

# PFN Researcher Interview

Lekan Molu

Smilow Center for Translational Research,  
The University of Pennsylvania, Philadelphia, PA

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# Personal Background

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- PhD in Electrical and Computer Engineering, University of Texas at Dallas, Richardson, USA. 2014–2019
  - *“A Multi-DOF Soft Robot Mechanism for Patient Motion Correction and Beam Orientation Selection in Cancer Radiation Therapy.”*
- Master of Science in Engineering in Control Systems, The University of Sheffield, Sheffield, United Kingdom. 2012
  - *“Autonomous Navigation of a Rotorcraft Unmanned Aerial Vehicle using Machine Vision.”*

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- Deep BOO: Automating Beam Orientation Selection in Intensity Modulated Radiation Therapy  
**Olalekan Ogunmolu**, Michael Folkerts, Dan Nguyen, Nicholas Gans, and Steve Jiang.  
*Algorithmic Foundations of Robotics XIII, International Workshop (WAFR)*, Mérida, Mexico. December 2018.
- Minimax Iterative Dynamic Game: Application to Nonlinear Robot Control Tasks  
**Olalekan Ogunmolu**, Nicholas Gans, and Tyler Summers.  
*IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*, Madrid, Spain. October 2018. DOI: 10.1109/IROS.2018.8594037.

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- Mechanism and Constitutive Model of a Continuum Robot for Head and Neck Cancer Radiotherapy.  
**Olalekan Ogunmolu**, Xinmin Liu, Nicholas Gans, and Rodney Wiersma  
Submitted to *Robotics and Automation Letters (ICRA 2020)*, September 2019.
- A Fast Deep Learning Approach for Beam Orientation Selection Using Supervised Learning with Column Generation on IMRT Prostate Cancer Patients  
Azar Sadeghnejad Barkousaraie, **Olalekan Ogunmolu**, Steve Jiang, and Dan Nguyen.  
Submitted to *Medical Physics (An AAPM Journal)*, May 2019.
- Full publications list available at my online [publications page](#).

# Select Research Experience

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- **Visiting Postdoctoral Scholar**, Department of Radiation and Cellular Oncology, Pritzker School of Medicine, [The University of Chicago](#), Chicago, IL, USA. Summer 2019.
- **Postdoctoral Scholar**, Perelman School of Medicine, [The University of Pennsylvania](#), Philadelphia, PA, USA. 2019 - Present.
- **Hardware Integration Intern**, Amazon Robotics LLC, North Reading, MA, USA. 2016.

# Technical Presentation

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# Robustness in Deep Policies: An Overview

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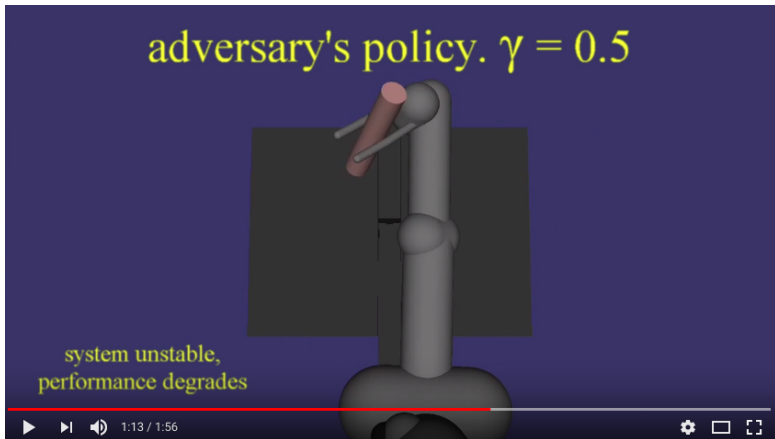
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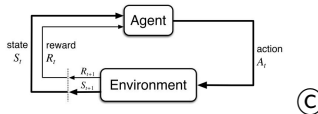
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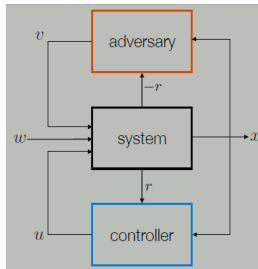


# The robustness conundrum

- How to know *a priori* a policy's robustness limits?



