

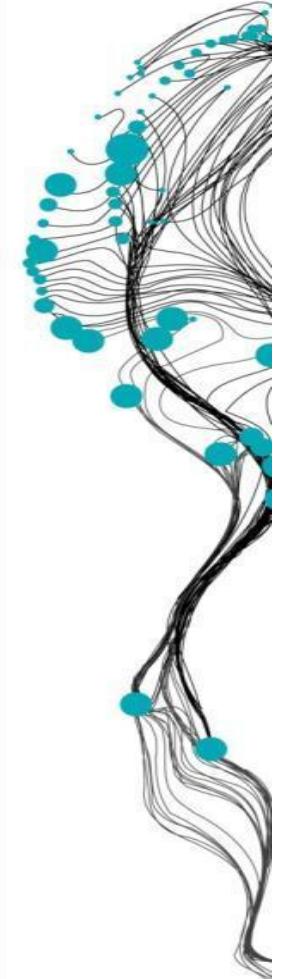
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PROSPECTS FOR GLOBAL MONITORING OF THE SDG SLUM INDICATOR WITH EARTH OBSERVATION

SLIUZAS, R., KUFFER, M., WANG, J., NAGENBORG, M., PFEFFER, K., KOHLI, D.,
AND PERSELLO, C.



FACULTY OF GEO-INFORMATION SCIENCE AND EARTH OBSERVATION



CONTENTS

- Some basic parameters for slum mapping (with EO)
- Slum mapping research at ITC
- Prospects and issues for global slum mapping

The nature of slum dwellers and slums

UN-HABITAT 2002

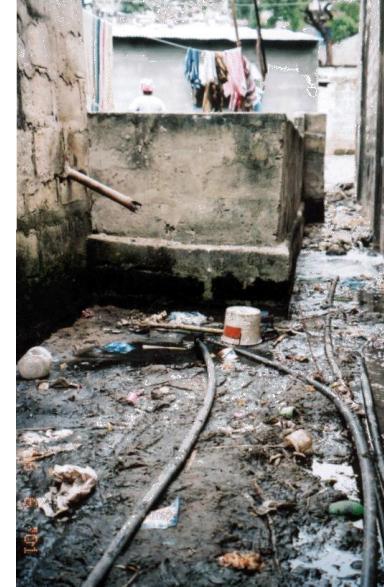
Who are slums dwellers?
Urban households lacking
at least 1 of the following:

- Adequate water
- Adequate sanitation
- Sufficient living space
- Secure tenure
- Durable housing (quality of structures & environment – hazards)

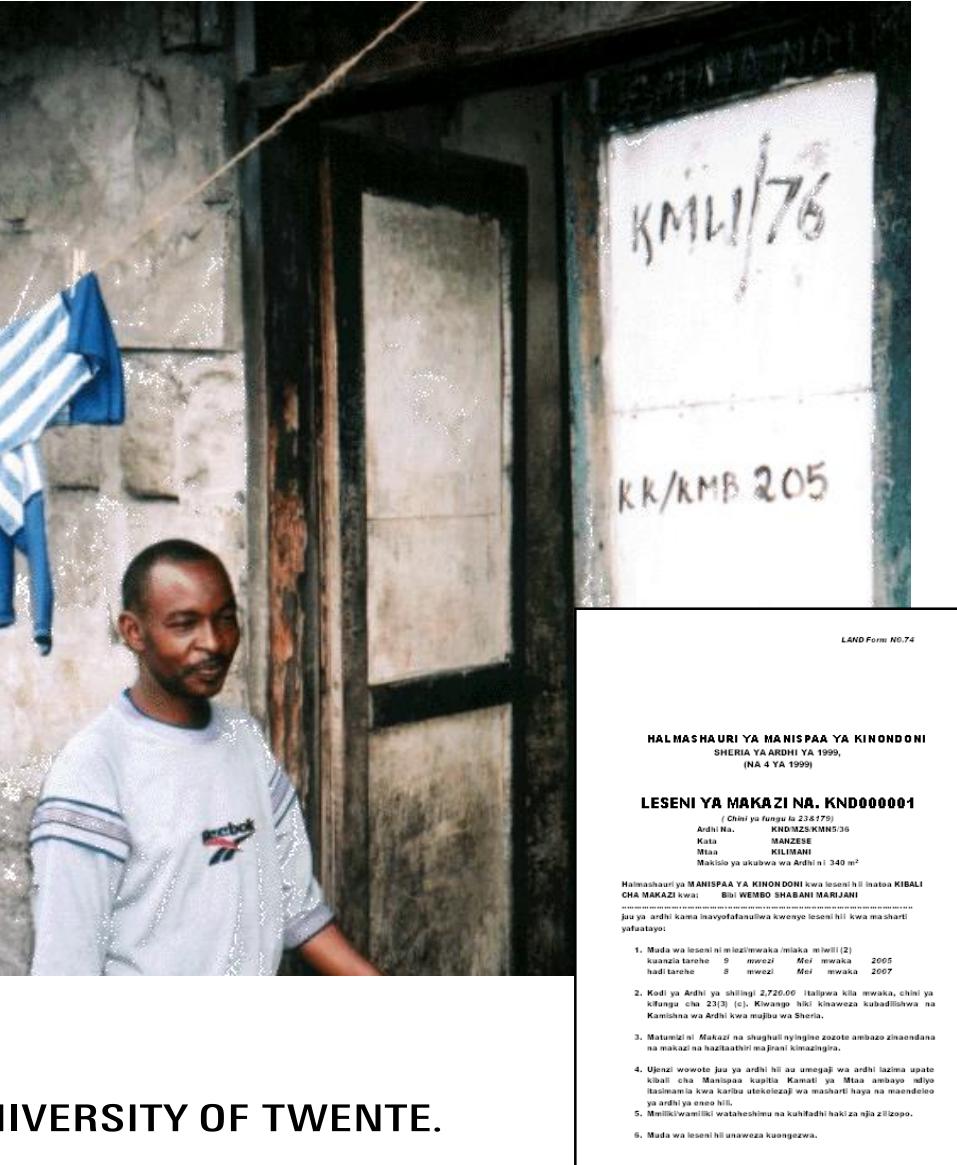
Large scale surveys:
Census, DHS
City and Settlement surveys



