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Protecting
Our World,
Conquering
New Frontiers

Artificial Intelligence Techniques for Extracting Impervious Surface Areas from Satellite Imagery

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Outline of Presentation

- Background and Motivation**
- Research Problem**
- Basis for solution**
- General Methodological Workflow (derived)**
- Results**
- What our paper will discuss...**
- What our paper will conclude...**
- Justification and Significance of Research**

Urbanization and Impervious Surfaces

- Urbanization is a phenomenon that is globally growing at an accelerating rate
- Result in construction and development of Impervious Surfaces (IS)
- IS are artificial hard areas that does not allow water to seep into the ground and are recognized as **“built-up”, “developed”** or **“urban areas”**

Buildings



Parking Lots



Roads



Sidewalks and Pavements



- IS are key quantifiable indicators of urbanization.
- IS information is important to assist in quantifying urbanization and have proper urban planning and environmental management.

Direct Results of Urbanization

- Construction and development usually includes removal of the Earth's natural land covers which disturbs local ecological systems



Small Island Developing States (SIDS)

- Small size and Remoteness
- Exposure to Global Environmental Changes
- “Hotspots” of Climate Change
- Emits the least carbon into the atmosphere
- Environmental and Socio-economic vulnerabilities for development challenges
- Currently less resilient and less prosperous than larger developed countries
- Limited Resources
- Struggling to meet Sustainable Development Goals
- Requires assistance from businesses, academia, and society.



Caribbean SIDS

- Natural Disasters (Tropical Location)
- Low Elevations
- Difference in size, geology terrain, landscape, vegetation types
- “the highest debt percentage (76% of their GDP in 2014)”



- Poor Building and Infrastructural Development



- Unplanned Development in Urban Settlements

- Geospatial Data and Resource Challenges



Research Problem: How can we get **Impervious Surface Area Information**, needed to address the **urbanization issues**?

- We need imagery **data**....
- We need to **extract** the impervious surface features from the **data** – providing us information.....

Basis for Solution:

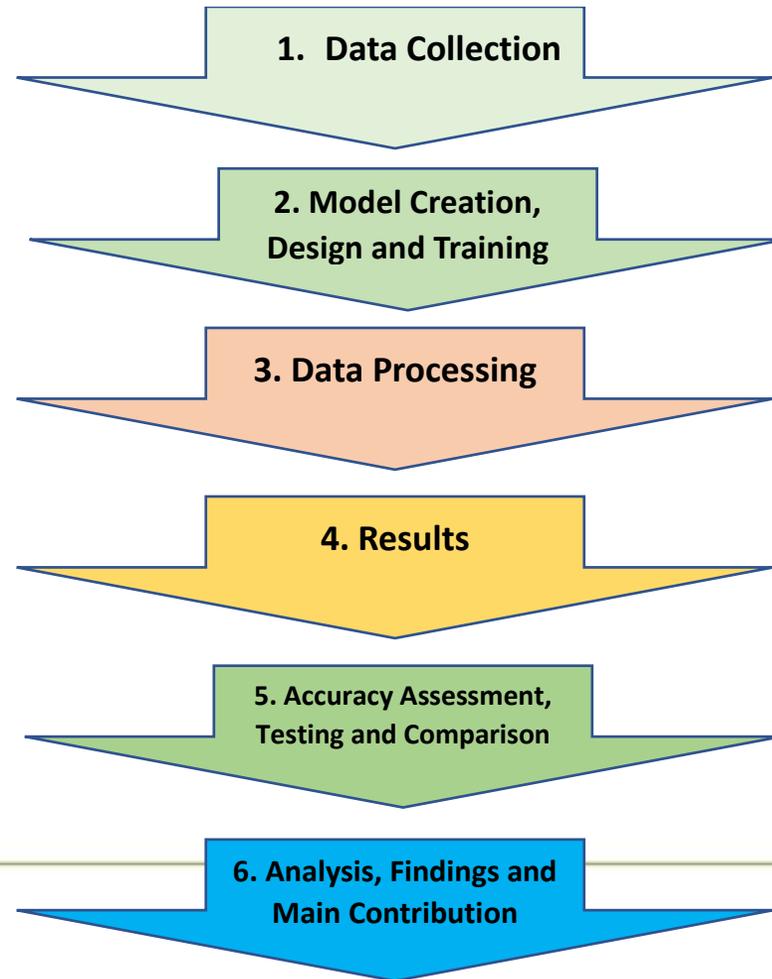
- Open-source satellite imagery: **Landsat 8** and **Sentinel-2**
- Artificial Intelligence for feature Extraction – **Machine Learning** and **Deep Learning**



General Methodological Workflow



Perform case studies in Trinidad (proof of concept) – Areas of different urban characteristics



Tested afterwards in some Caribbean SIDS – in islands of varying geologies and geographies, e.g., Volcanic Islands, Coral Islands, and Limestone Islands.

