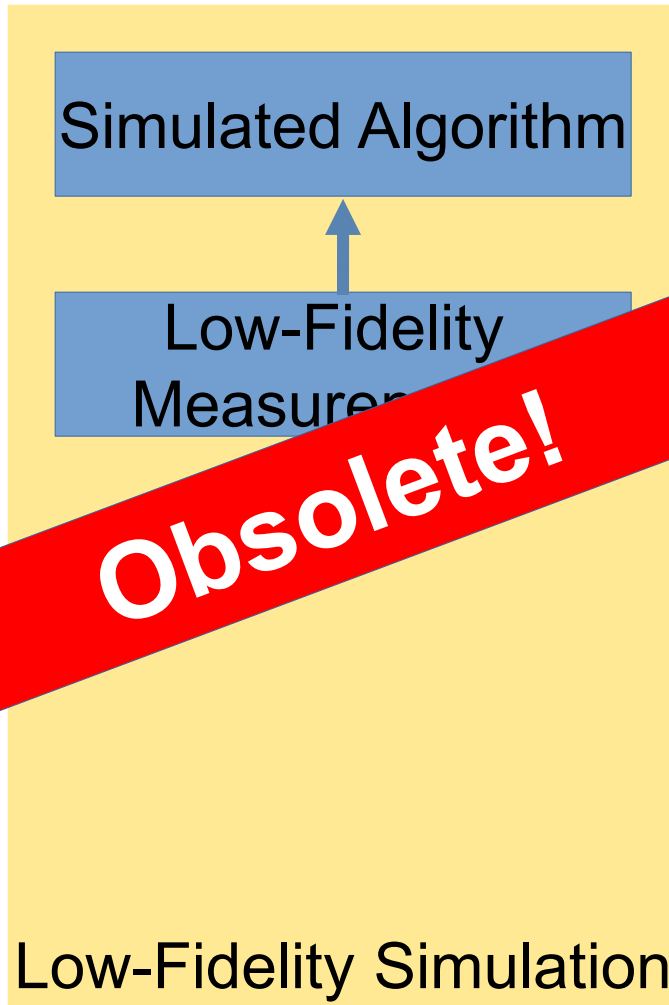
Typical Development



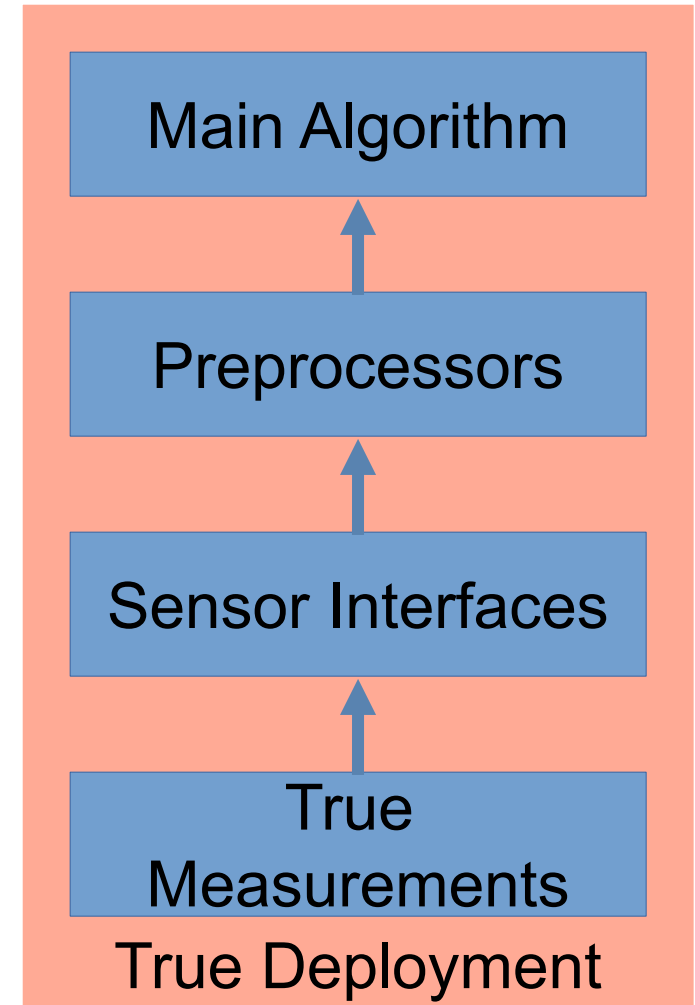
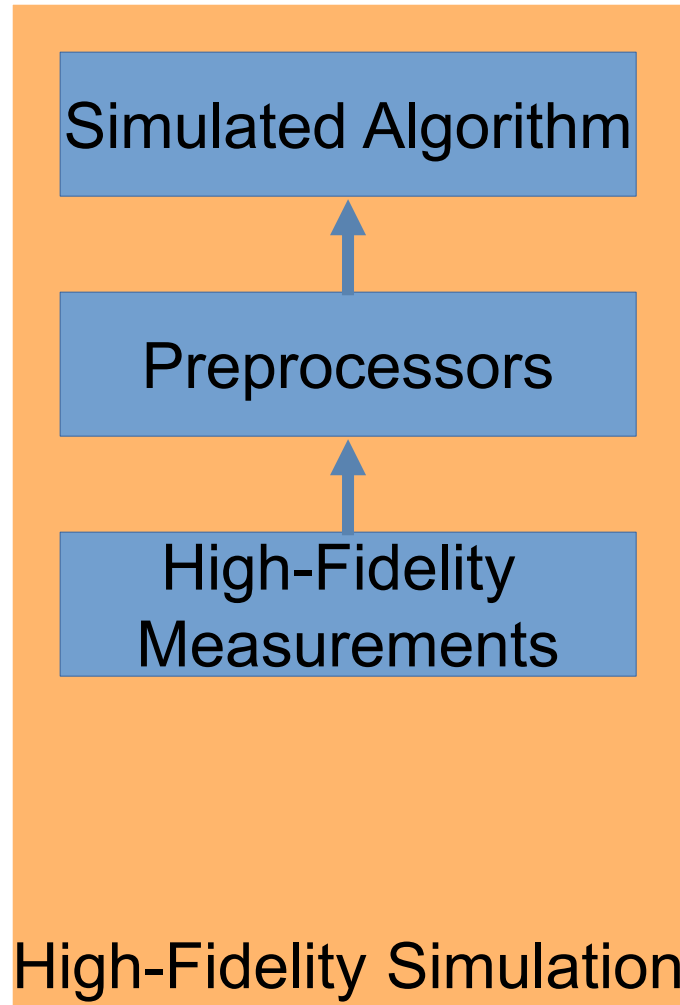
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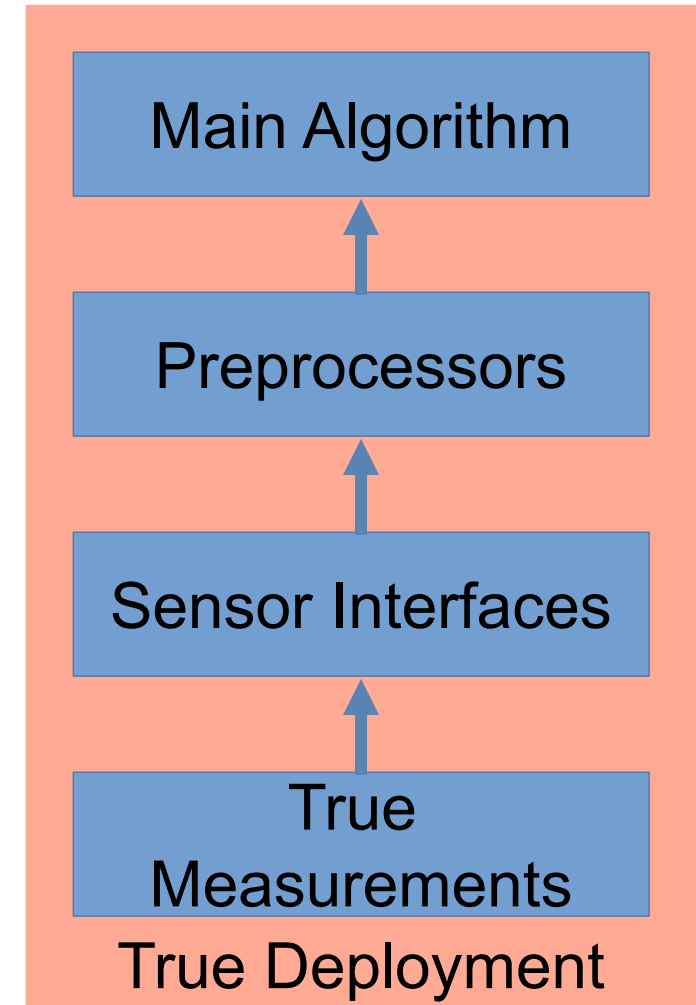
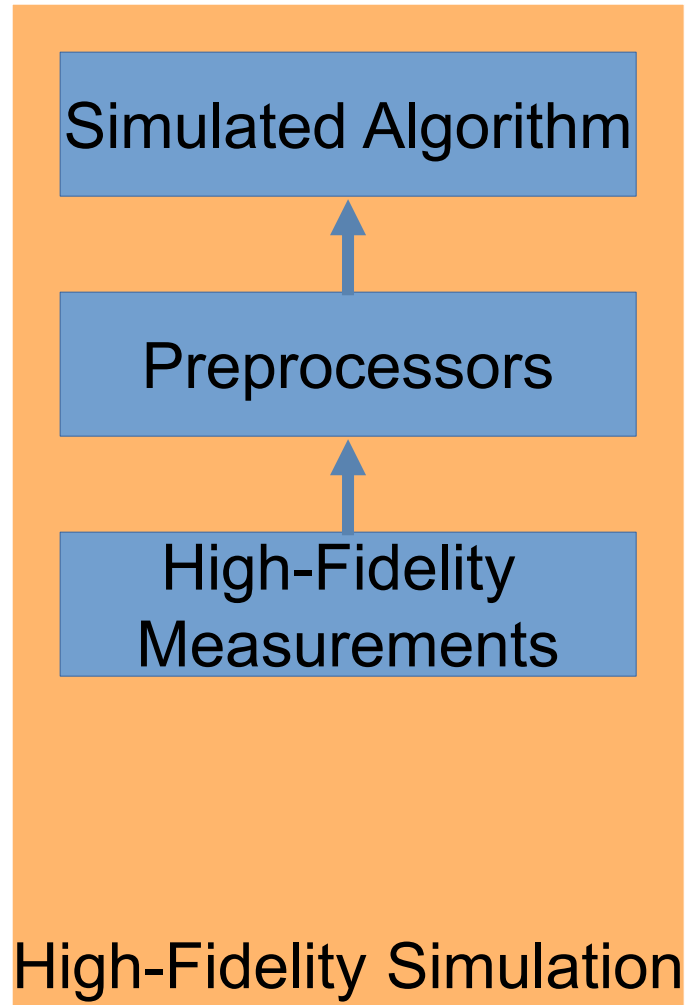
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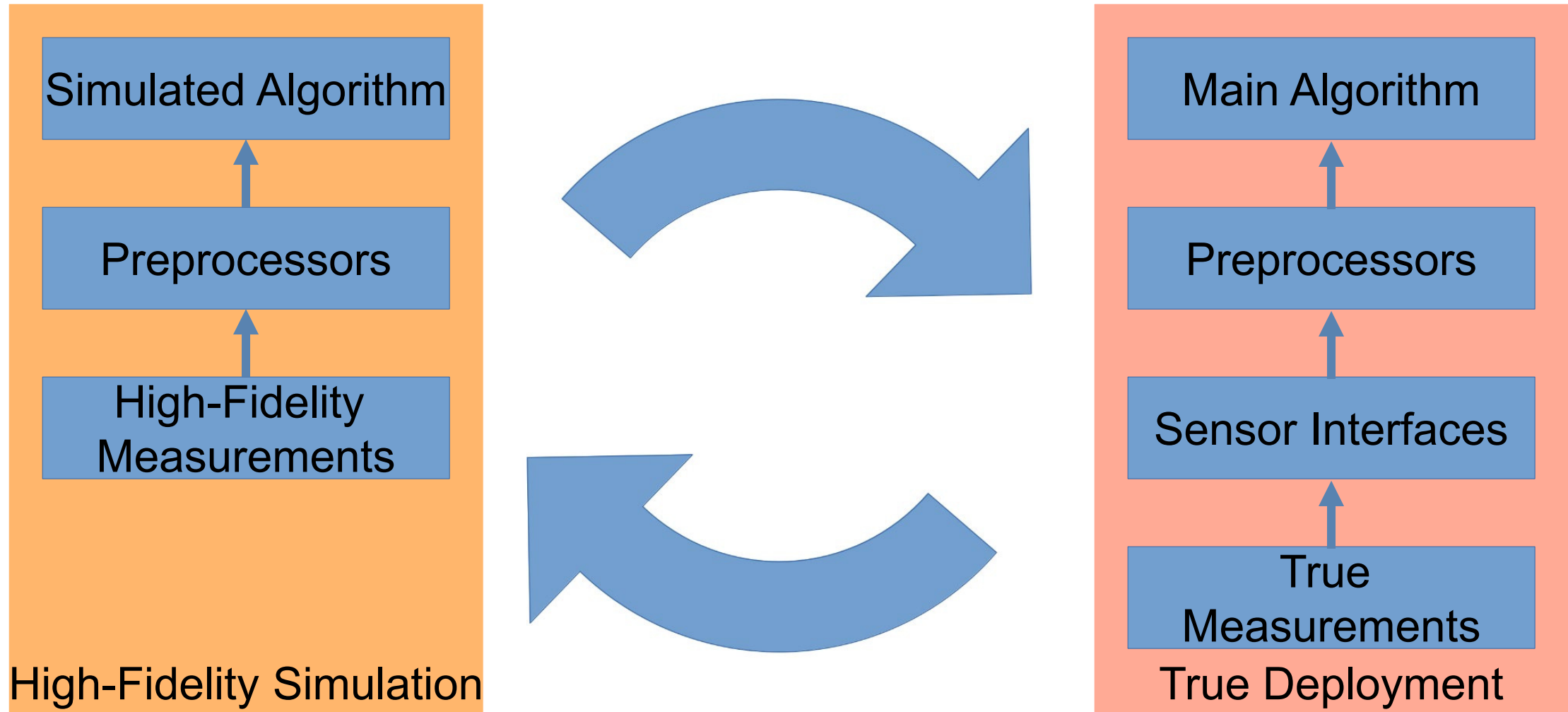
Obsolete!