INDICATORS FOR SLUM DWELLERS ADEQUATE WATER AND ADEQUATE SANITATION



TENURE SECURITY



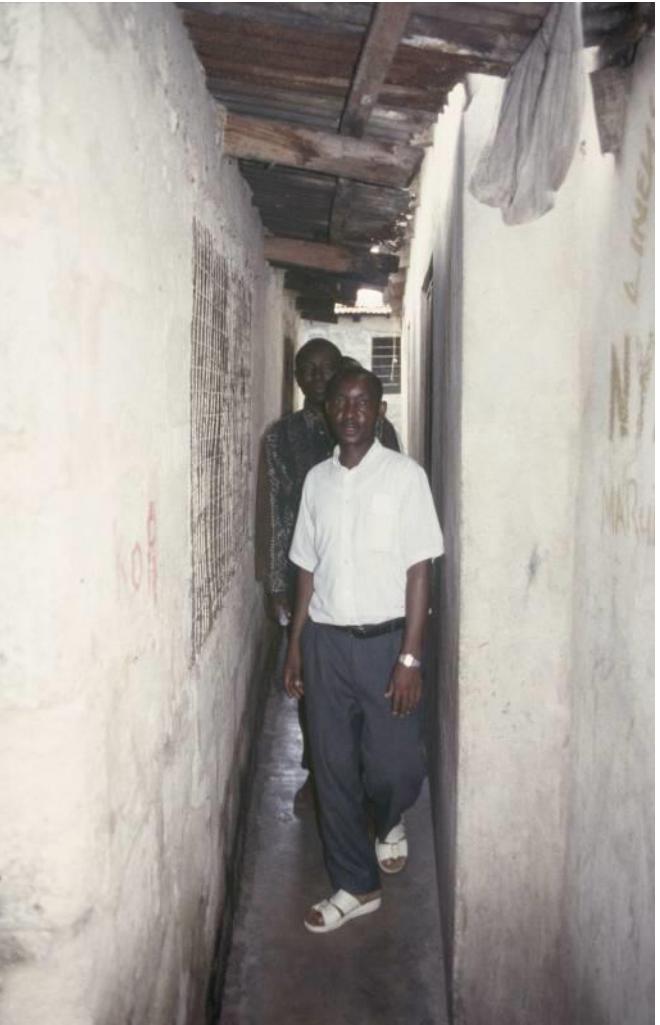
OVERCROWDING



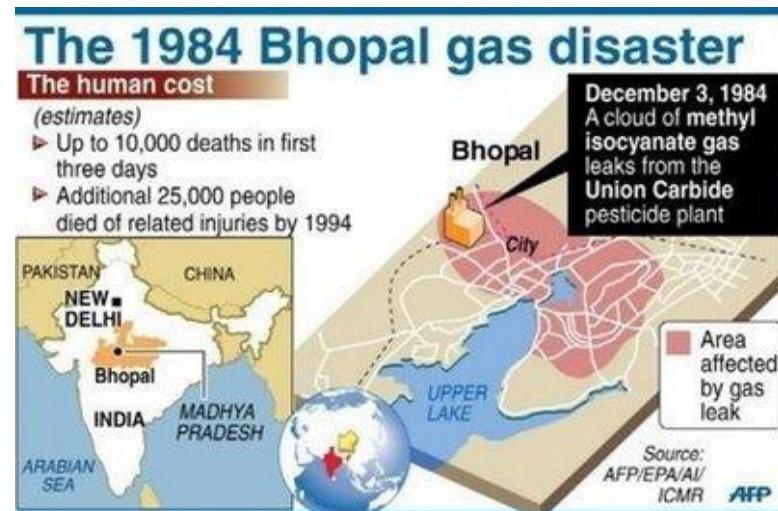
> 3 persons per room

DURABLE HOUSING

PRIVATE VS PUBLIC SPACE, BUILDING AND PLANNING STANDARDS



DURABLE HOUSING: SAFE FROM NATURAL AND TECHNOLOGICAL HAZARDS



HAZARDS AND SLUM FORMATION: DAR ES SALAAM



INFORMAL VS FORMAL URBAN DEVELOPMENT

Adapted from Baros

Occupation (land)

Building

Servicing

**Planning
(regularization,
eviction, resettlement)**

**Monitor (guide)
citizen led
development**



Sliuzas et al., 2017 <https://ieeexplore.ieee.org/document/7924589>

Planning

Servicing

Building

Occupation

**Monitor plan
implementation**



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SLUMS: spatial concentration of slum dwellers - diversity of physical forms and settings



