All comments will be made public as-is, with no edits or redactions. Please be careful to not include confidential business or personal information, otherwise sensitive or protected information, or any information you do not wish to be posted.

Comment Template for Responses to NIST Artifical Intelligence Risk Management

| General RFI Topics (Use as many lines as you like) | Response # | Responding organization | Responder's name | Paper Section (if applicable) | Response/Comment (Include ration |
|---|------------|--|---------------------|-------------------------------------|--|
| | | | | | |
| Responses to Specific Request for information (pages 11,12, 13 and 14 of the RFI) | | | | | |
| 1. The greatest challenges in improving how AI actors manage AI-related risks – where "manage" means identify, assess, prioritize, respond to, or communicate those risks; | | 1 Infocomm Media Development Authority | | | The lack of common metrics that can safety metrics and risk profile of an A understand assessment tools which A Current state-of-the-art tools have de challenge for different AI actors to co There are lack of tools in many classe unsupervised learning and deep learn It is also a challenge to identify a com be preoccupied with performance ov Lastly, there is a lack of mitigation me use to address fairness problem when |
| | | | | | |

Submit comments by August 19, 2021:

| ale) | Suggested change |
|---|------------------|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| be used in different AI trust scenarios to measure the different principles in trustworthiness is a key challenge. For example, the | - |
| I system used in self-driving vehicles will be different compared to that of a facial recognition system. There is a lack of easy-to- | |
| I actors can use to evaluate the trustworthiness of the AI systems. | |
| | |
| fined their own metrics and is often vague and open to different interpretations. Across the variety of AI use cases, it is a | |
| me to a consistent interpretation of the raw output of tools. | |
| | |
| s of AI such as graph neural networks or multiclass machine learning. For instance, it will be a challenge to assess fairness in | |
| ing as most fairness assessment tools support some tasks (e.g. classification) in supervised learning. | |
| | |
| mon framework for all stakeholders to agree on the AI-related risks that have to be managed. For example, data scientists may | |
| er fairer outcomes, while compliance teams may be preoccupied with risk and liability. | |
| | |
| asures that can help the AI actors to respond to AI-related risks. For instance, there is no guideline that the data scientists can | |
| n they identify bias in the datasets that might cause harmful outcome. | |
| | |
| | |
| | |
| | |
| | |
| | |



| 2. How organizations currently define and manage characteristics of AI trustworthiness and whether there are important characteristics which should be considered in the Framework besides: accuracy, explainability and interpretability, reliability, privacy, robustness, safety, security (resilience), and mitigation of harmful bias, or harmful outcomes from misuse of the AI; | 1 | Infocomm Media Development Authority | - | - | We define trustworthiness based on tw should be human-centric with user wel In addition, we provide guidance in fou The four key areas are: internal govern operations management, and stakehold framework set up effective structures, |
|--|---|--|---|---|--|
| 3. How organizations currently define and manage principles of AI trustworthiness and whether there are important principles which should be considered in the Framework besides: transparency, fairness, and accountability; | 1 | Infocomm Media Development Authority | | | We think that the following 10 attribute - Transparent (Information about AI systems - Fairness (Ensure that there is no unint - Accountability (Making sure that AI accountability (Making sure that AI accountability (Ability to understand account) - Explainability (Ability to understand account) - Safety (AI systems must be reliable and - Safety (AI systems must be secured - Data Governance (Training data must - Repeatability (For purpose of verifical - Reproducibility (For purpose of verifical - Reproducibility (For purpose of verifical - Robustness (Ability to handle unexpect AI systems should be explainable and t outputs over multiple runs demonstrated reproduce the same results by a third-puthereby reinforcing trust by users. AI systems should also be robust to have falling trees covering part of traffic signed the AI systems such that it will not caused system. |
| 4. The extent to which AI risks are incorporated into different organizations' overarching enterprise risk management – including, but not limited to, the management of risks related to cybersecurity, privacy, and safety; | 1 | Infocomm Media Development Authority | - | | We believe that AI risks should be cont In most cases, AI-related risks (e.g. data practices and standards applied in othe communications on managing AI-relate |
| | | | | | |

two guiding principles. Firstly, the AI used in the decision-making must be explainable, transparent and fair. Secondly, the AI used well-being and safety being primary considerations in designing and deploying the AI solution.