Typical Development

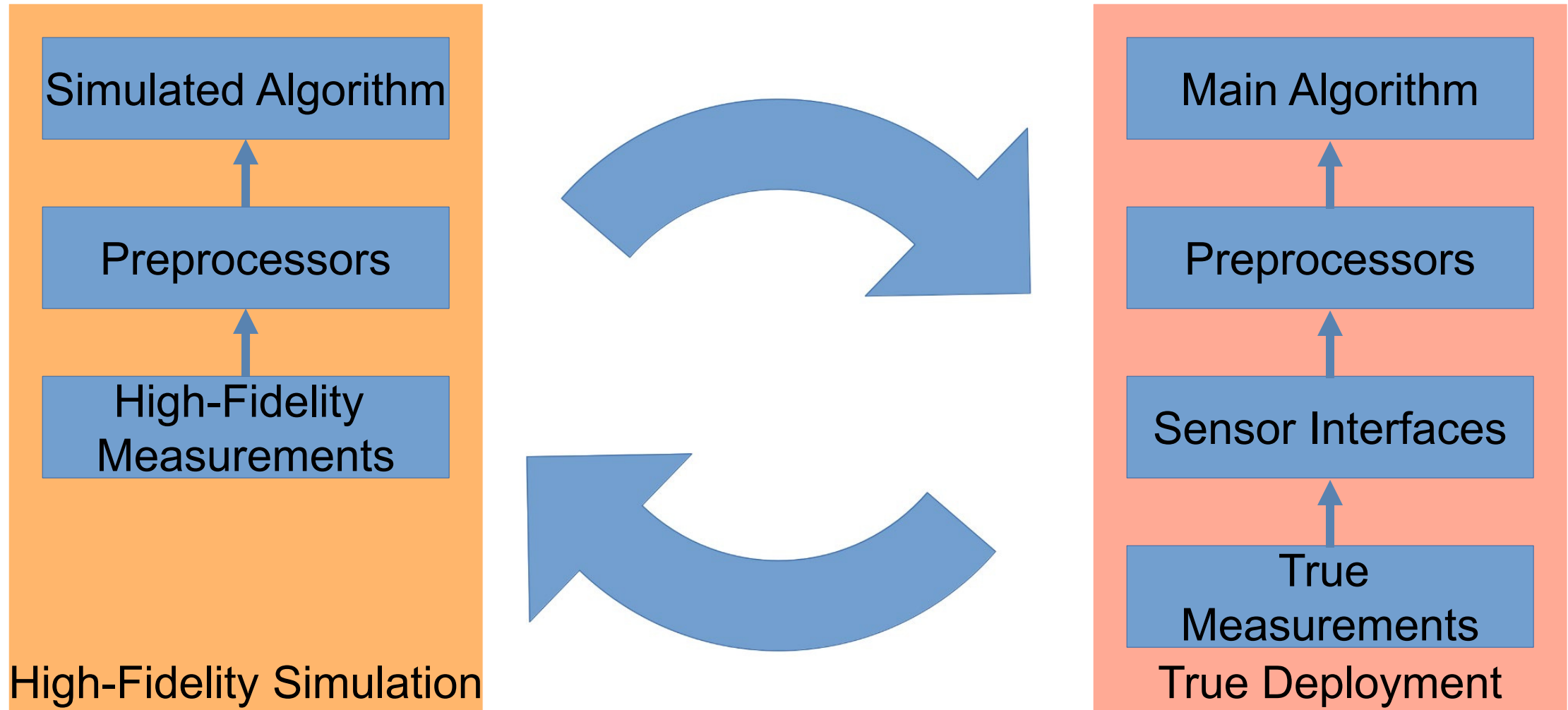


Typical Development



Typical Development

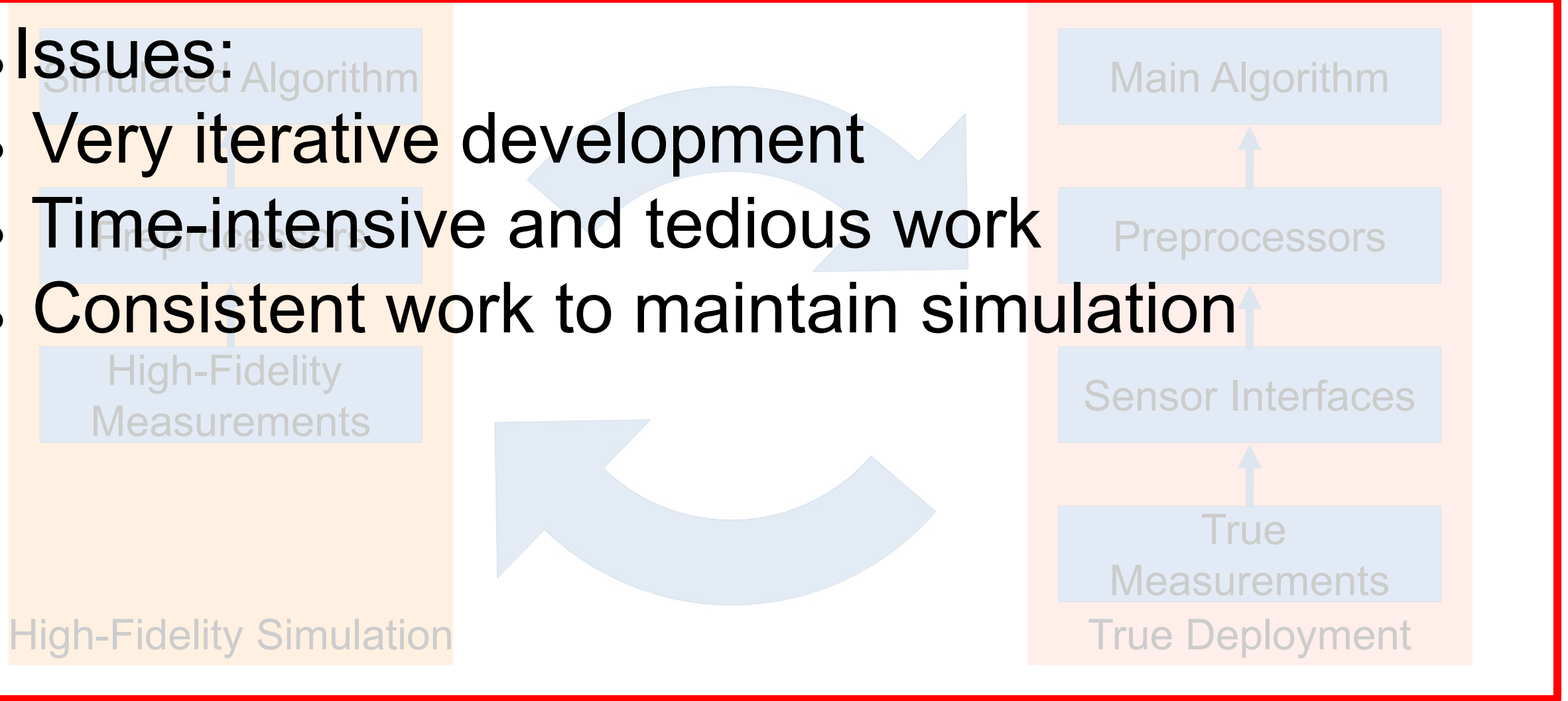
Functionally Identical



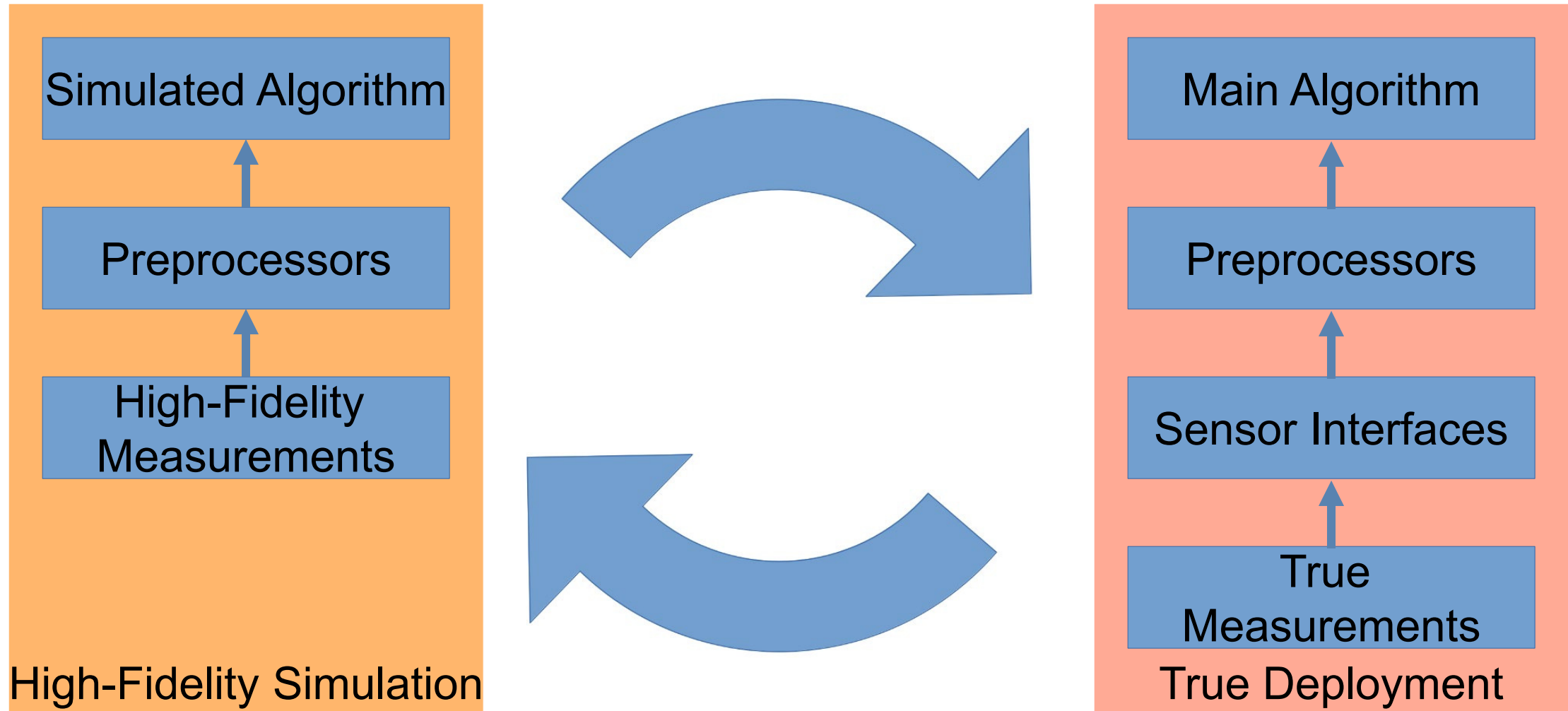
Typical Development

Functionally Identical

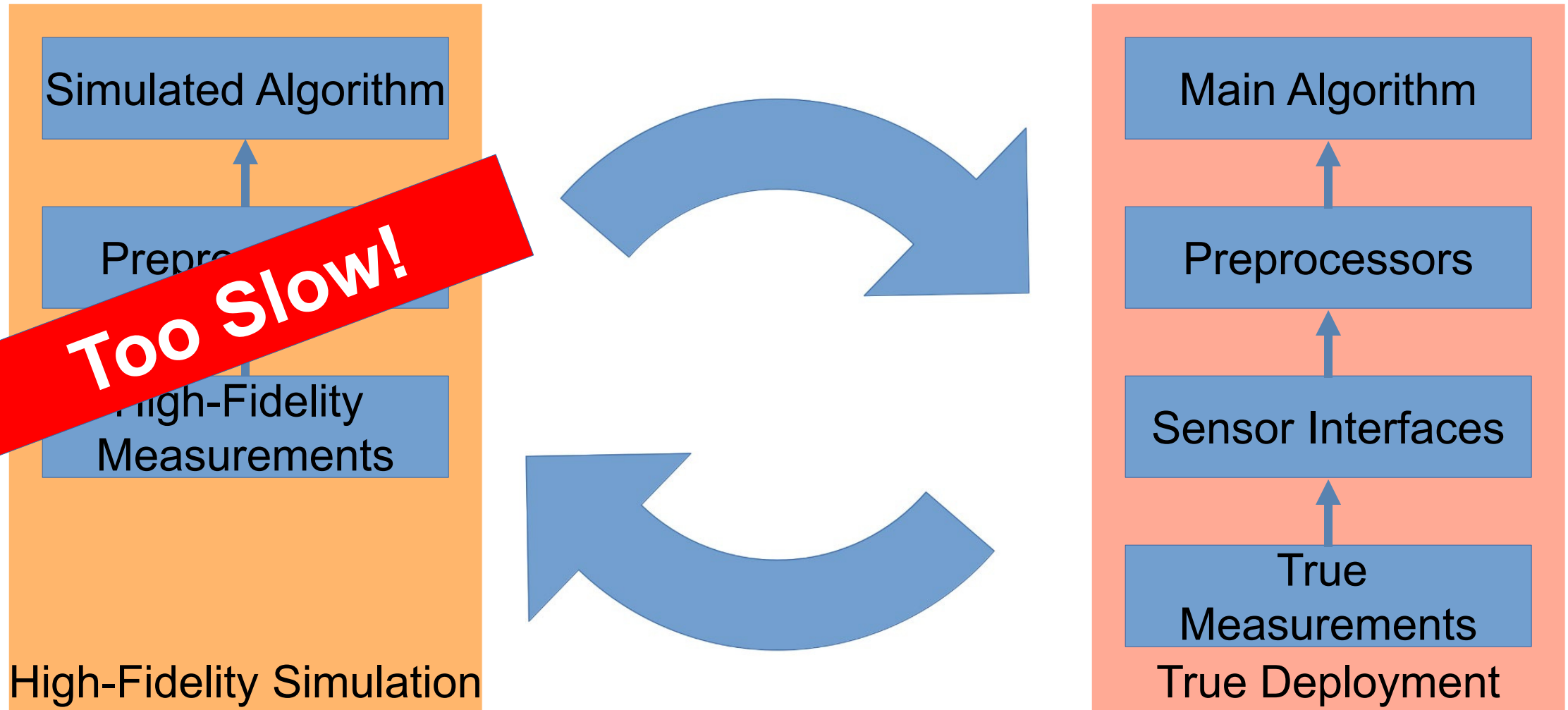