- How to inculcate robustness into multistage decision policies?





# Problem Setup

- To quantify the brittleness, we optimize the stage cost

$$\max_{\mathbf{v}_t \sim \psi \in \Psi} \left[ \sum_{t=0}^T \underbrace{c(\mathbf{x}_t, \mathbf{u}_t)}_{\text{nominal}} - \gamma \underbrace{g(\mathbf{v}_t)}_{\text{adversarial}} \right]$$

- To mitigate lack of robustness, we optimize the *cost-to-go*

$$\mathcal{J}_t(\mathbf{x}_t, \pi, \psi) = \min_{\mathbf{u}_t \sim \pi} \max_{\mathbf{v}_t \sim \psi} \left( \sum_{t=0}^{T-1} \ell_t(\mathbf{x}_t, \mathbf{u}_t, \mathbf{v}_t) + L_T(\mathbf{x}_T) \right),$$

- and seek a saddle point equilibrium policy that satisfies

$$\mathcal{J}_t(\mathbf{x}_t, \pi^*, \psi) \leq \mathcal{J}_t(\mathbf{x}_t, \pi^*, \psi^*) \leq \mathcal{J}_t(\mathbf{x}_t, \pi, \psi^*),$$

# Results: Brittleness Quantification

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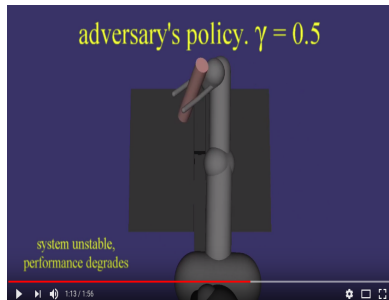
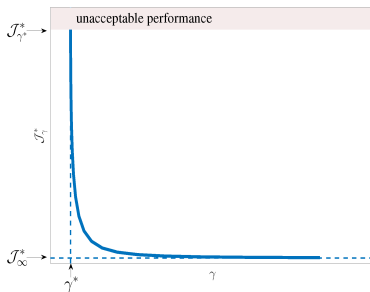
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# ILQG Algorithm Example

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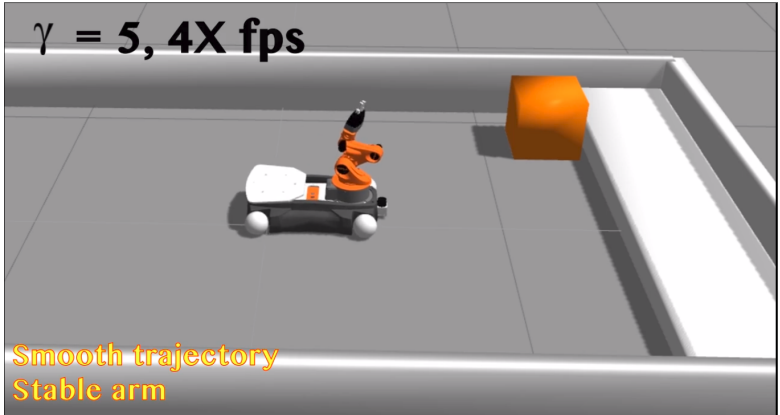
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# Results: Iterative Dynamic Game

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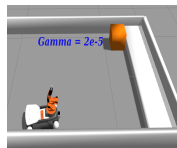
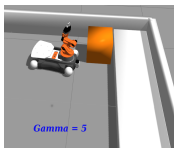
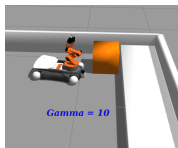
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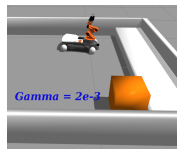
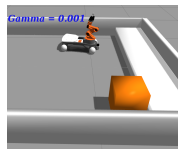
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$x_1^*$



$x_2^*$



End pose of the KUKA platform with our iDG formulation given different goal states and  $\gamma$ -values