Results: Visuals

Landsat 8 *MLC*



RF



SVM



U-NET



Sentinel-2



ESRI's High resolution World Imagery

	Urban/Built-up
	Non-Urban/Un-Developed

Classification Results: Statistics

Actual Labels	Predicted Labels	
	Impervious	Permeable
Impervious	TP	FN
Permeable	FP	TN

Confusion Matrix

Landsat 8					Sentinel 2				
Method	TP	TN	FP	FN	Method	TP	TN	FP	FN
MLC	7481	715	190	1155	MLC	76802	5311	1470	2799
RF	8075	584	325	557	RF	77625	5071	1600	2086
SVM	8118	576	333	514	SVM	78420	5305	1470	1187
UNET	8047	588	318	588	UNET	64271	7250	395	14466

Accuracy Measures and Assessment

Accuracy Assessment:

Precision = $TP / (TP + FP)$

Recall = $TP / (TP + FN)$

Accuracy = $(TP + TN) / (TP + FN + TN + FP)$

F-score = $2 \times \text{recall} \times \text{precision} / (\text{recall} + \text{precision})$

Mean intersection over union (MIOU) = $TP / (FP + FN + TP)$

AI Method	Satellite	Technique	Accuracy	MIOU	F-Score	Recall	Precision
<i>Machine Learning</i>	Landsat 8	<i>MLC</i>	0.8590	0.8476	0.9175	0.8663	0.9752
		<i>RF</i>	0.9076	0.9015	0.9482	0.9355	0.9613
		<i>SVM</i>	0.9112	0.9055	0.9504	0.9405	0.9606
	Sentinel-2	<i>MLC</i>	0.9506	0.9473	0.9730	0.9648	0.9812
		<i>RF</i>	0.9573	0.9547	0.9768	0.9738	0.9798
		<i>SVM</i>	0.9692	0.9672	0.9833	0.9851	0.9816
<i>Deep Learning</i>	Landsat 8	<i>U-Net</i>	0.9050	0.8988	0.9467	0.9319	0.9620
	Sentinel-2	<i>U-Net</i>	0.8280	0.8122	0.8964	0.8163	0.9939

Discussed Points in our paper

- What data was needed?
- How data was acquired?
- Why and what AI image classification techniques were chosen?
- Steps of Data Processing?
- Suitable accuracy measures adopted?
- How was accuracy measured?
- Performance of AI techniques
- Importance of Research and relevance of addressing urbanization issues.

Concluded points in our paper

- ✓ References to key journal articles and papers used
- ✓ Suggestion on software consideration
- ✓ Revealing the technique with the best performance
- ✓ Comments on AI (machine learning and deep learning)
- ✓ Suggestion on enhancing results for unique environments
- ✓ Recommendations

Justification/Significance of the Research

Affordable Solution
(wider user population)

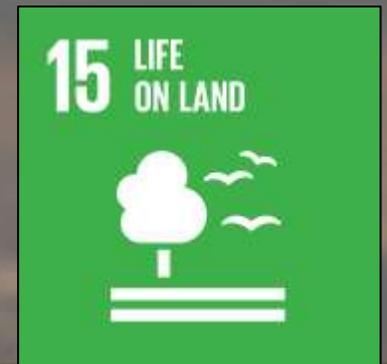
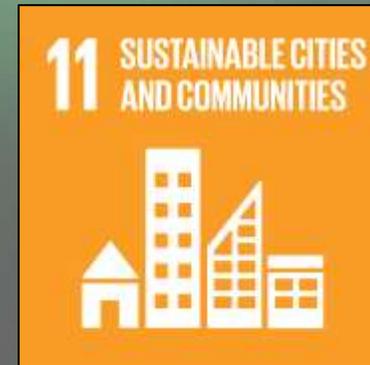
Land Use & Environmental
Planning

Disaster
Management

The United Nations
SDGs **11, 13, and 15**

Generate
Fundamental
Geospatial Datasets

Body of Knowledge
in Surveying, Land
Information and
Geoinformatics



The End – Check out our paper!!!

*Artificial Intelligence Techniques for Extracting Impervious
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