- Issues:
- Very iterative development
- Time-intensive and tedious work
- Consistent work to maintain simulation



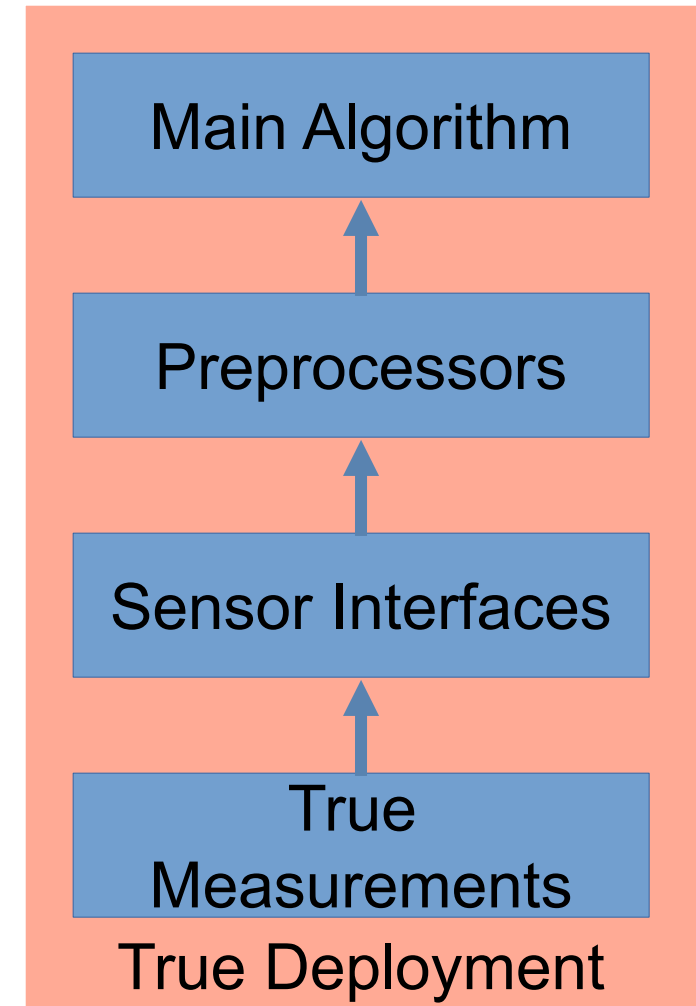
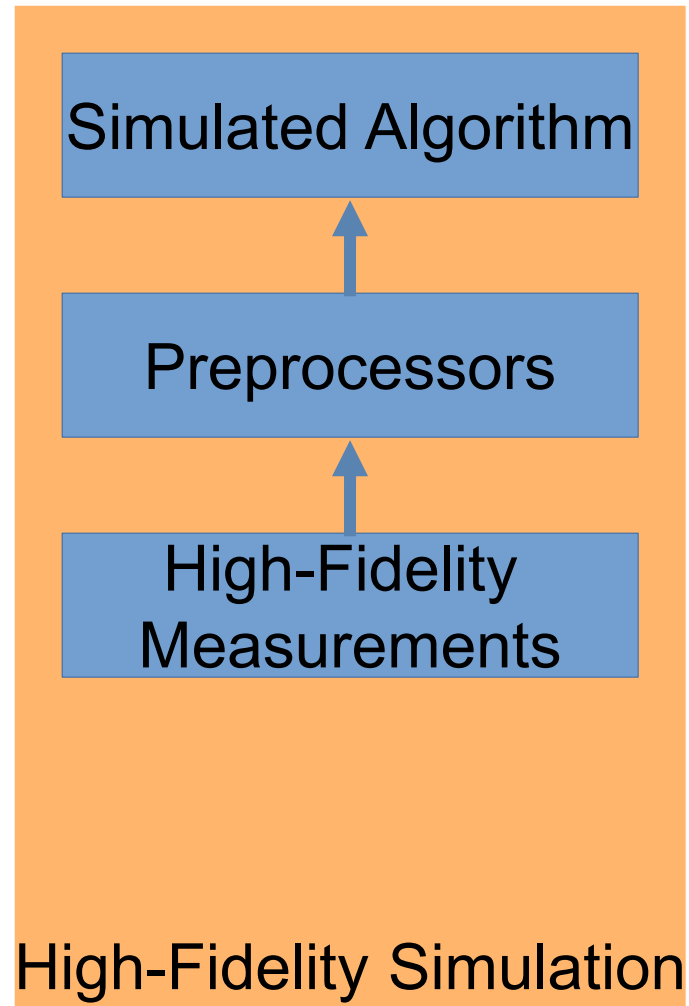
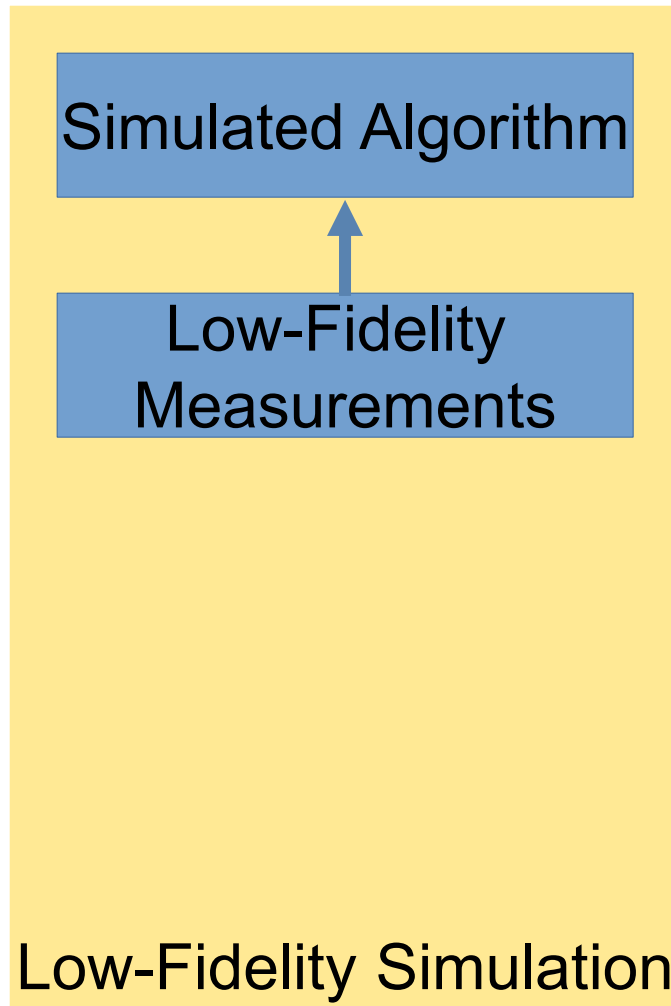
Typical Development



