

Opportunity Title: Low Shot Training and Testing of Machine Learning Algorithms for Detection of Items of Concern **Opportunity Reference Code:** ICPD-2022-42

Organization Office of the Director of National Intelligence (ODNI)

Reference Code ICPD-2022-42



Complete your application – Enter the rest of the information required for the IC Postdoc Program Research Opportunity. The application itself contains detailed instructions for each one of these components: availability, citizenship, transcripts, dissertation abstract, publication and presentation plan, and information about your Research Advisor co-applicant.

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Application Deadline 2/28/2022 6:00:00 PM Eastern Time Zone

Description Research Topic Description, including Problem Statement:

- X-Ray scanners are deployed for security screening of bags at various security checkpoints: in airports, other transport hubs, and at critical national infrastructure.
- There is a great interest to assess the effectiveness of machine learning (ML) algorithms deployed on X-Ray scanners to detect concealed items of concerns.
- To date, it has been considered that a large number and variety of images are required to train a ML algorithm for effective threat detection. However, it may not be possible to provide a large image set of items of concern to developers. Therefore, it is of importance to be able to develop effective ML algorithms using a small number oft raining images-termed low shot training
- It is also of interest to explore the feasibility of independently testing an algorithm with a small test set of images, termed low shot testing.
- A further option to overcome difficulties with provision of large image sets of images of concern, is to assess the effectiveness of using synthetic imagery.

Example Approaches:

A small training set of images of items of concern will be provided (open format, not security classified) to the postdoc to be used to develop a low shot trained detection algorithm. A small test set of images will be provided to be used to test the algorithm. A large data set of related images will be provided to train and test an algorithm with a conventional supervised learning approach. A comparison is to be made of the low shot learning approach and the supervised learning approach using the large data set. Similarly, a comparison is to be made of the low shot testing approach with testing using the large test set.

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Based on the provided images for items of concern, the postdoc will create synthetic images and develop a study to compare the effectiveness of synthetic images, real images, and a combination of both.

Combine the approaches of low shot learning and synthetic imagery to assess the most effective way to train aML algorithm for items of concern in the absence of large datasets.

Relevance to the Intelligence Community:

It is often very difficult/impossible to provide large image sets of items of concern to developers. This may be because these image sets do not exist and would require significant resources of experts and funding to produce; or the images may be security classified. It would therefore be advantageous to the intelligence community toknowifitispossibletoeffectivelytrainMLalgorithmswith:farfewerimagesand/or

synthetic images.

Testing the effectiveness of security screening equipment with deployed algorithms is a lengthy, costly process that requires facilities and staff capable of handling items that are hazardous and/or security classified. A new approach to testing is proposed that, if successful, could make equipment and algorithm testing significantly faster and cheaper. This rapid testing will be essential for the intelligence community to keep pace with and exploit the many novel machine learning algorithms that are anticipated to be developed as this technology matures.

Key Words: Machine Learning, X-Ray Screening, Artificial Intelligence, Low Shot Learning, Low Shot Testing, Synthetic Imagery

Qualifications Postdoc Eligibility

- U.S. citizens only
- Ph.D. in a relevant field must be completed before beginning the appointment and within five years of the application deadline
- Proposal must be associated with an accredited U.S. university, college, or U.S. government laboratory
- Eligible candidates may only receive one award from the IC Postdoctoral Research Fellowship Program

Research Advisor Eligibility

- Must be an employee of an accredited U.S. university, college or U.S. government laboratory
- Are not required to be U.S. citizens

Eligibility Requirements

Citizenship: U.S. Citizen Only Degree: Doctoral Degree.

Discipline(s):

- Chemistry and Materials Sciences (12. (12)
- Communications and Graphics Design (2. •)
- Computer, Information, and Data Sciences (16)
- Earth and Geosciences (21 (19)



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- Engineering (<u>27</u> ^(©))
- Environmental and Marine Sciences (14 (*)
- Life Health and Medical Sciences (45 (19)
- Mathematics and Statistics (<u>10</u>)
- Other Non-Science & Engineering (2.)
- Physics (<u>16</u> [●])
- Science & Engineering-related (1.)
- Social and Behavioral Sciences (27.
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