Kampala
Uganda



Cairo
Egypt

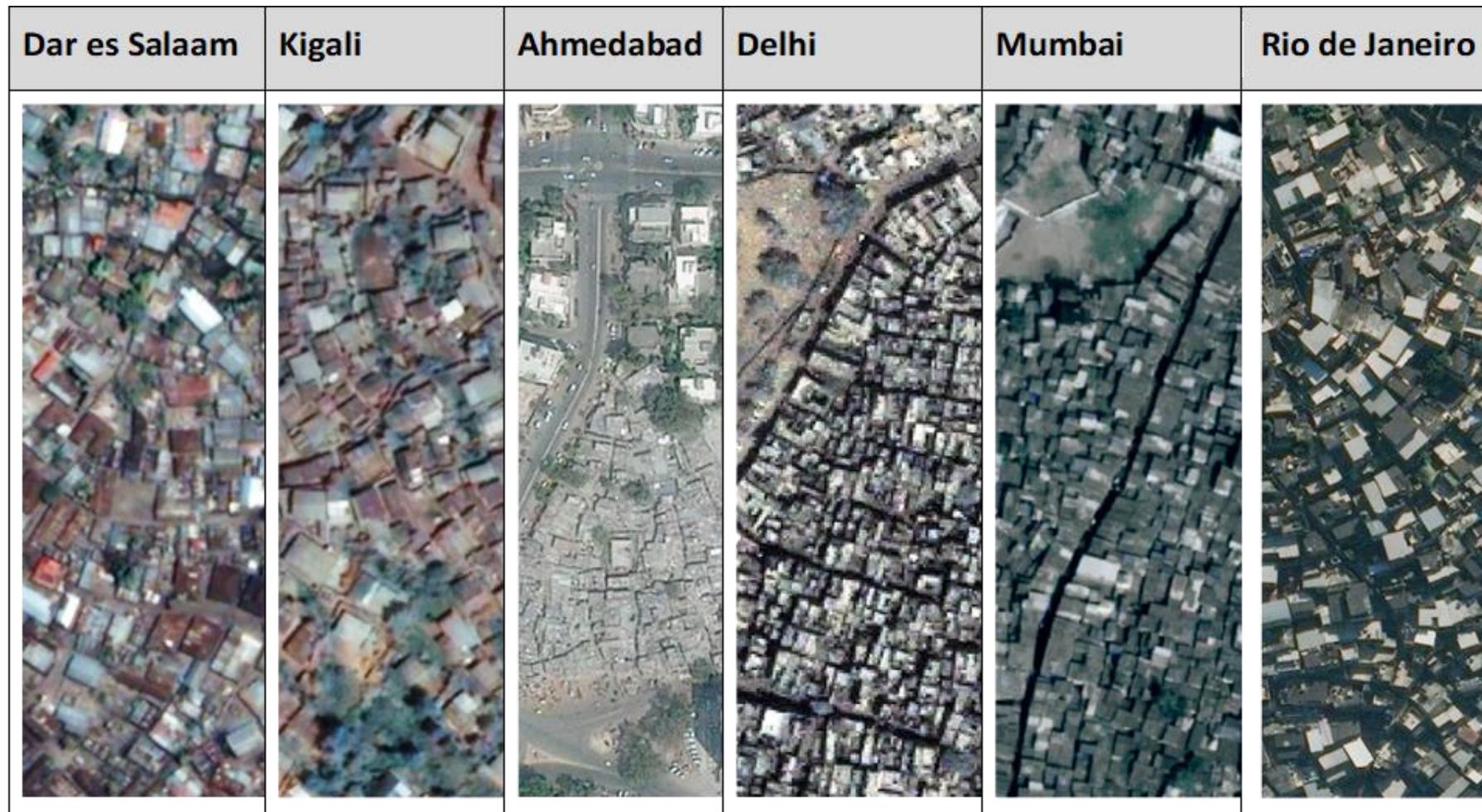


Kisumu
Kenya



Ahmedabad
India

THE URBAN DIVIDE – THE MORPHOLOGY



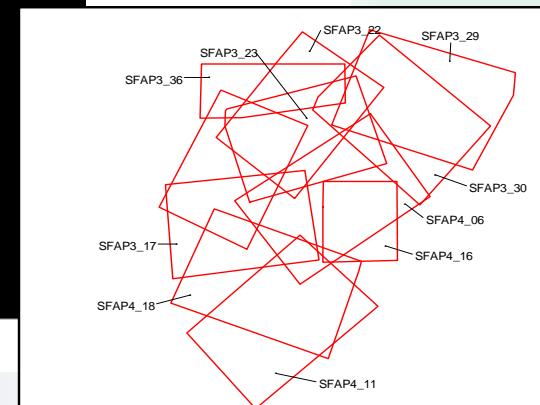
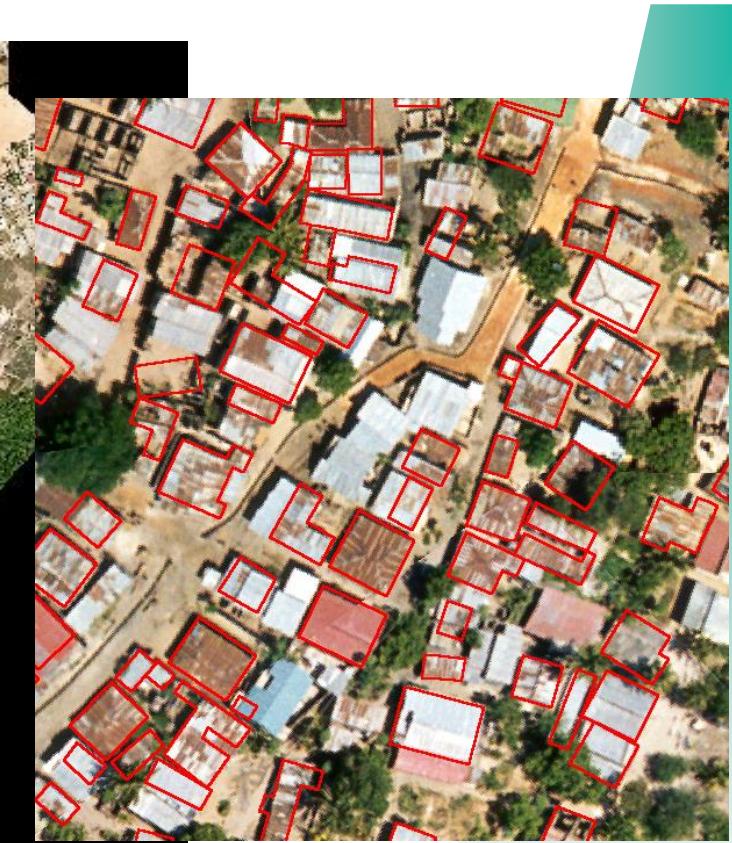
MORPHOLOGY OF SLUMS – FROM SPACE

What is specific to slums?

Features	Slums	Planned areas
Size	<ul style="list-style-type: none">• Small building sizes	<ul style="list-style-type: none">• Generally larger building sizes
Density	<ul style="list-style-type: none">• High densities (roof coverage)• Lack of public (green) spaces	<ul style="list-style-type: none">• Low – moderate density areas• Provision of public (green spaces)
Pattern	<ul style="list-style-type: none">• Organic layout structure	<ul style="list-style-type: none">• Regular layout pattern
Site Aspects	<ul style="list-style-type: none">• Hazardous locations• Access to livelihood opportunities• Etc...	<ul style="list-style-type: none">• Formal development with services and infrastructure provision



SLUM MAPPING FROM SMALL FORMAT AERIAL PHOTOS



POINT CLOUD FROM UAV IMAGES, KIGALI, RWANDA FOR 2D AND 3D ANALYSIS AND PLANNING

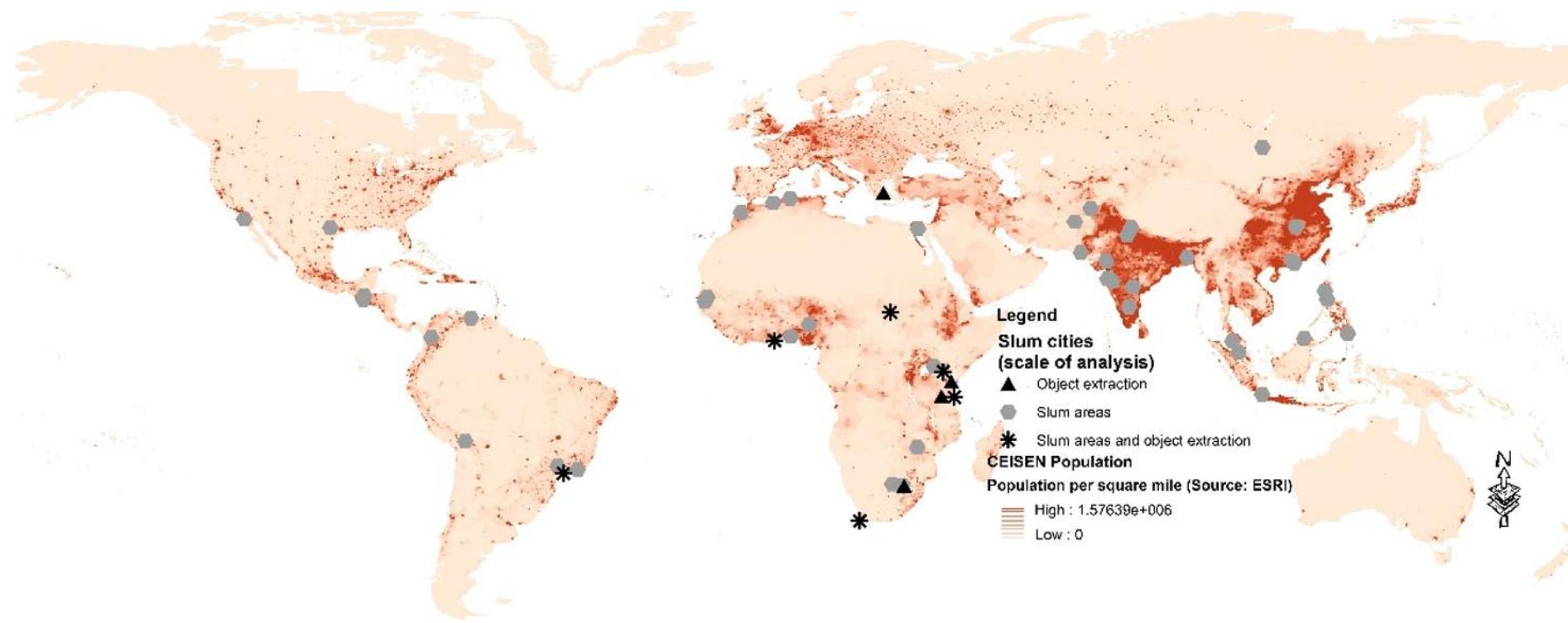


(IMAGE BY C. GEVAERT)