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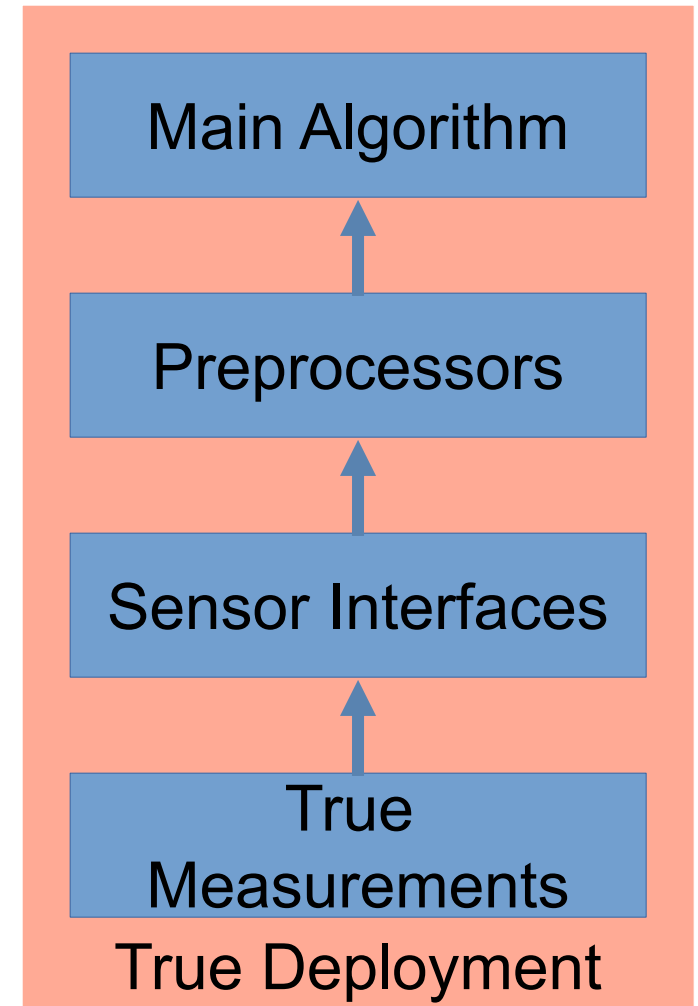
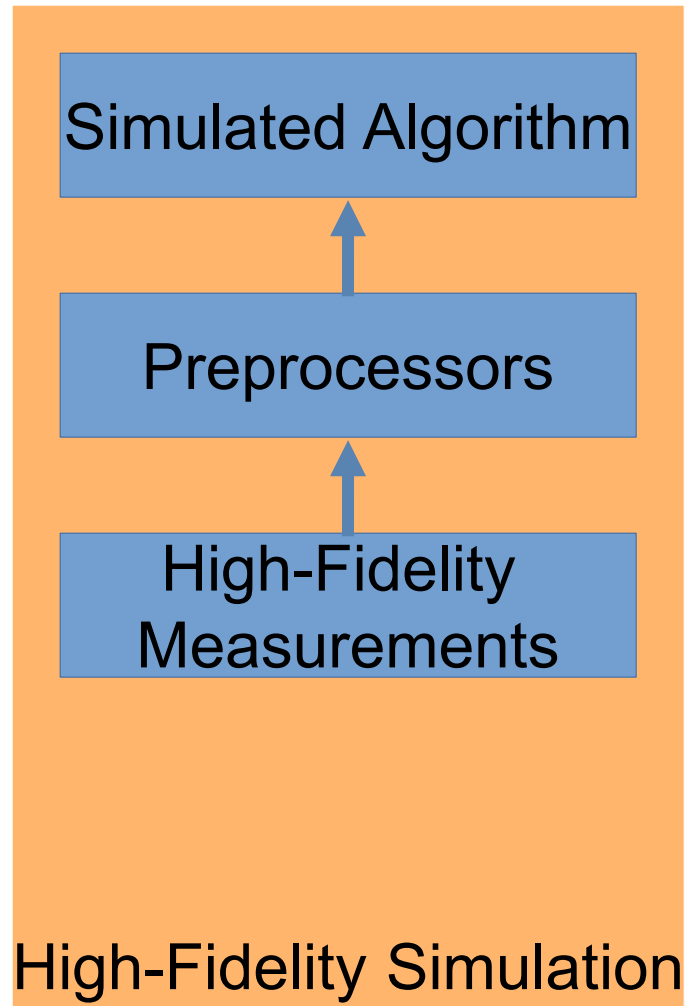
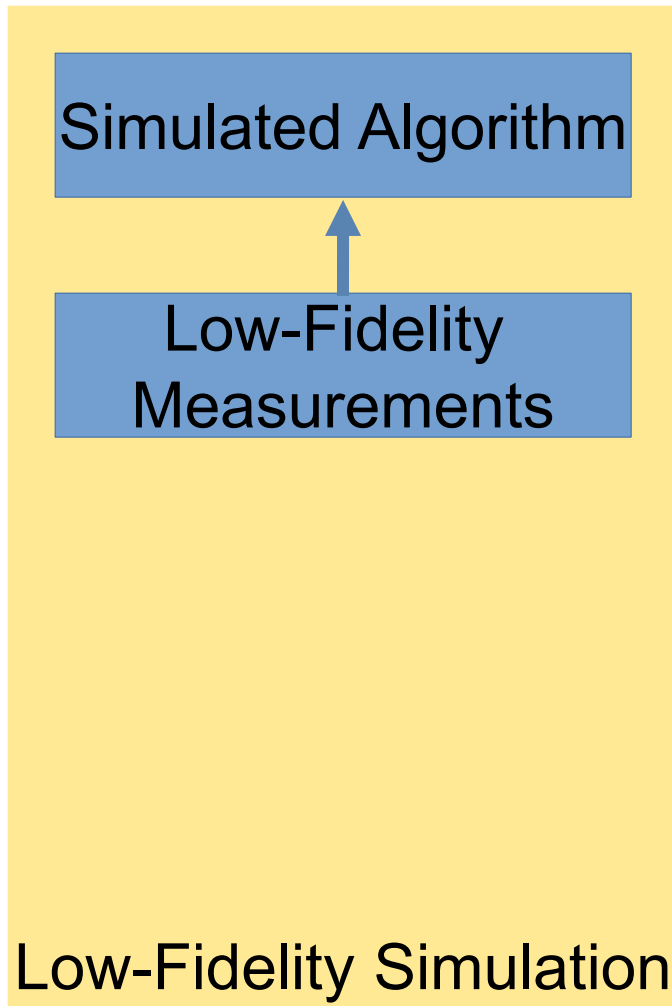


Typical Development



Typical Development

Three Code Bases



Typical Development

Three Code Bases

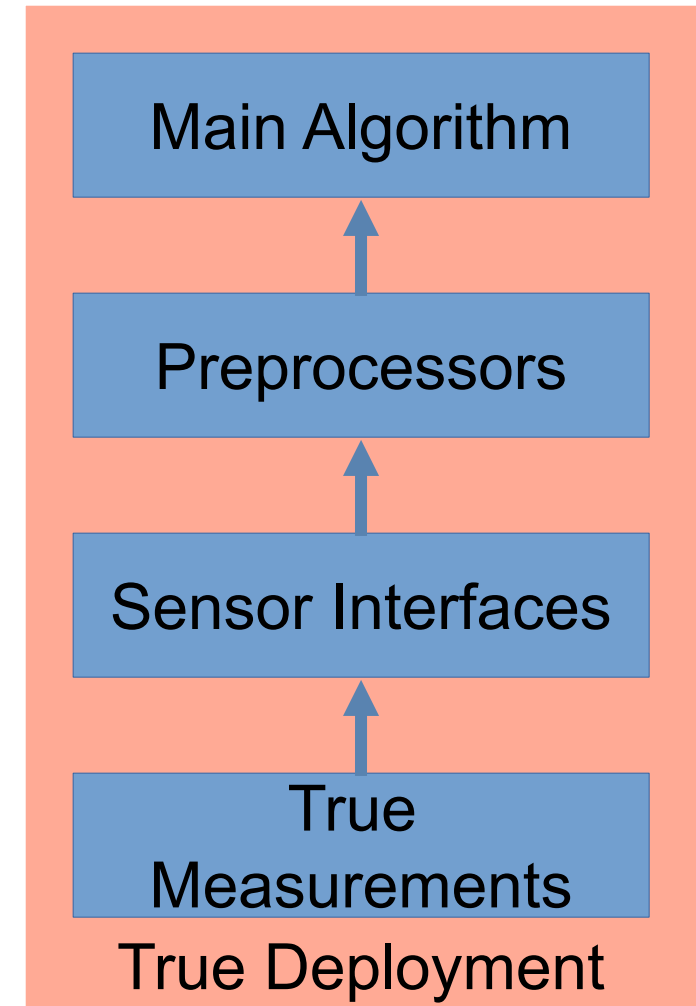
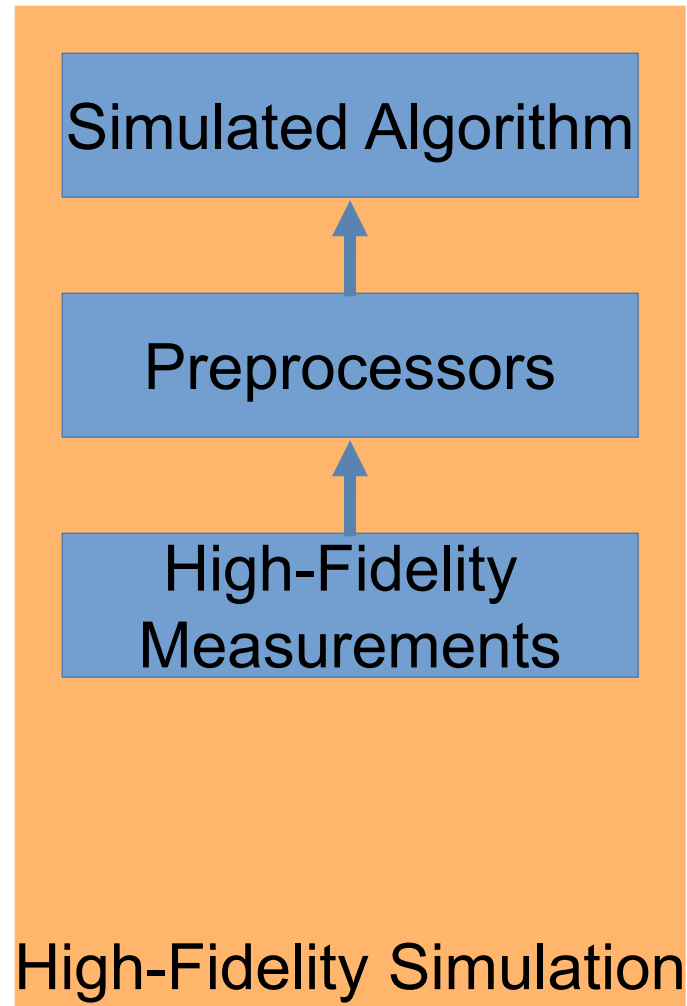
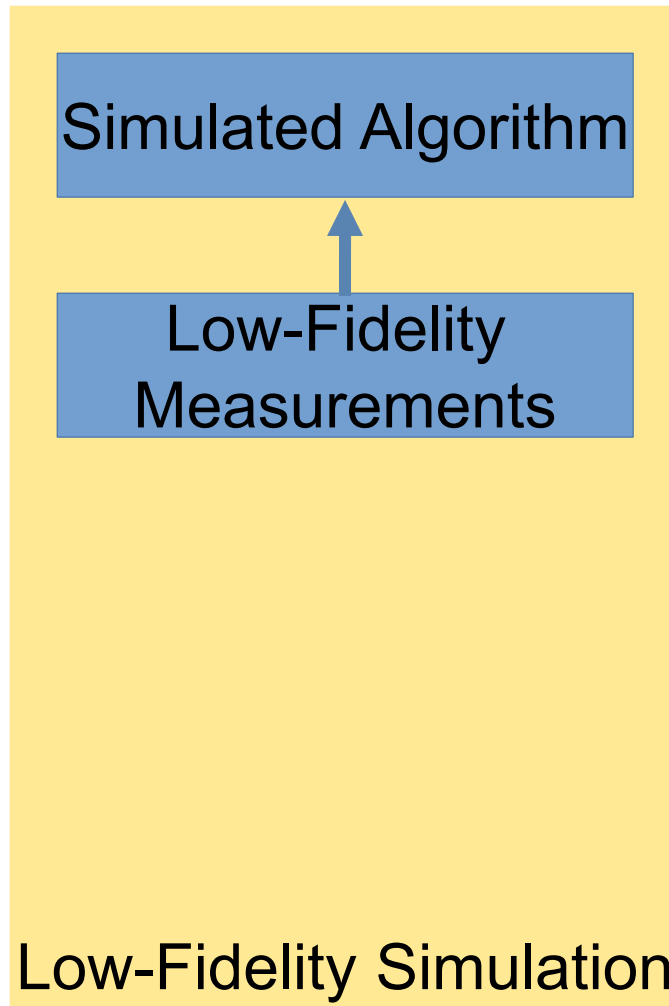
- Result:
- Complex simulations lead to fragmented code
- Fragmented code is expensive to maintain
- Multiple simulations leads to:
 - Uncaught bugs
 - Untested deployment code