our key areas that help promote responsible use of AI, in a manner that is algorithm, technology, and business agnostic.

ernance structures and measures, risk assessment to determine the level of human involvement in AI-augmented decision-making, nolder interaction and communication to concretely operationalise AI ethics principles. These help organizations adopting the es, determining acceptable risks, establish good data governance practices, and mitigate bias in data and AI models.

| outes are important principles that must be considered in the Framework. | - |
|--|---|
| system is available) | |
| nintentional discrimination in the results produce by the AI system) | |
| l actor is held responsible for the decision) | |
| d and interpret what AI system is doing) | |
| and wil not cause harm) | |
| ed from cyber attacks) | |
| ust be managed properly and the quality must be ensured) | |
| ication & validation of the results) | |
| rification & validation of the rseults by independent party) | |
| pected input/adversarial attacks) | |
| | |
| d transparent so the users of the system can trust the predicted results. Al systems that are able to repeat and achieve consistent | |
| rate the reliability and stability of these systems. Repeatable results are also a key complement to explainable AI. Being able to | |
| d-party team helps to verify claims about the AI systems, allowing the systems to be accredited or certified to perform as claimed, | |
| | |
| | |
| handle unexpected input to produce correct output. Such unexpected input may or may not be due to adversarial attacks (e.g., | |
| igns but the autonomous vehicle would still be able to recognise the traffic signs correctly). This would help to ensure the safety of | |
| ause harm. Al systems must also be secured from cyberattacks to ensure that they remain safe and accurate for the users of the | |
| | |
| | |
| ontextualized with respect to an overarching risk management framework. | - |
| | |
| lata privacy, security of the system) should be aligned to the organization's internal policies and compliance. This ensures the | |
| ther areas are also consistently applied on the AI systems. It will improve trust in these systems and simplify stakeholder | |
| ated risks as an extension of existing policies. | |
| | |
| | |
| | |
| | |



| 5. Standards, frameworks, models, methodologies, tools, guidelines and best practices, and principles to identify, assess, prioritize, mitigate, or communicate AI risk and whether any currently meet the minimum attributes described above; | 1 | Infocomm Media Development Authority | - | - Singapore's Model AI Governance Fr - ISO/IEC TR 24028:2020: Overview of |
|---|---|--|---|---|
| 6. How current regulatory or regulatory reporting requirements (e.g., local, state, national, international) relate to the use of AI standards, frameworks, models, methodologies, tools, guidelines and best practices, and principles; | 1 | Infocomm Media Development Authority | | Currently, there is no regulatory requi provided by Singapore Government v The Singapore Government provides considerations and measures to be im |
| 7. AI risk management standards, frameworks, models, methodologies, tools, guidelines and best practices, principles, and practices which NIST should consider to ensure that the AI RMF aligns with and supports other efforts; | 1 | Infocomm Media Development Authority | | Singapore's Model AI Governance Fra |
| 8. How organizations take into account benefits and issues related to inclusiveness in AI design, development, use and evaluation – and how AI design and development may be carried out in a way that reduces or manages the risk of potential negative impact on individuals, groups, and society. | 1 | Infocomm Media Development Authority | | Organizations should weigh in all the they are found to use or develop unfa AI systems that are properly designed The developers of the AI systems sho AI developers should use available to in the model development phase. Apa In general, organizations should take are suited for purpose and do not pro understand how datasets could be inf develop mechanisms around accurate datasets for training, testing and valid |
| 9. The appropriateness of the attributes NIST has developed for the AI Risk Management Framework. (See above, "AI RMF Development and Attributes"); | 1 | Infocomm Media Development Authority | - | Highly appropriate |

Framework of trustworthiness in Al

| uirements to mandate the use of specific standards or best practices. Organisations are encouraged to adopt the guidelines voluntarily. | - |
|---|---|
| s guidelines such Model AI Governance Framework and Self-Assessment Guide for Organisations (ISAGO) for organizations on key mplemented in designing and deploying the AI solution. | |