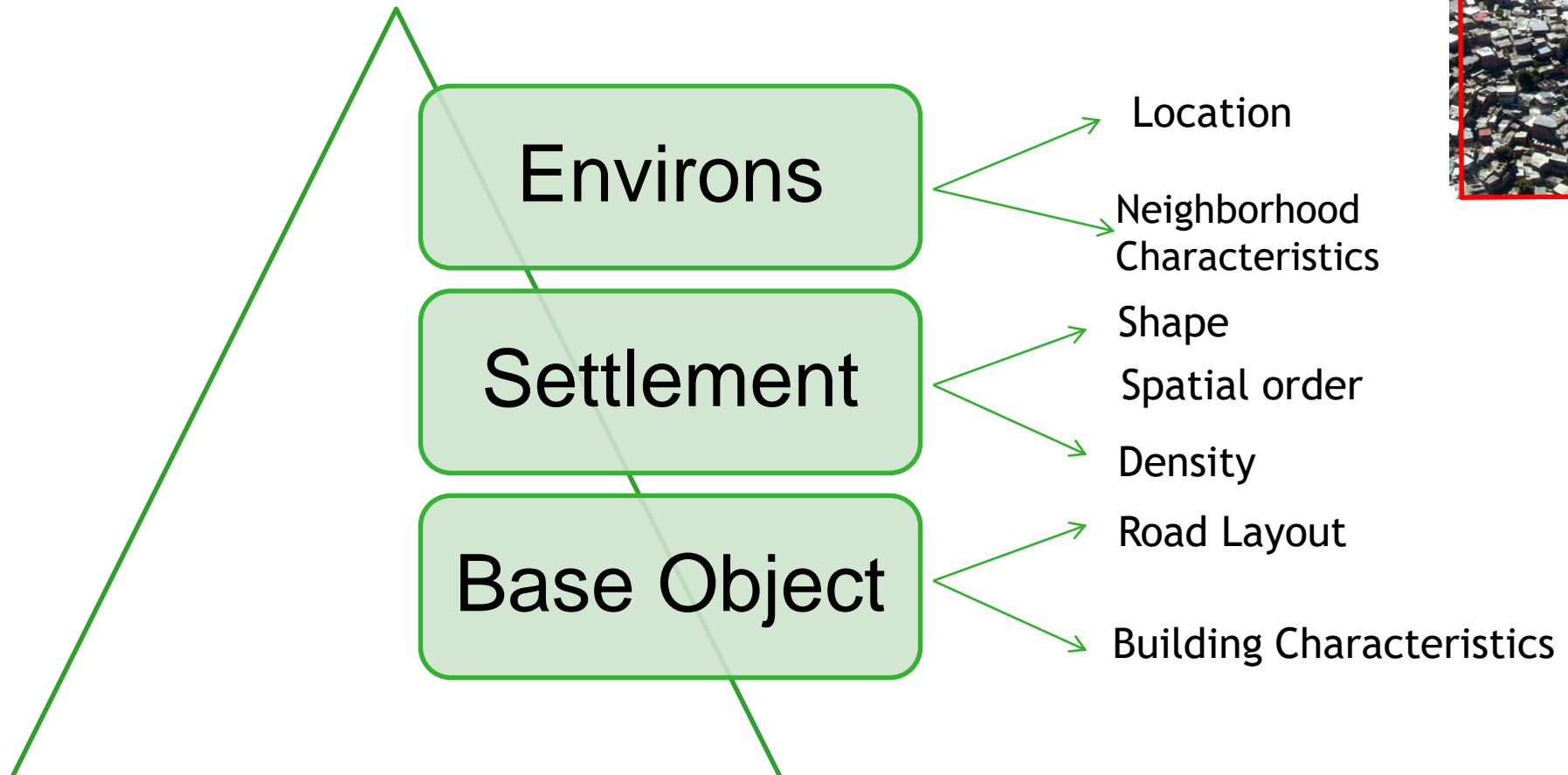
THE URBAN DIVIDE

What do we know about global slum developments



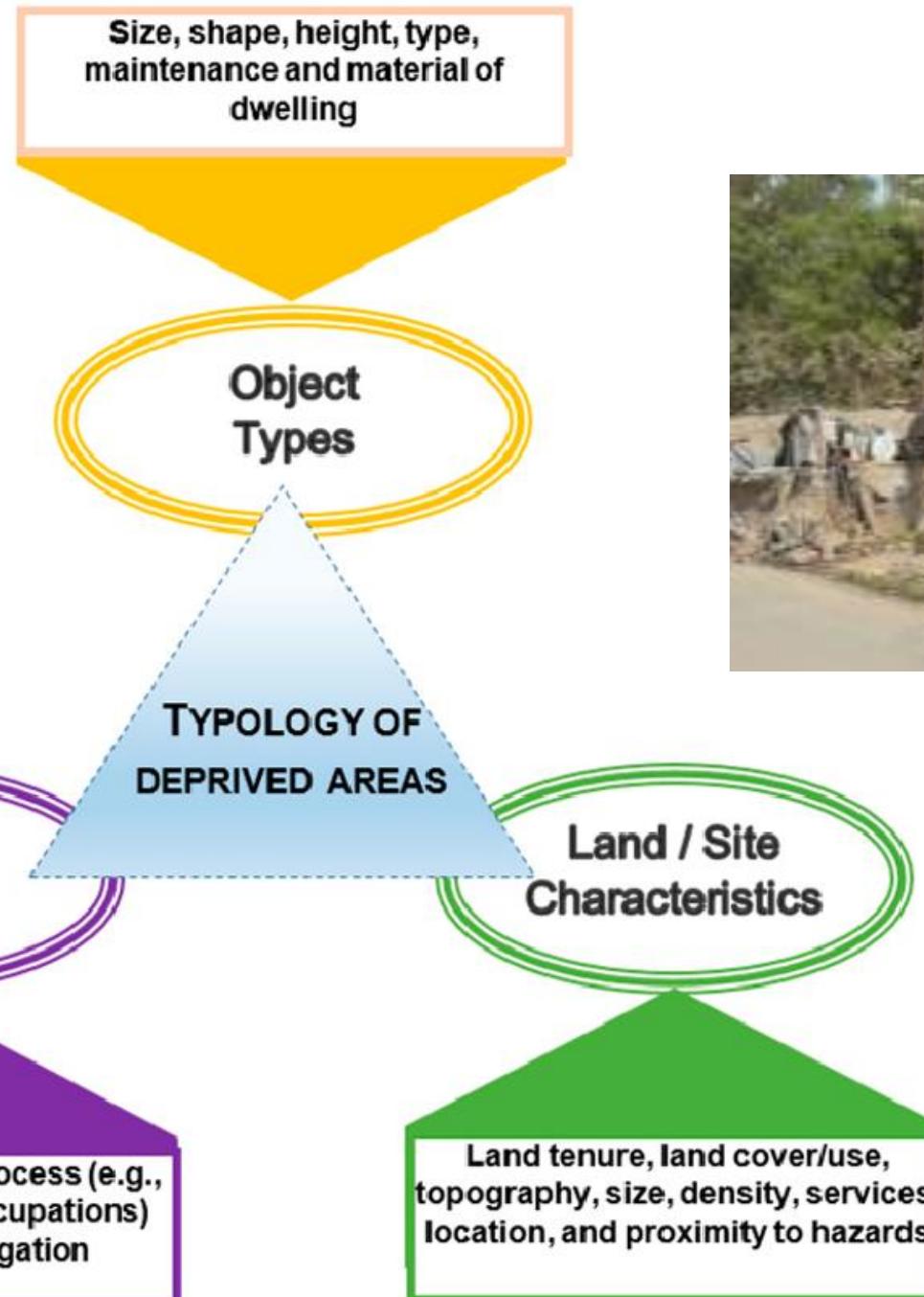
- 15 years of slum mapping using remote sensing
(Kuffer, Pfeffer and Sliuzas, 2016)
- Based on 87 publications selected and reviewed