Low-Fidelity Simulation

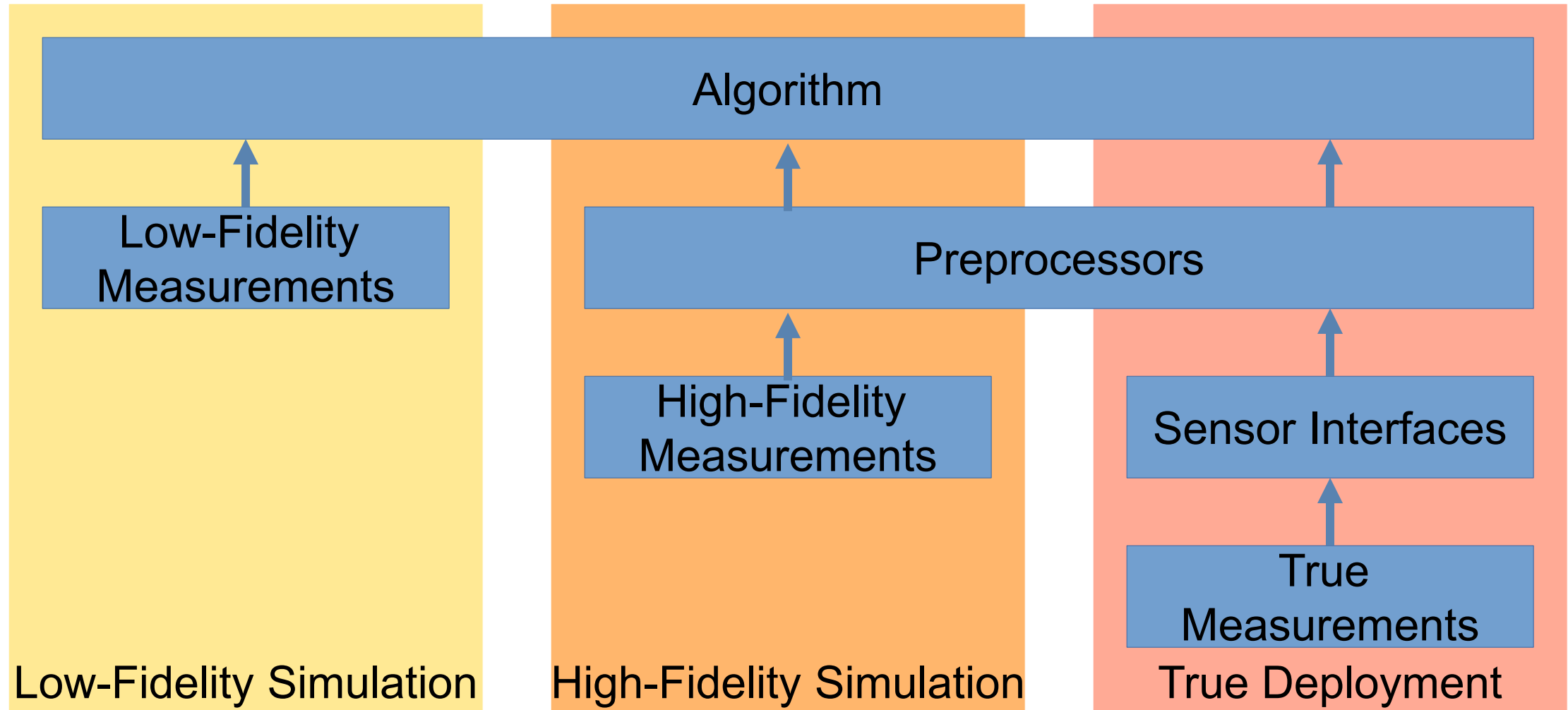
High-Fidelity Simulation

True Deployment

Typical Development

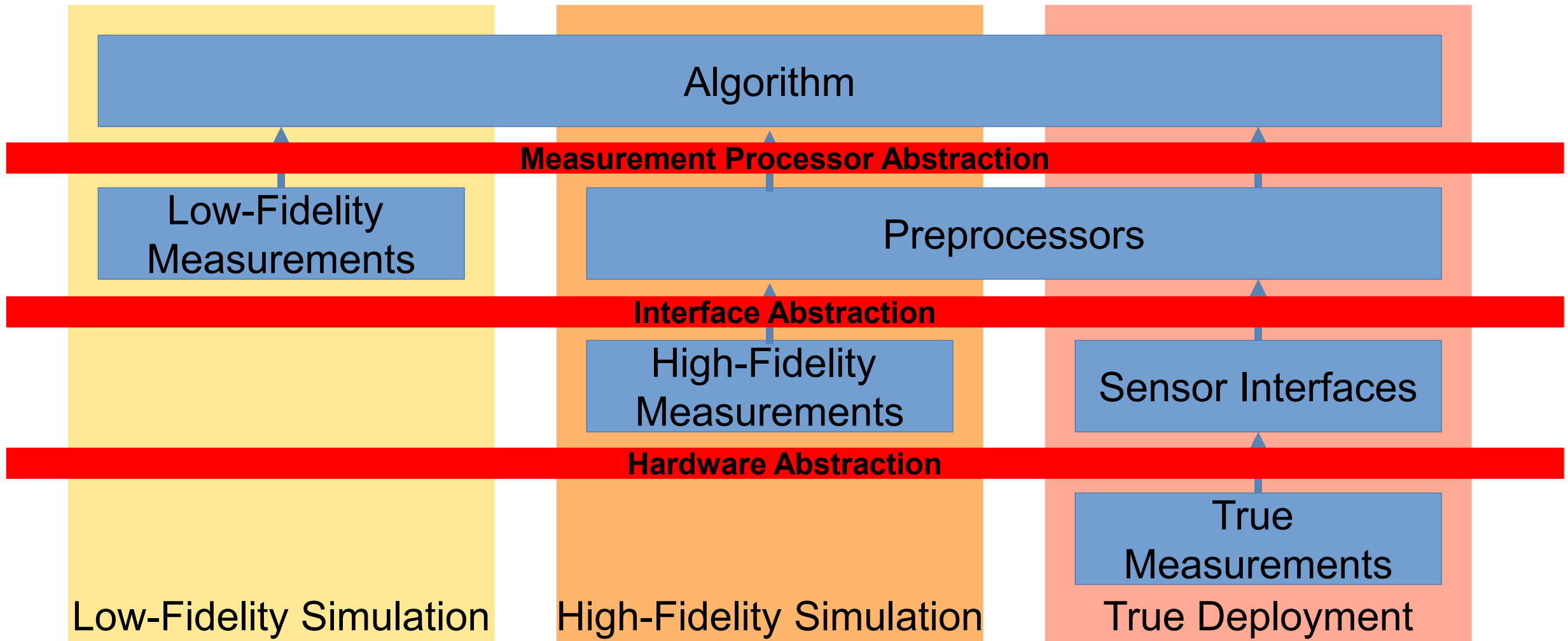


Proposed Solution

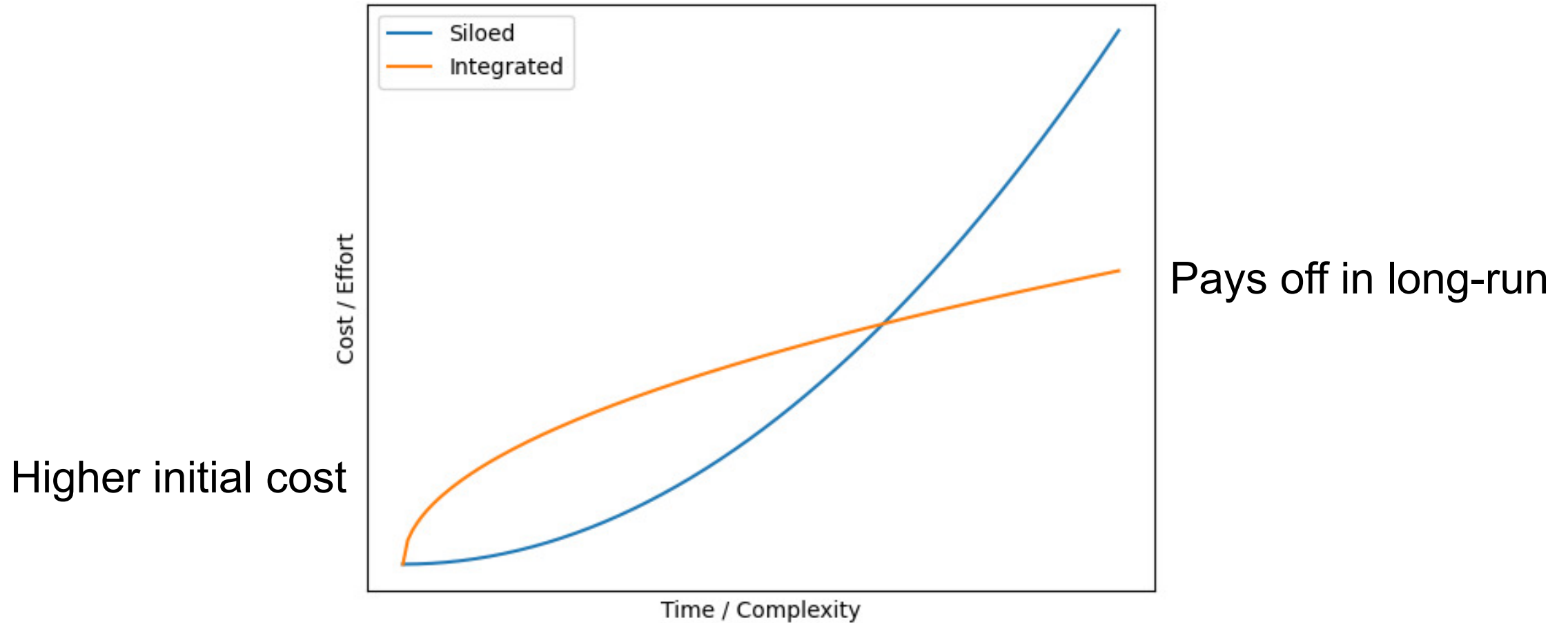


Proposed Solution

Abstraction Layers!