amework

| possible damages that could happen to the organization when such issues arise. Organizations might develop poor reputation if ir AI systems. This could erode the customer base and hinder future businesses that are done by the organizations. Whereas for and tested to be inclusive will increase the value of the organizations. | - |
|---|---|
| uld be made aware of the negative impact on individuals, groups, and society when it comes to inclusiveness in AI development. It of properly test the datasets and machine learning models. This can help to highlight any potential bias issues that could arise Int from that, there should be a proper process in ensuring accountability in the decisions that are made by the AI developers. | |
| reasonable efforts to address multivariate sources of bias in AI. Specifically, it is critical that datasets used for AI model training pagate inherent flaws in data to effectively manage the risks of inaccuracy or bias. Organizations should institute processes to herently biased and develop a strategy to detect, understand and mitigate these issues. For instnace, organizations may have to data tagging, periodic review, and validating sources of datasets. Where the dataset permits, organizations may apply different ation to minimize negative impact. | |
| | |
| | - |



| 10. Effective ways to structure the | 1 | Infocomm Media | - | - | Singapore's Model AI Governance Fran |
|-------------------------------------|---|----------------|---|---|--|
| goals including but not limited | | Authority | | | assossment. Al developers can build in |
| to integrating Al rick | | Authonity | | | assessment, Al developers can build in |
| to, integrating Allisk | | | | | Al system graceruny. |
| management processes with | | | | | The France work can also be offectively |
| organizational processes for | | | | | interramework can also be effectively |
| developing products and services | | | | | integrating AFrisk processes with organ |
| trustwarthings and management | | | | | I claims for developing Al products and s |
| of All ricks. Despendents are asked | | | | | |
| to identify any surrent models | | | | | |
| which would be offective. These | | | | | |
| could include but are not limited | | | | | |
| to the NIST Cybersocurity | | | | | |
| Framework or Privacy | | | | | |
| Framework which focus on | | | | | |
| outcomes functions categories | | | | | |
| and subcatogories and also offer | | | | | |
| and subcategories and also orien | | | | | |
| reflecting current and desired | | | | | |
| approaches as well as tiers to | | | | | |
| describe degree of framework | | | | | |
| | | | | | |
| 11. How the Framework could be | 1 | Infocomm Media | - | - | Today, it is difficult to recruit the suitab |
| developed to advance the | | Development | | | (e.g. business groups vs technical group |
| recruitment, hiring, development, | | Authority | | | occur at various stages in the lifecycle of |
| and retention of a knowledgeable | | | | | set a clear indication to the DevSecOps |
| and skilled workforce necessary | | | | | |
| to perform AI-related functions | | | | | It will also be useful to reference other |
| within organizations. | | | | | These frameworks collectively improve |
| | | | | | frameworks of knowledge used in the |
| | | | | | |
| 12. The extent to which the | 1 | Infocomm Media | - | - | The Framework should include detailed |
| Framework should include | _ | Development | | | However, guidelines need to retain a de |
| governance issues, including but | | Authority | | | |
| not limited to make up of design | | | | | With reference to Singapore's Model A |
| and development teams. | | | | | assign clear roles and responsibilities for |
| monitoring and evaluation, and | | | | | |
| grievance and redress. | | | | | |
| | | | | | |
| | | | | | |

| mework highlights the importance of conducting risk assessment prior to the design of an AI system. This framework suggests everity of impact and possibility of impact on the individuals, taking into consideration context of use. Based on the risk n risk mitigating measures including the level of human involvement in AI-augmented decision making and the need to shut down | - |
|---|---|
| v structured in such a way that it provides AI actors with guidelines to test the principles against the AI systems apart from nizational processes. The risks could be managed but it will be much more effective to include testings to validate and verify the services. | |
| | |
| ble candidate to run AI-related functions within the organizations as most of the AI-related risks are siloed among the AI actors ups). The Framework could lay out the different aspects of AI-related risks (both non-technical risks and technical risks) that can of the AI systems. Each risk can be elaborated with guidelines that contain detail on how to assess, test and respond to. This will s, AI and audit community the skillsets that are required to perform AI-related functions within the organization. | - |
| r standards, methodologies or frameworks that are relevant or complementary to AI design, development, and deployment. e upon our appreciation of AI risks, and will chart essential skills for workforce learning and development, similar to other technology sector. | |
| | |
| d guidelines on how organizations can adapt existing or set up internal governance structures and risk mitigation measures. Jegree of flexibiltiy for it to be adapted to different business sectors and risk profiles. | _ |
| AI Governance Framework, this guides organizations to recognise and document AI-related risks in their internal policies and for the design and deployment of AI solutions. | |
| | |