THE GENERIC SLUM ONTOLOGY

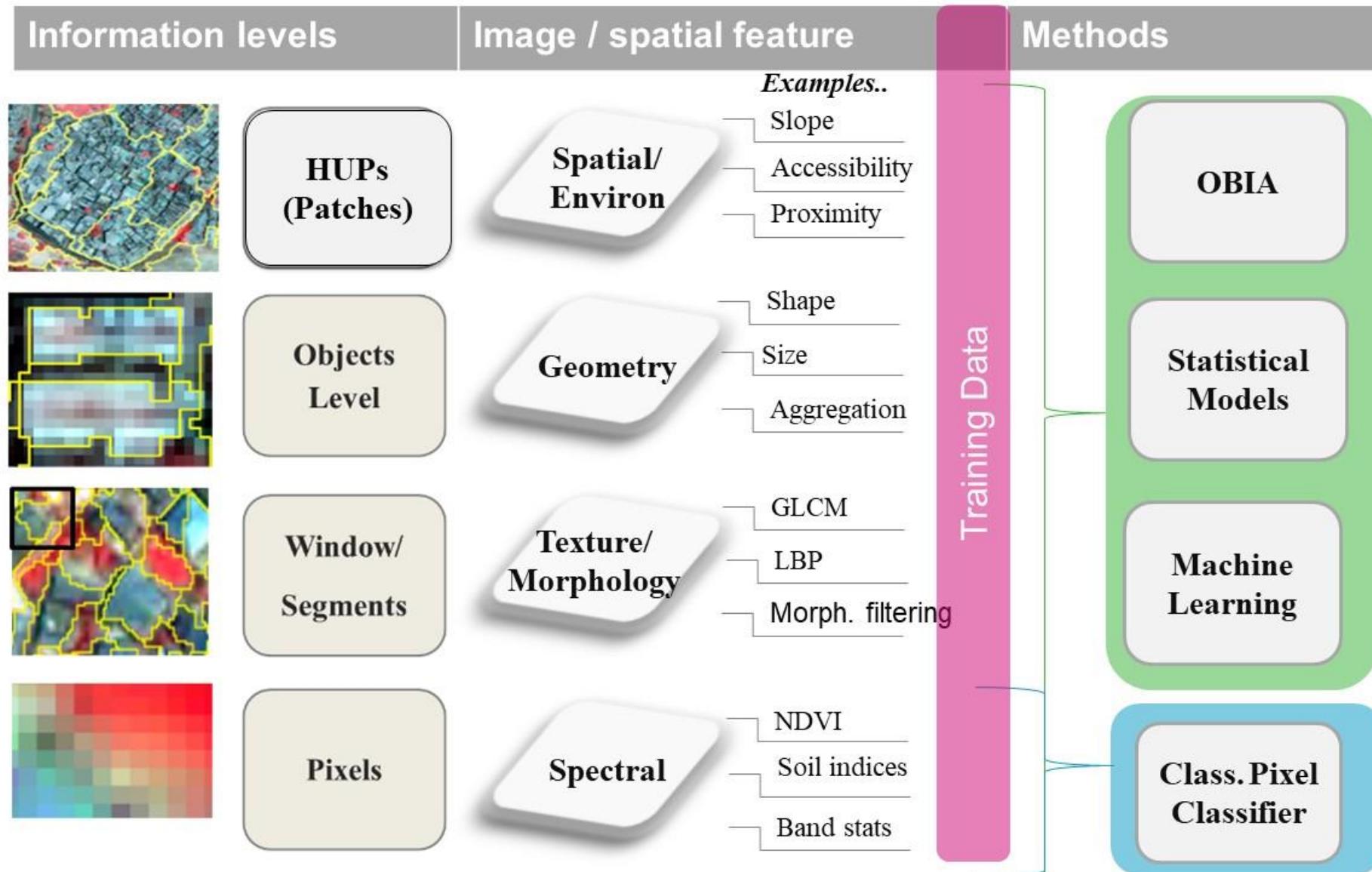


Kohli, D.; Sliuzas, R.V.; Kerle, N.; Stein, A. An ontology of slums for image-based classification. *Comput. Environ. Urban Syst.* **2012**, *36*, 154–163.