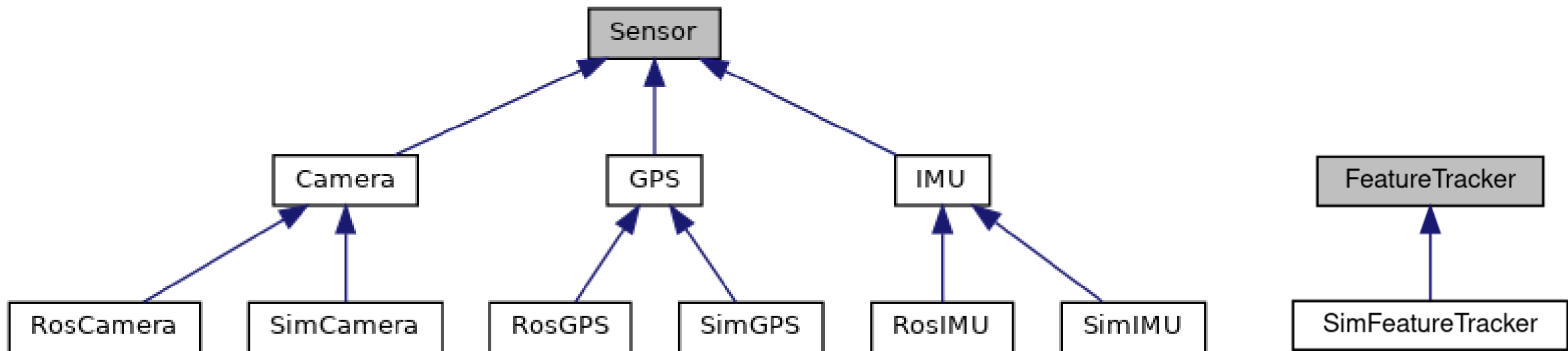


Cost of Integrated Coding

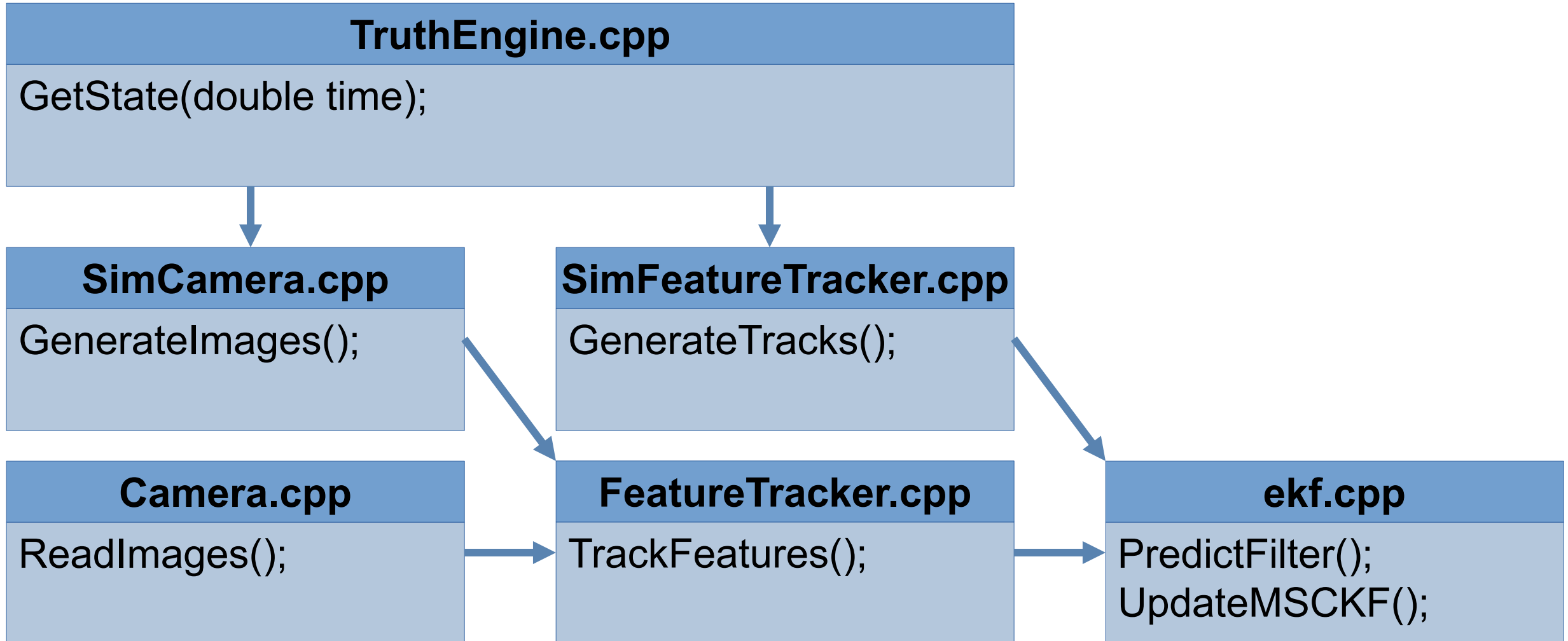


Abstraction - Example

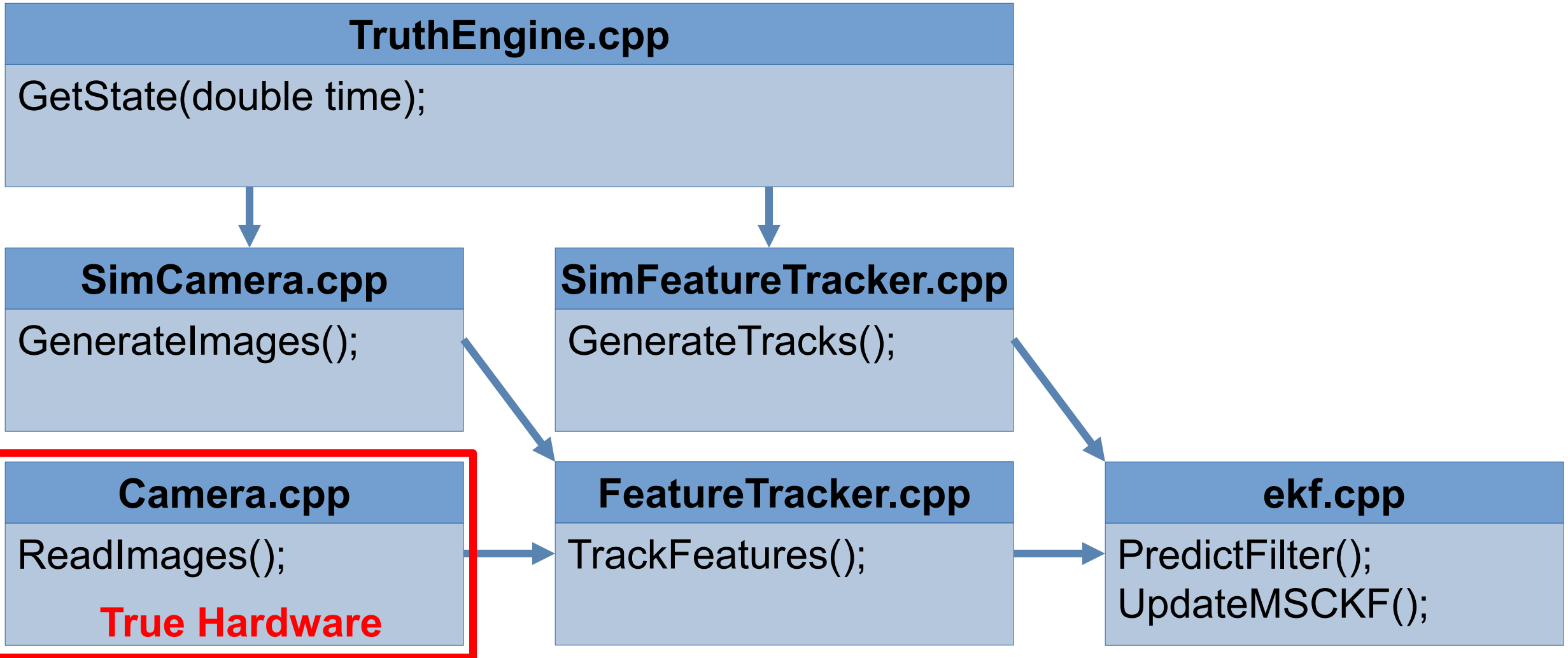
- Sensors call updates to filter / algorithm
 - Utilize real or simulated sensor messages
- Feature tracker utilizes camera measurements
 - True deployment utilizes true camera measurements
 - High-Fidelity simulation provides ray-traced images
 - Low-Fidelity simulation provides “pre-tracked features”



