







Leverage Artificial Intelligence to Learn, Optimize, and Wargame (LAILOW) for Navy Ships

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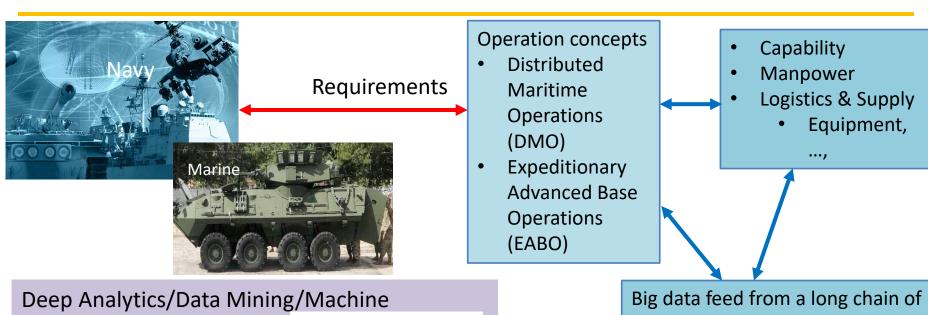
In the Special Webinar *Developing Artificial Intelligence in Defense Programs f*or the
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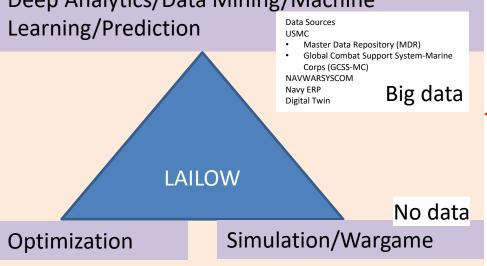


Challenges, Needs, and LAILOW Methods









Big data feed from a long chain of readiness components

- Deployment
- Demand, CASREPs
- Maintenance
- Supply

Analytics

- Requisition
- Transportation
- Finance processes
- Activities
- Decisions



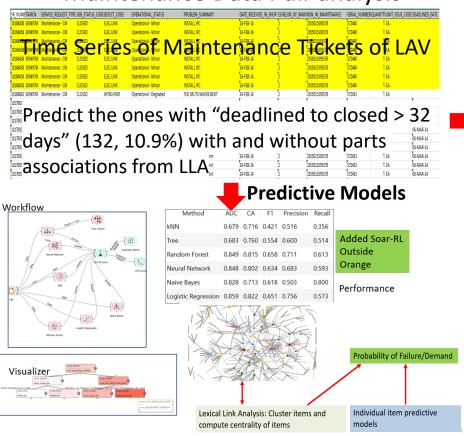
Use Case 1: Marine Maintenance and Supply System