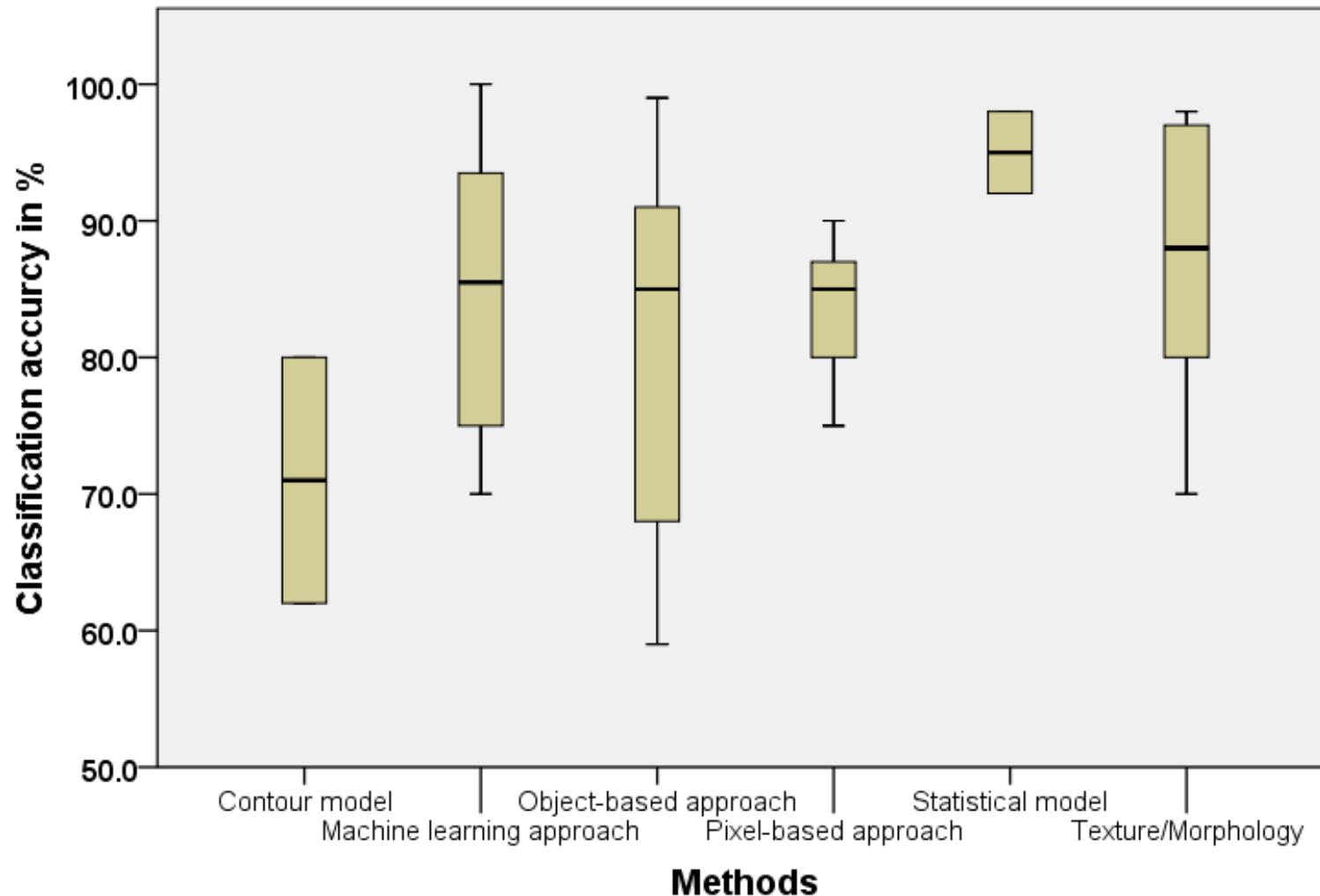
DIFFERENCES



SLUM MAPPING

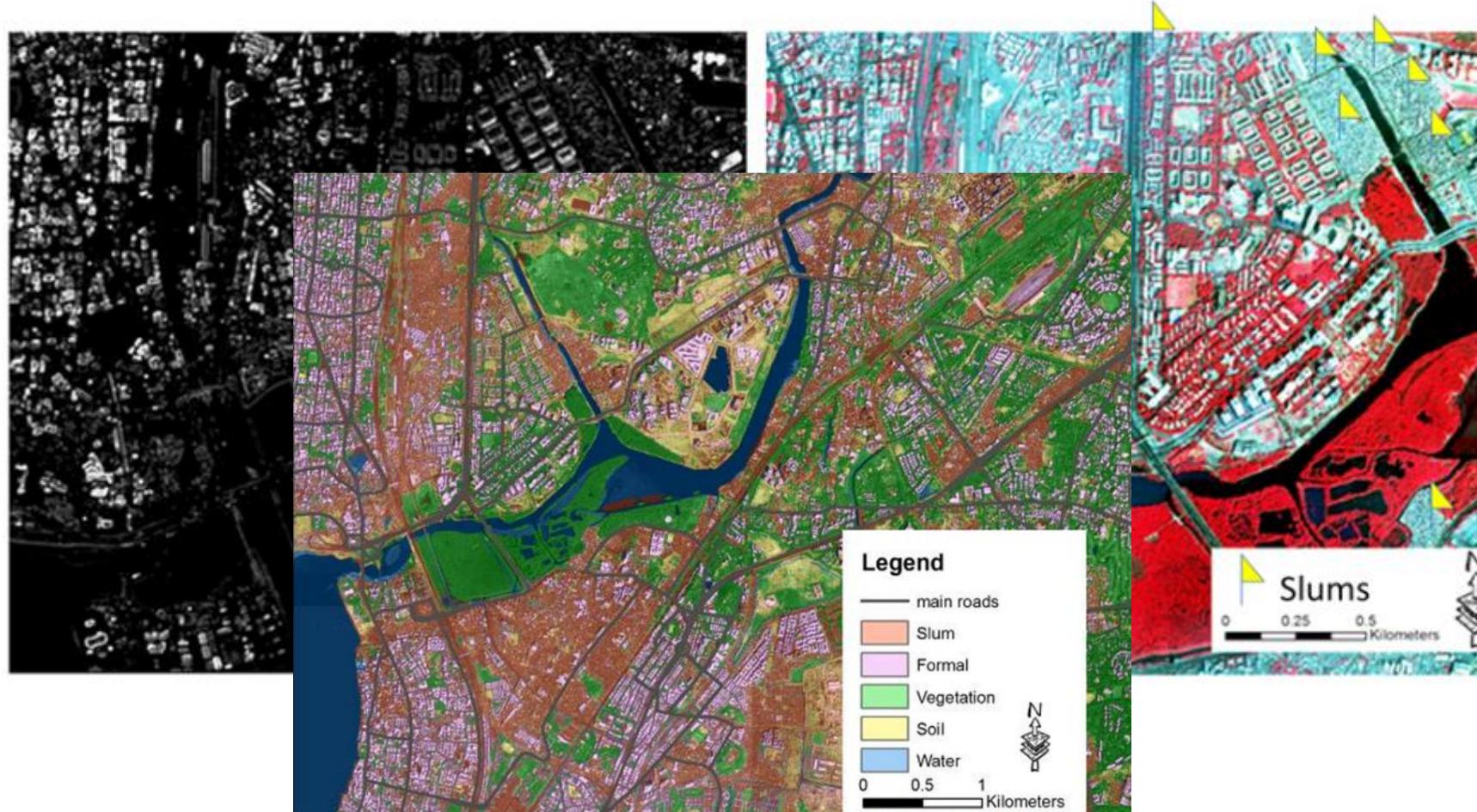


ACCURACIES OF METHODS TO MAP SLUMS



THE URBAN DIVIDE AND MACHINE LEARNING

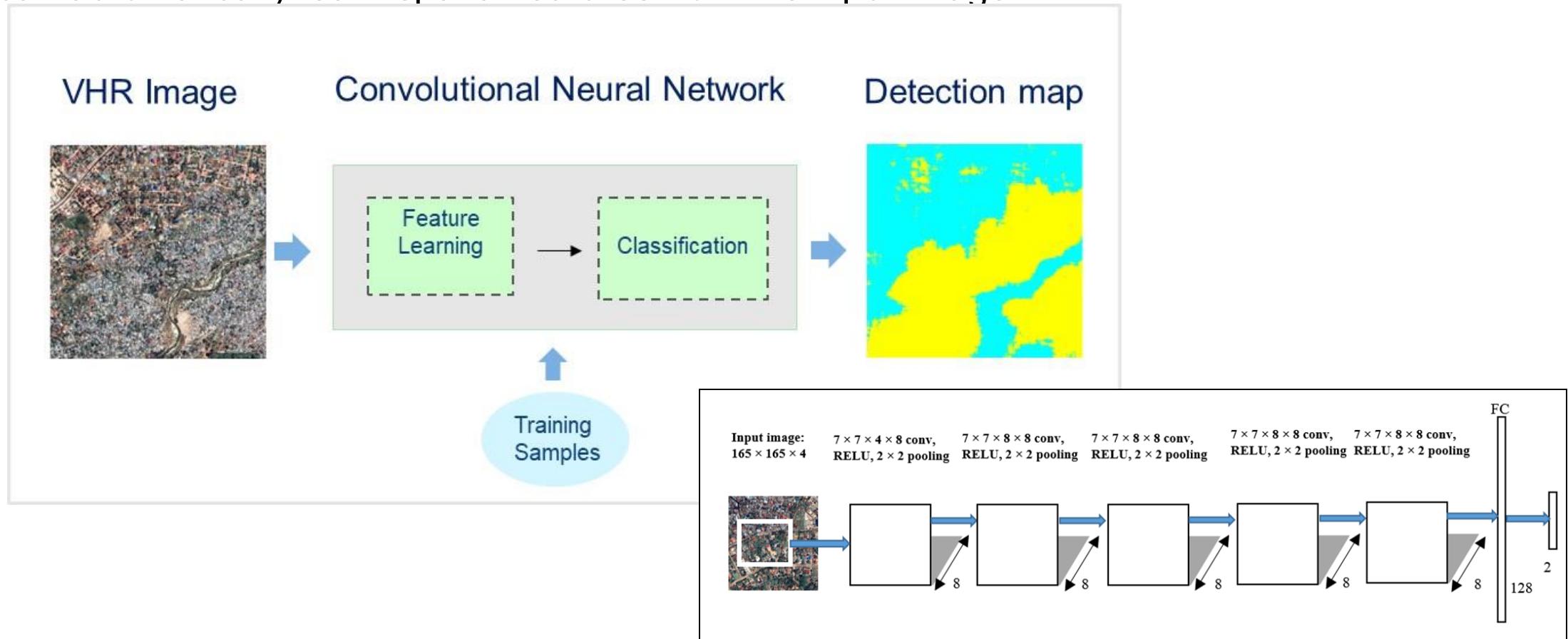
GLCM (Gray Level Co-Occurrence Matrix) - Example Mumbai



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THE URBAN DIVIDE - DEEP LEARNING APPROACH

Deep learning methods such as **Convolutional Neural Networks** can automatically learn spatial features from the input image.



Mboga, Persello, Bergado, Stein, "Detection of Informal Settlements from VHR Satellite Images using Convolutional Neural Networks, IGARSS 2017.

DEGREE OF DEPRIVATION USING CNNs



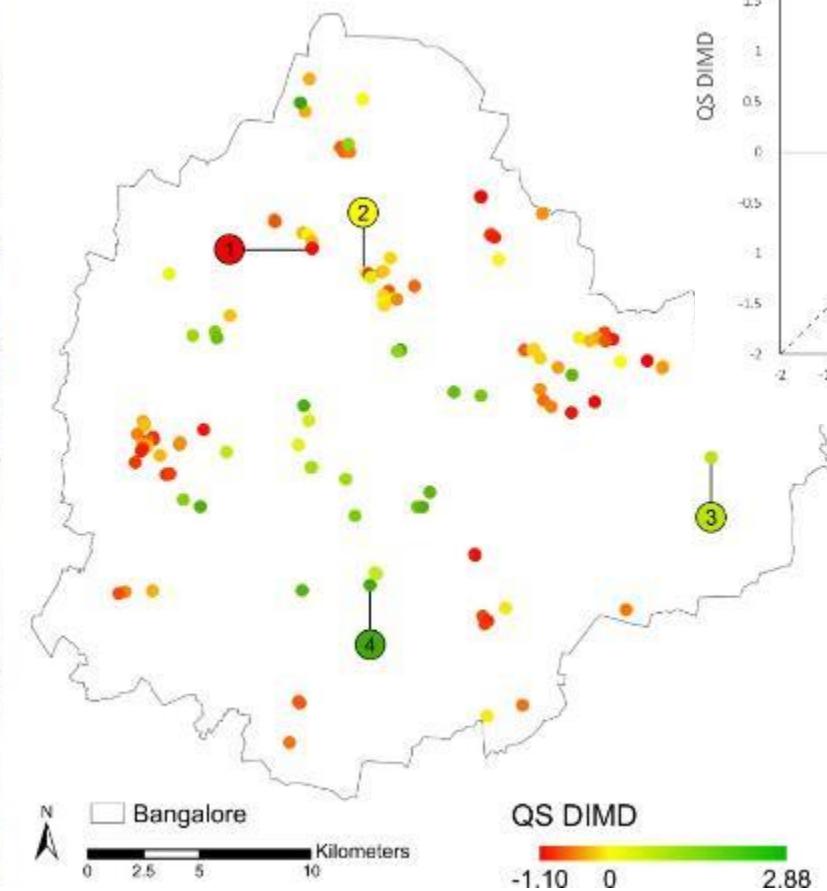
CNN-based model Transfer learning

Classification problem
Distinguishing slum from formal
2000 samples for training

Distinctive features

Regression problem
Predicting Deprivation indices
<121 samples for training

CNN model



Source: A. Ajami et al. forthcoming



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PROSPECTS FOR GLOBAL SLUM MAPPING?