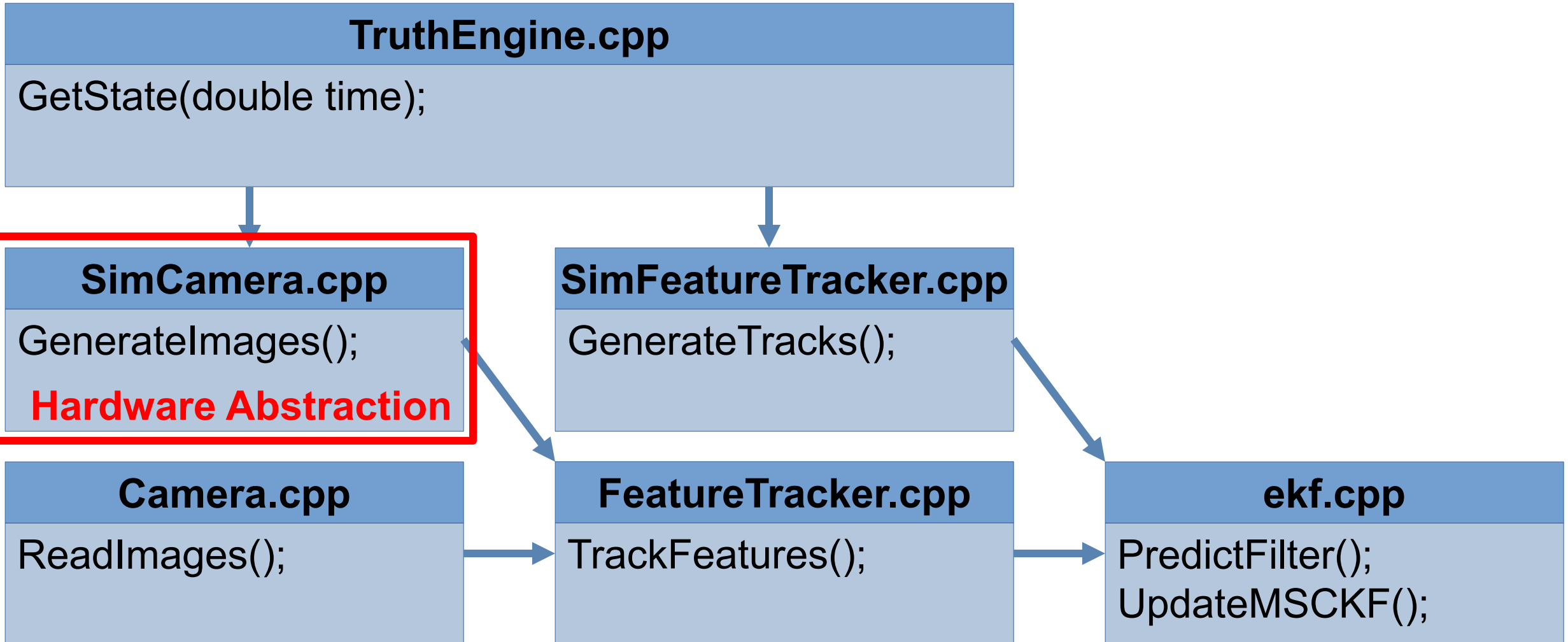
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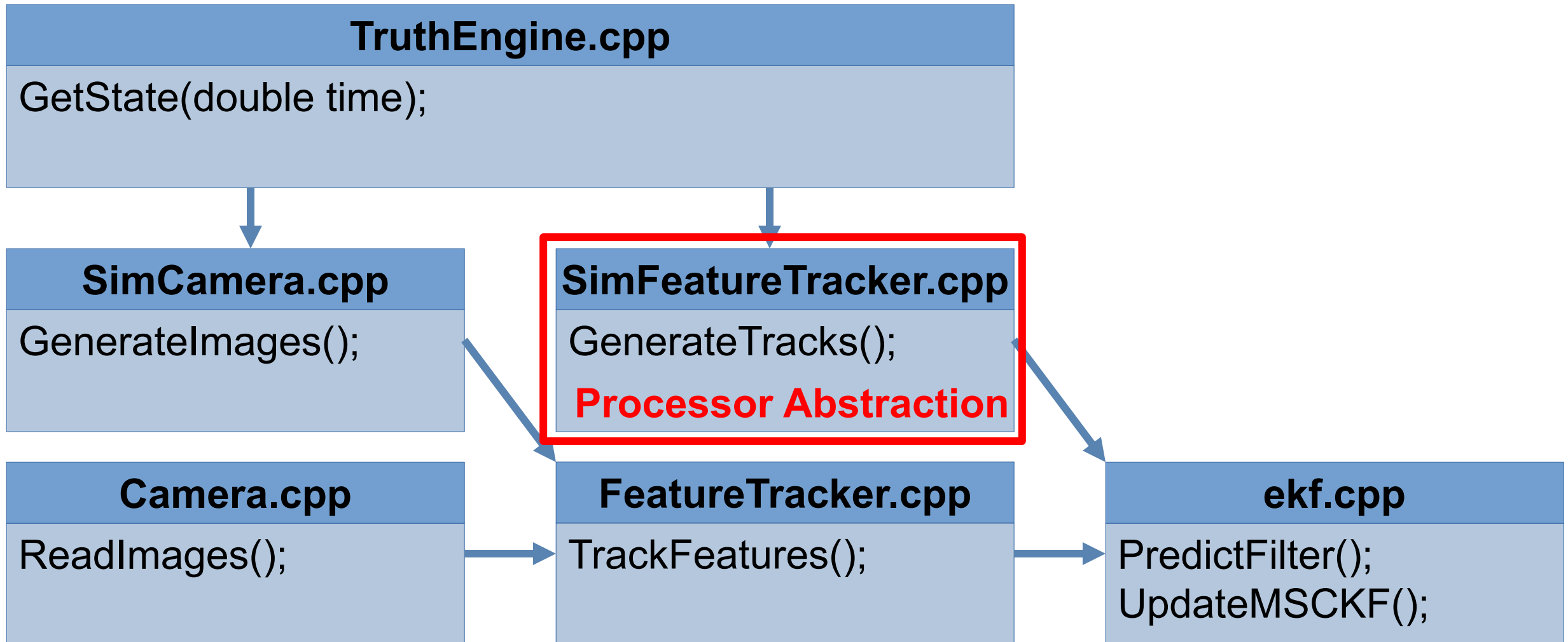
Abstraction - Example



Abstraction - Example

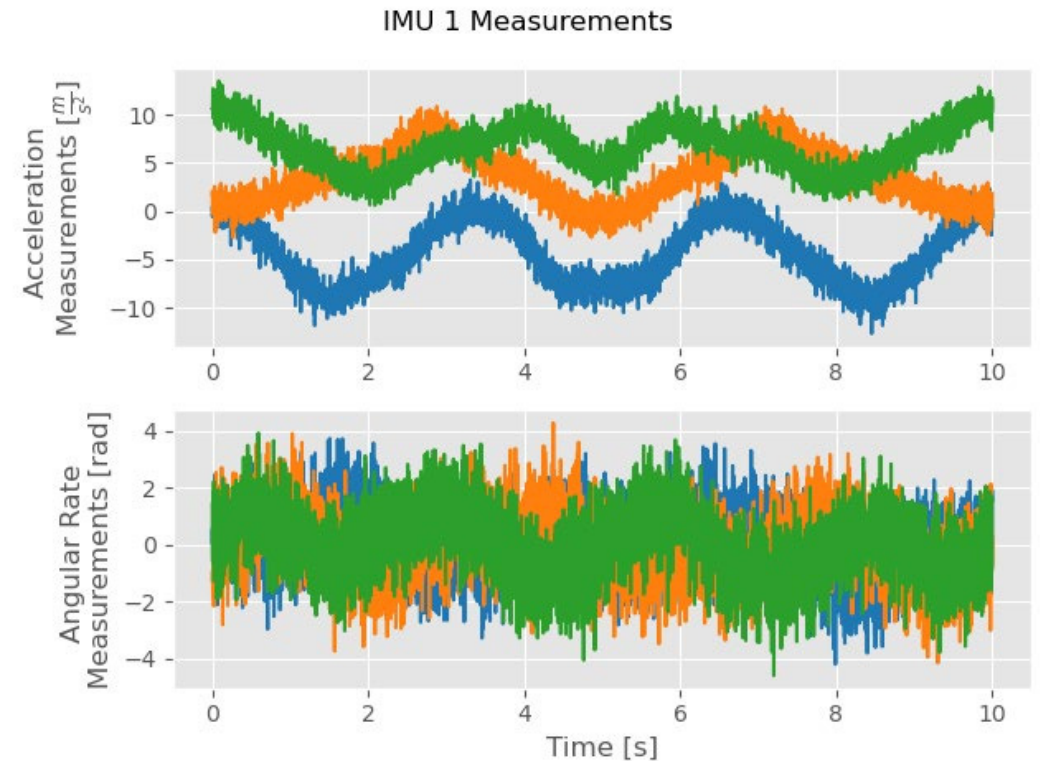


Abstraction - Example



Abstraction Models: IMU

$$\begin{bmatrix} \tilde{a}_x \\ \tilde{a}_y \\ \tilde{a}_z \end{bmatrix} = \begin{bmatrix} S_{ax} & 0 & 0 \\ 0 & S_{ay} & 0 \\ 0 & 0 & S_{az} \end{bmatrix} \begin{bmatrix} 1 & \alpha_{a1} & \alpha_{a2} \\ \alpha_{a3} & 1 & \alpha_{a4} \\ \alpha_{a5} & \alpha_{a6} & 1 \end{bmatrix} \left(\begin{bmatrix} a_x \\ a_y \\ a_z \end{bmatrix} + \begin{bmatrix} b_{ax} \\ b_{ay} \\ b_{az} \end{bmatrix} \right) + \begin{bmatrix} n_{ax} \\ n_{ay} \\ n_{az} \end{bmatrix}$$



Abstraction Models: Camera

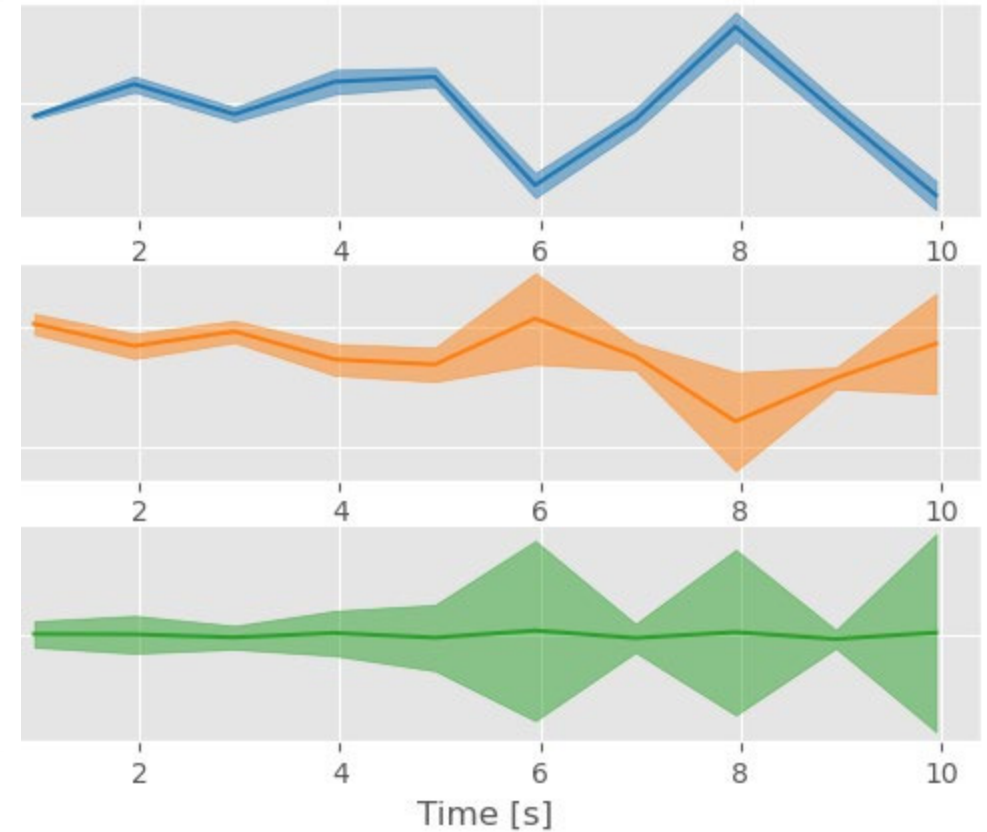
$$x_n = \frac{c \mathbf{p}_x f_x + c_x}{w^c \mathbf{p}_z} - 1$$

$$y_n = \frac{c \mathbf{p}_y f_y + c_y}{h^c \mathbf{p}_z} - 1$$

$$r^2 = x_n^2 + y_n^2$$

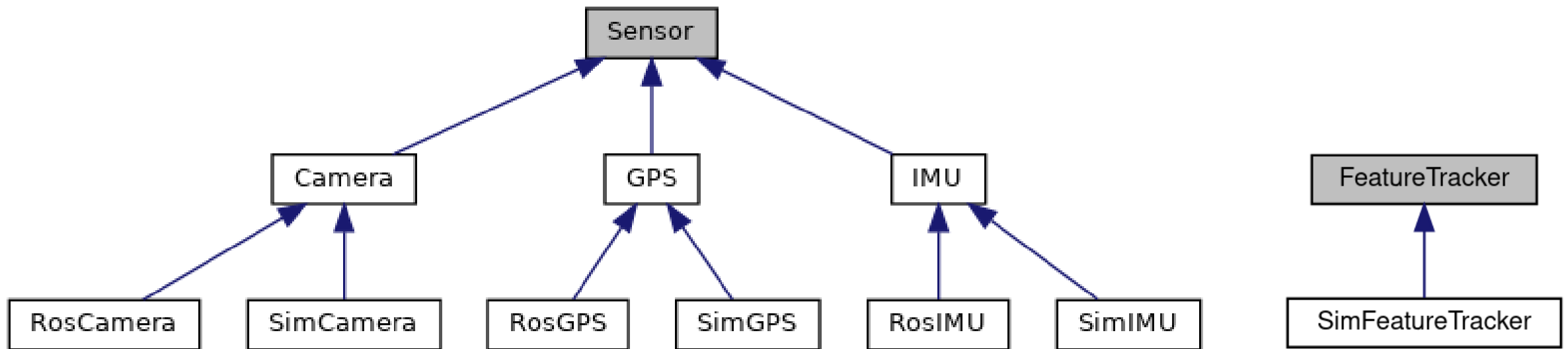
$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x_n \\ y_n \end{bmatrix} \cdot \underbrace{\begin{bmatrix} 1 + k_1 r^2 + k_2 r^4 + k_3 r^6 \\ 1 + k_1 r^2 + k_2 r^4 + k_3 r^6 \end{bmatrix}}_{\text{Radial}} + \underbrace{\begin{bmatrix} 2p_1 x_n y_n + p_2 \cdot (r^2 + 2x_n^2) \\ p_1 \cdot (r^2 + 2y_n^2) + 2p_2 x_n y_n \end{bmatrix}}_{\text{Tangential}}$$