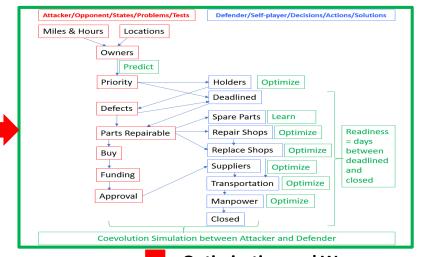


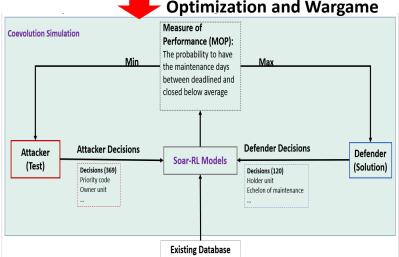


Land Armored Vehicle (LAV) (1/2013 to 1/2020), E0949 GCSS-MC

Maintenance Data Pull-analysis





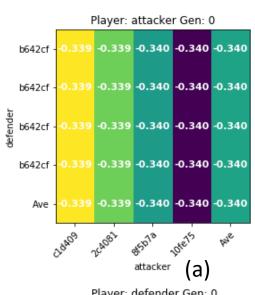


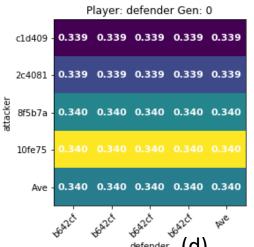


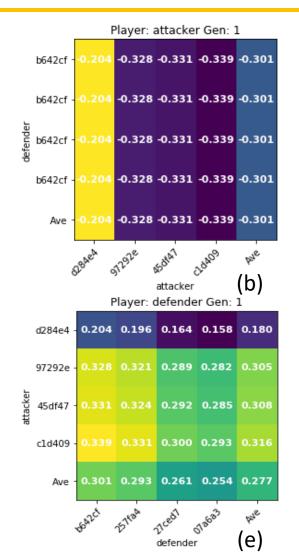
Use Case 1: Coevolution Process

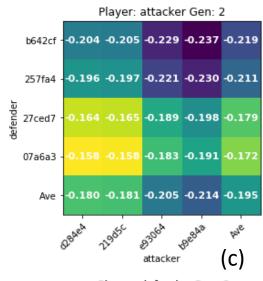


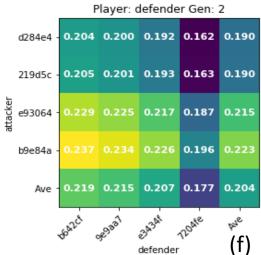














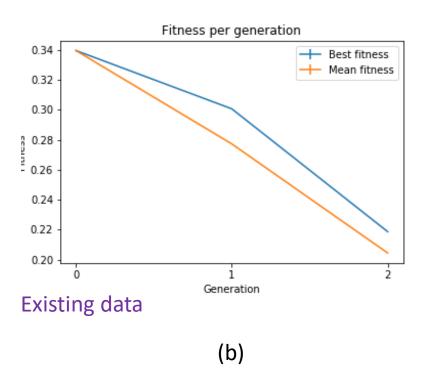
Use Case 1: Summary





Attackers/Problems

Defender/Solutions



The Wargame can systematically simulate and discover possible new tests or "vulnerabilities" for the complex system and evolve solutions accordingly.





Use Case 2: Predictive Risk Sparing Matrix (PRiSM)

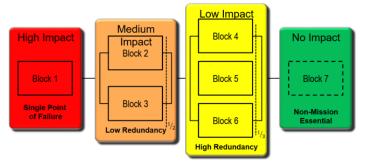






Goal/ Baseline: Determine which parts are likely to fail during operational deployments

"PREDICT" – The Data (AI/ML)



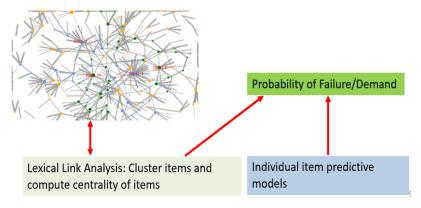
NAVSEA MRBD utilizes PEO C4I approved Reliability Block Diagrams

▼ NSWC Corona utilizes Machine Learning analytics with Python, NumPy, Scikit-Learn, Pandas, Tableau, data structures, and algorithm design for data science and advanced programming and techniques

Focus on Predictive Results

Source: Predictive Risk Sparing Matrix (PRiSM), Integrated Product Support/Logistics, Fleet Readiness Directorate (FRD), NAVWARSYSCOM

Applied LAILOW's LLA:



Results: For a test incorporated LLA recently, Theodore Roosevelt Carrier Strike Group (TR CSR) 51 of 64 (~80%) high impact C4I parts that were identified failed and had actually, matched with either LLA or PRiSM, improved from 36 matched from PRiSM alone. PRiSM and LLA are complementary for prediction.



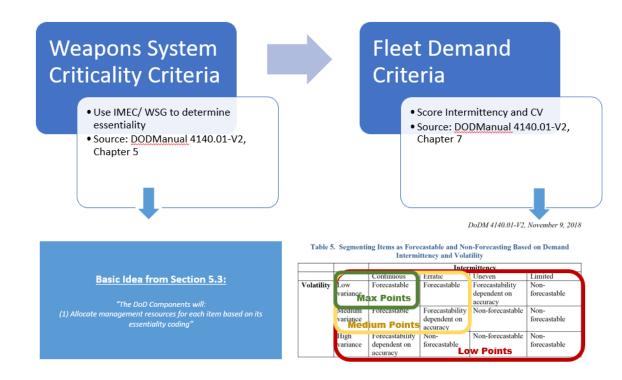
Use Case 3: Deep Analytics for Readiness Impacts of Underfunding Spares Backlogs





Baseline

- Objective: Prioritize the demanded parts (items) in the Financially Restricted Work Que (FRWQ)
- In the process to see
 if feasible and helpful
 to apply LAILOW's
 lexical link analysis to
 prioritize the items





Conclusions, Recommendations, Acknowledges, and Disclaimer





Conclusions

- The LAILOW framework provides a holistic predictive and simulation platform to improve readiness
- Soar-RL results: comparable to other predictive machine learning algorithms, rule-based and explainable, integrated with the coevolutionary algorithm
- The wargame with the Soar-RL and coevolutionary algorithm integration can systematically
 - Simulate and discover possible new tests or "vulnerabilities" of the value chain
 - Evolve solutions or "resiliency" accordingly

Recommendations for Navy Ships

- Adopt more deep analytics, machine learning and AI algorithms for big data or no data
- Focus on the entire spectrum or end-to-end (E2E) logistic planning for

Acknowledges

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- **Disclaimer**: The views presented are those of the authors and do not necessarily represent the views of the U.S. Government, Department of Defense (DoD), or their Components.