- Producers, uses and users
- Incorporating different slum development stages, dynamics and typologies?
- Feature selection – training – assessment – which algorithms and reference data?
- Transferability of methodology (temporal – spatial)?
- How to upscale to global level?
- Suitable data (spatial resolution, cost...)



PRODUCERS, USES AND USER NEEDS

- Slum mappers: government, researchers, communities, NGOs
- Better understanding user requirements – bridge communication gap
- Making products relevant to support user needs
- Co-production of slum maps and data
- Data access, distribution and maintenance (slum mapping as a social-technical infrastructure)



TEMPORAL DYNAMICS



A) 2008



B) 2012

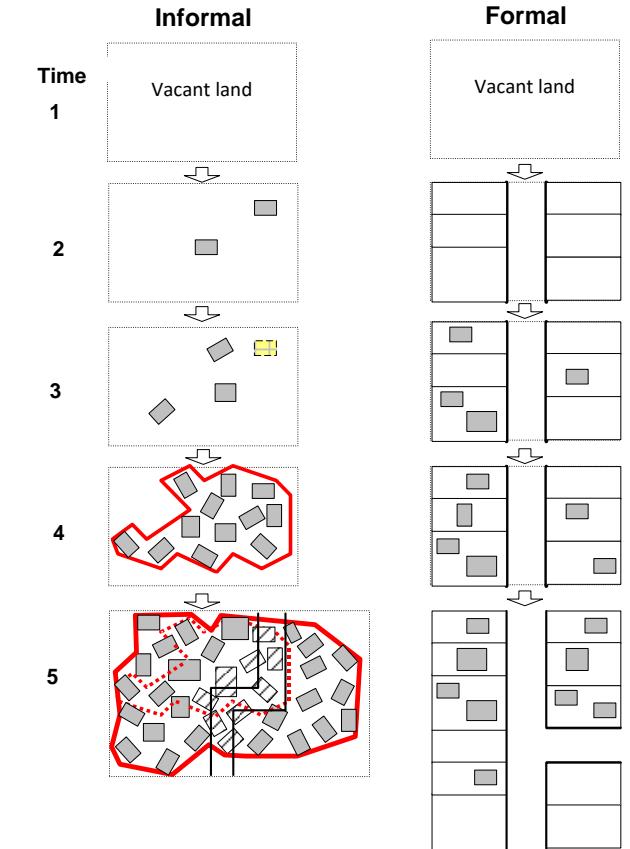


C) 2013



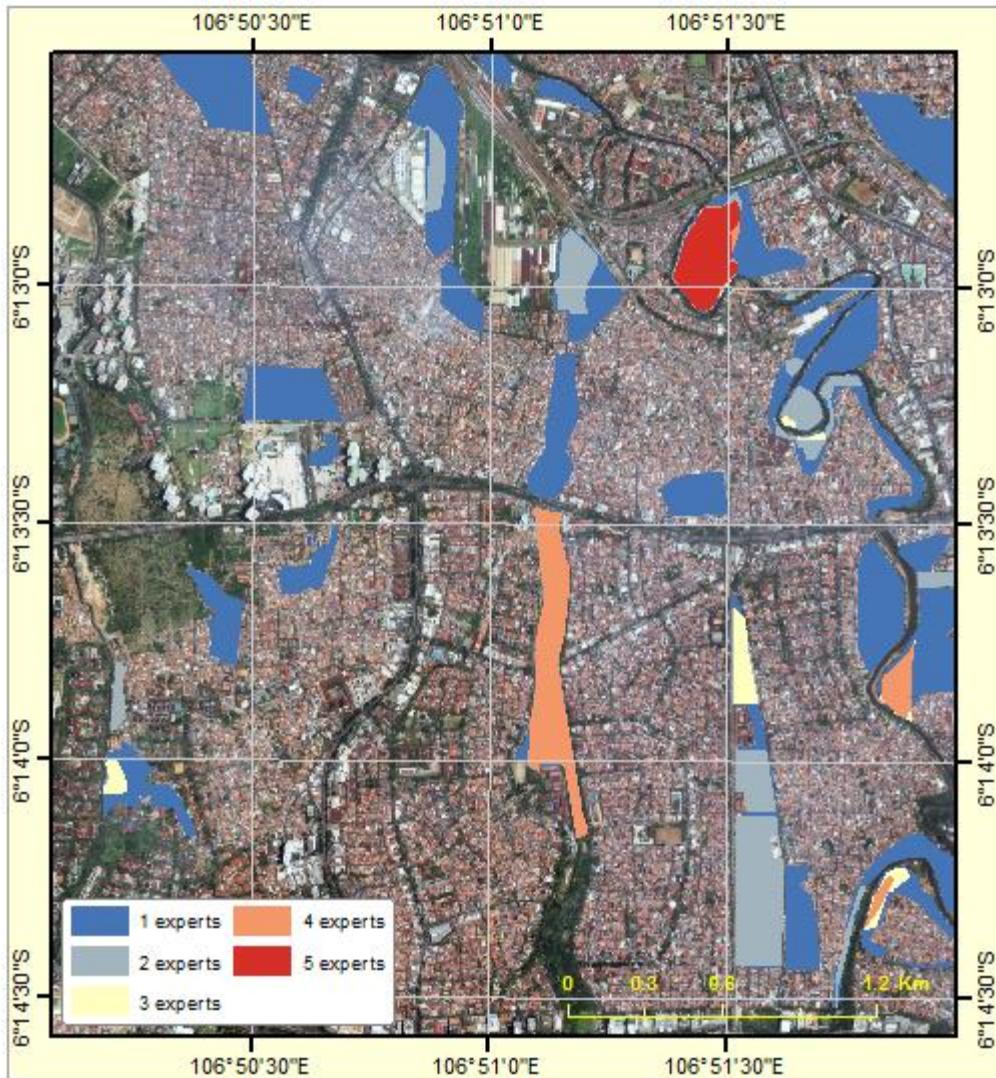
D) 2015

Emergence and Growth of a slum in Huidi, Bangalore (marked with a red polygon). a) Slums emerge near a construction Site in 2008. b) Slum grows near the same site. c) Slum disappear when construction is complete in 2013. d) A slum re-emerge at the same site in 2014 (Images— Google Earth) (Source: Dynaslam)



Rangelova et al. 2017
<https://www.tandfonline.com/doi/full/10.1080/22797254.2018.1535838>

UNCERTAINTIES IN THE REFERENCE DATA



In area with higher agreement shows (on the ground): poor building materials, high density and