Camera 3 Triangulation Errors



Abstraction - Benefits

- Improves accuracy of simulation
- Catches more bugs earlier
- Reduces rework (no code divergence)
- More beneficial unit testing
- Robust Monte Carlo testing



Unit Testing

*It's
Easy!*

- This architecture allows test-driven development
- Any tweaks or examples can become tests
- Tests ensure code accuracy and functionality
- Can be automated per commit / merge request

GoogleTest

+

CMake

+

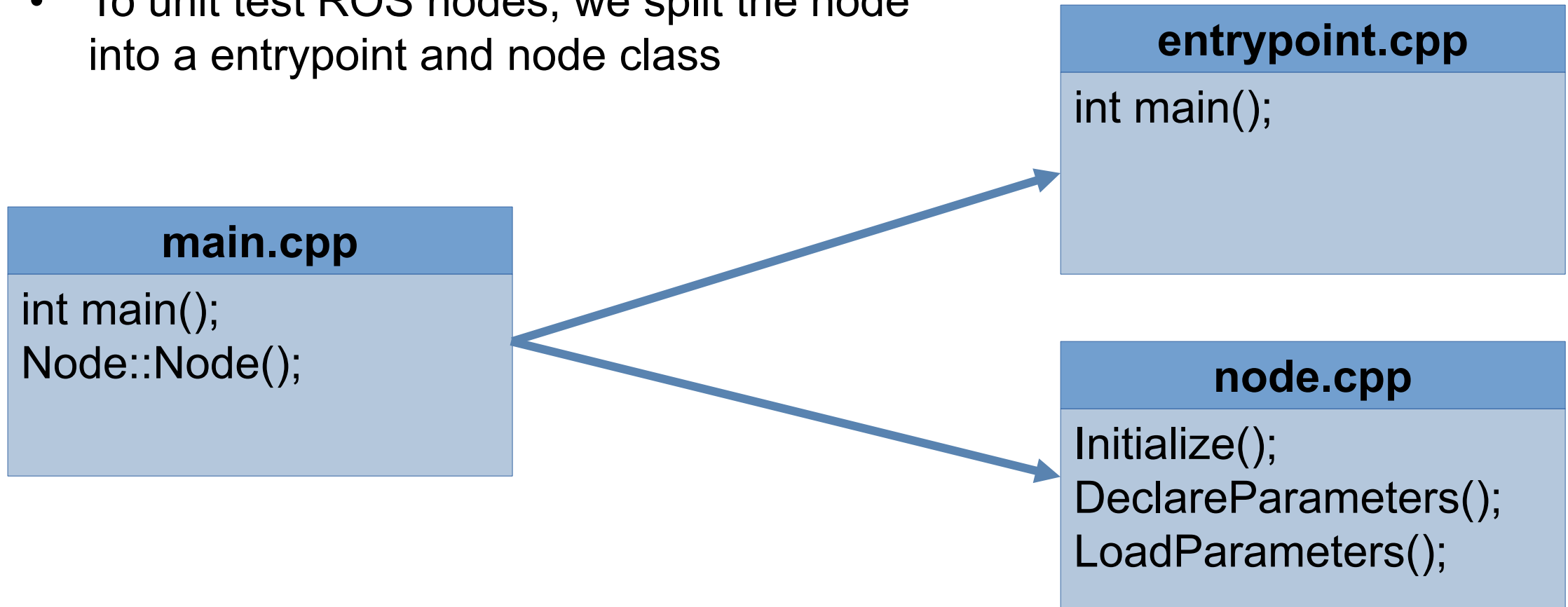
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Unit Testing Nodes

- Ideally, we should test as much as possible
- To unit test ROS nodes, we split the node into a entrypoint and node class



Unit Testing Nodes

entrypoint.cpp

```
int main();
```

node.cpp

```
Initialize();  
DeclareParameters();  
LoadParameters();
```

test.cpp

```
TEST_F(ExampleNode, ExampleNode_test)  
{  
    ExampleNode node;  
  
    node.Initialize();  
    node.DeclareParameters();  
  
    node.set_parameter(rclcpp::Parameter("param1"));  
    node.set_parameter(rclcpp::Parameter("param2"));  
  
    node.LoadParameters();  
}
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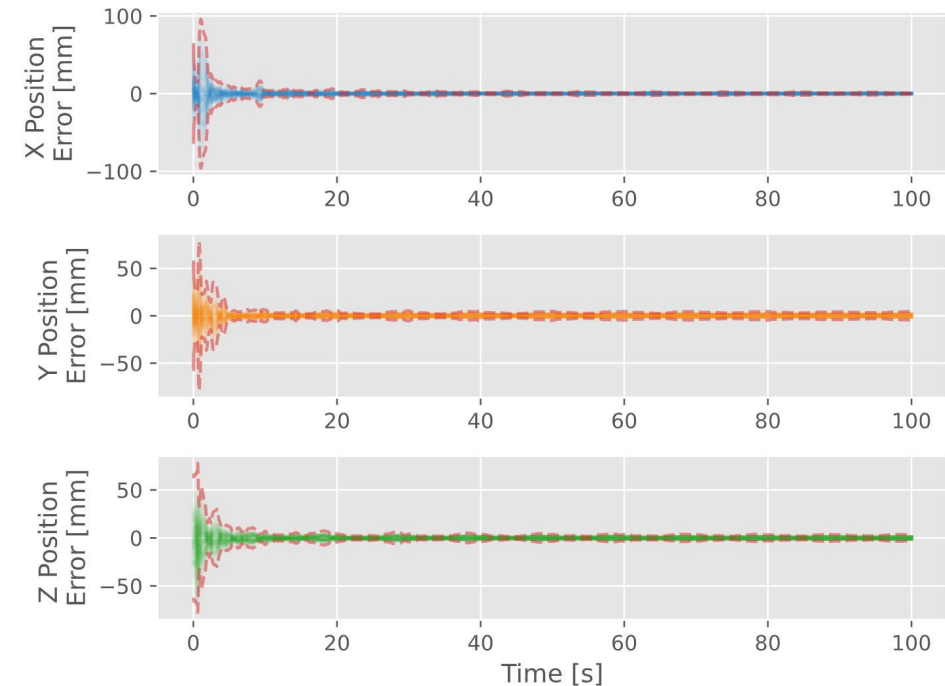
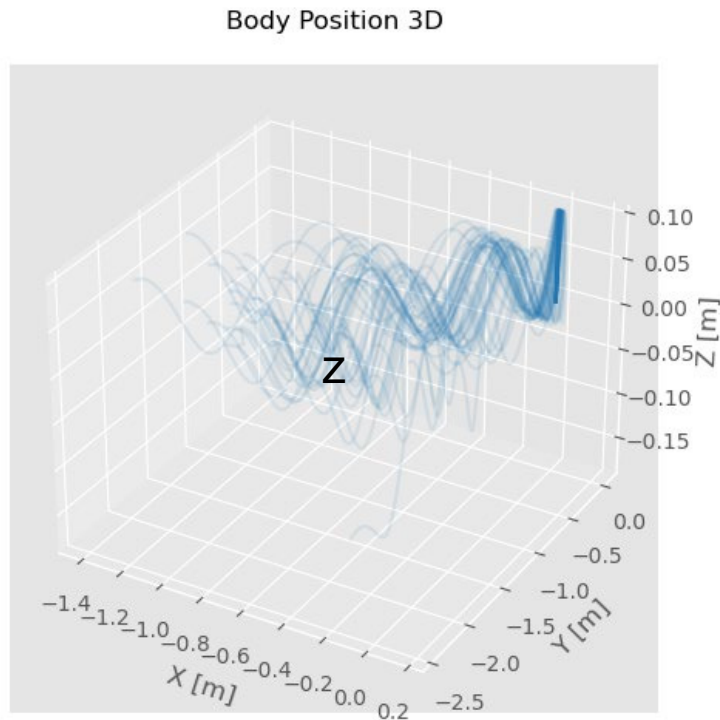
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```

Monte Carlo Testing

- With fast enough simulations, we can run thousands of example datasets
- Random initialization and measurement errors are inserted
- Utilizing abstractions increases confidence of filter stability



Key Takeaways

- Stop developing simulations separate from deployments
- Utilize existing models for integrated simulations
- Abstract layers as low as possible
- Utilize multiple layers of abstraction for various fidelity / execution speed
- Try out EKF-CAL! We love feedback and collaboration!

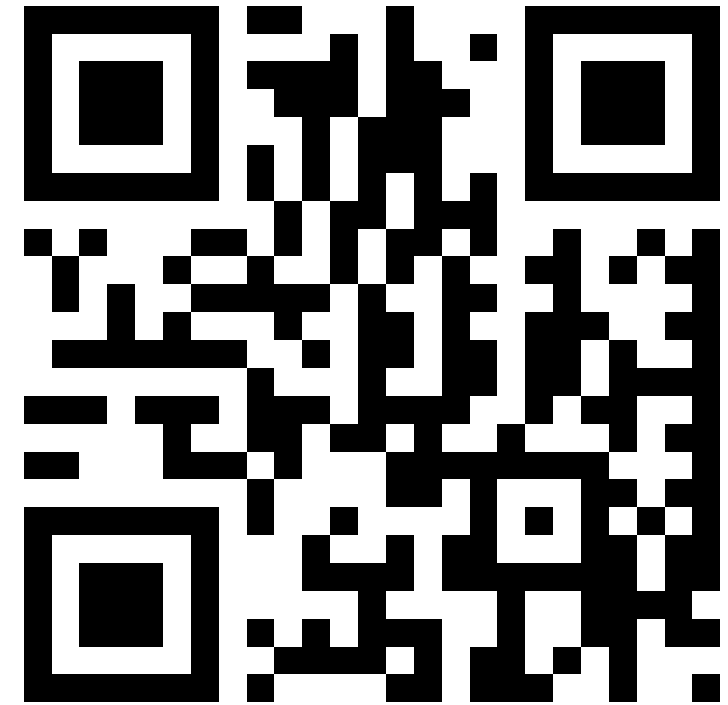
Presentation References

1. J. Hartzler and S. Saripalli, " Online Multi Camera-IMU Calibration", IEEE International Symposium on Safety, Security, and Rescue Robotics (SSRR), 2022. [IEEE](#), [arXiv](#)
2. P. Jiang and S. Saripalli, "LiDARNet: A Boundary-Aware Domain Adaptation Model for Point Cloud Semantic Segmentation," 2021 IEEE International Conference on Robotics and Automation (ICRA), Xi'an, China, 2021, pp. 2457-2464, [IEEE](#)
3. Experimental Evaluation of 3D-LIDAR Camera Extrinsic Calibration, S. Mishra, P. Osteen, G. Pandey and S. Saripalli, IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS, 2020), [arXiv](#)
4. Extrinsic Calibration of a 3D-LIDAR and a Camera, S. Mishra, G. Pandey and S. Saripalli, IEEE Intelligent Vehicles Symposium (IV, 2020) [arXiv](#)
5. Chustz, G., & Saripalli, S. (2021). ROOAD: RELLIS Off-road Odometry Analysis Dataset. [arXiv](#)

Questions?



[EKF-CAL Repository](#)



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