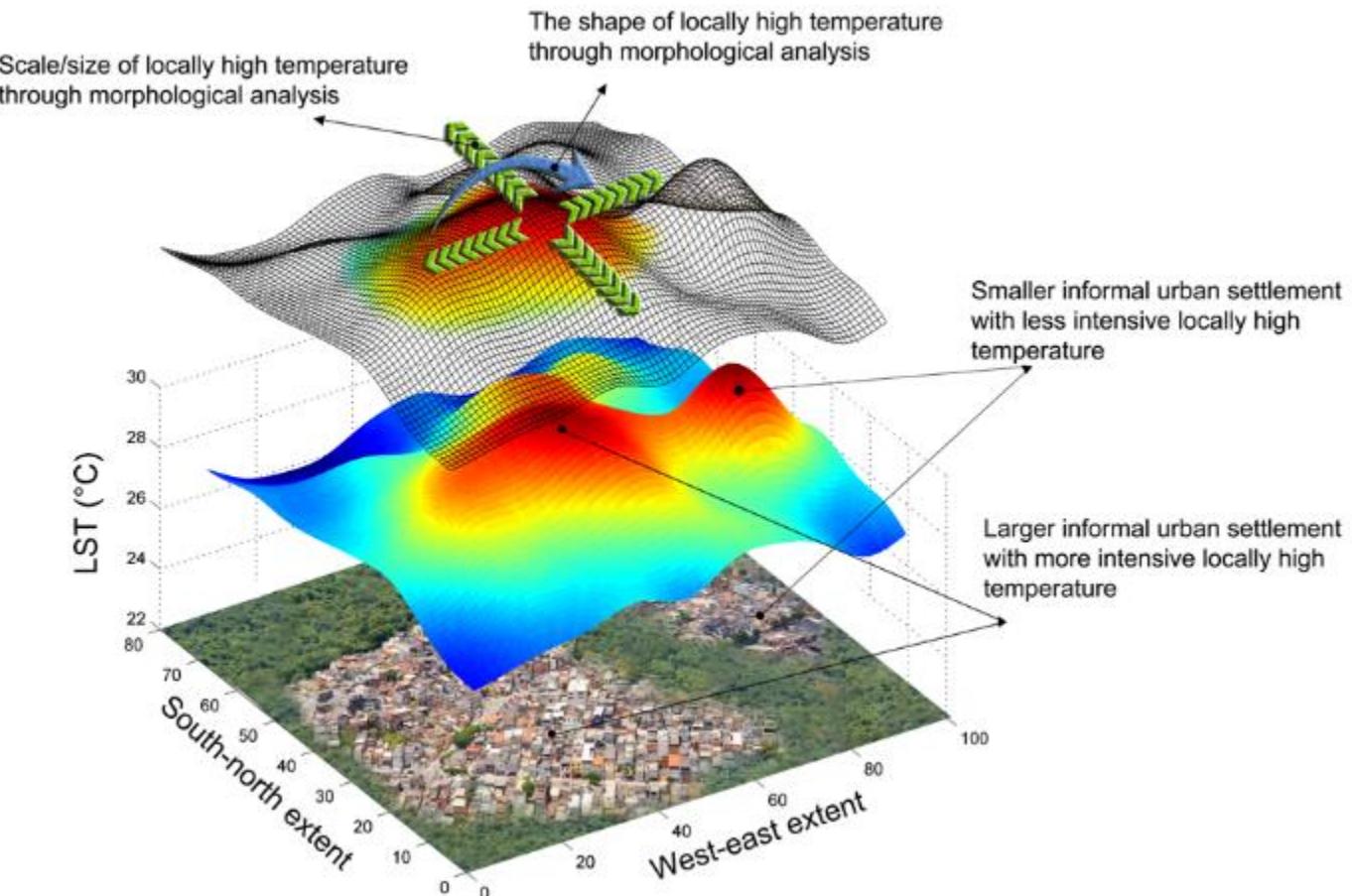


Misclassifications: high density and have a roof from asbestos, but not a slum

Opportunity to link to NGOs and communities slum dweller groups

Source: Pratomo et al., 2017:
<https://www.mdpi.com/2072-4292/9/11/1164>

UNDERSTAND BETTER ENVIRONMENTAL CONDITIONS OF SLUMS: HAZARDS – CLIMATE

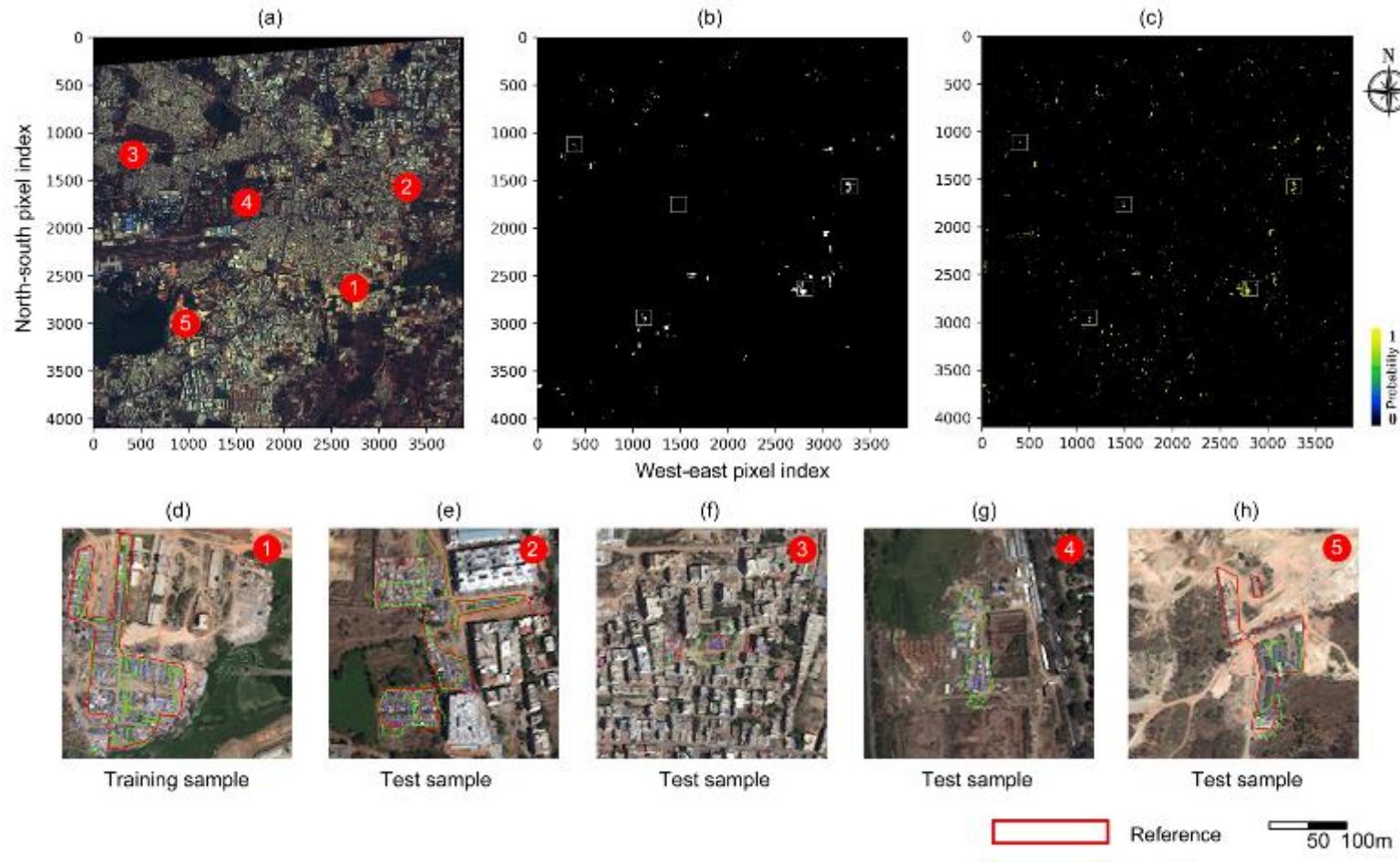


WANG, J., Sliuzas, R., Kuffer, M., Kohli, D.

<https://www.sciencedirect.com/science/article/pii/S0048969718337811>

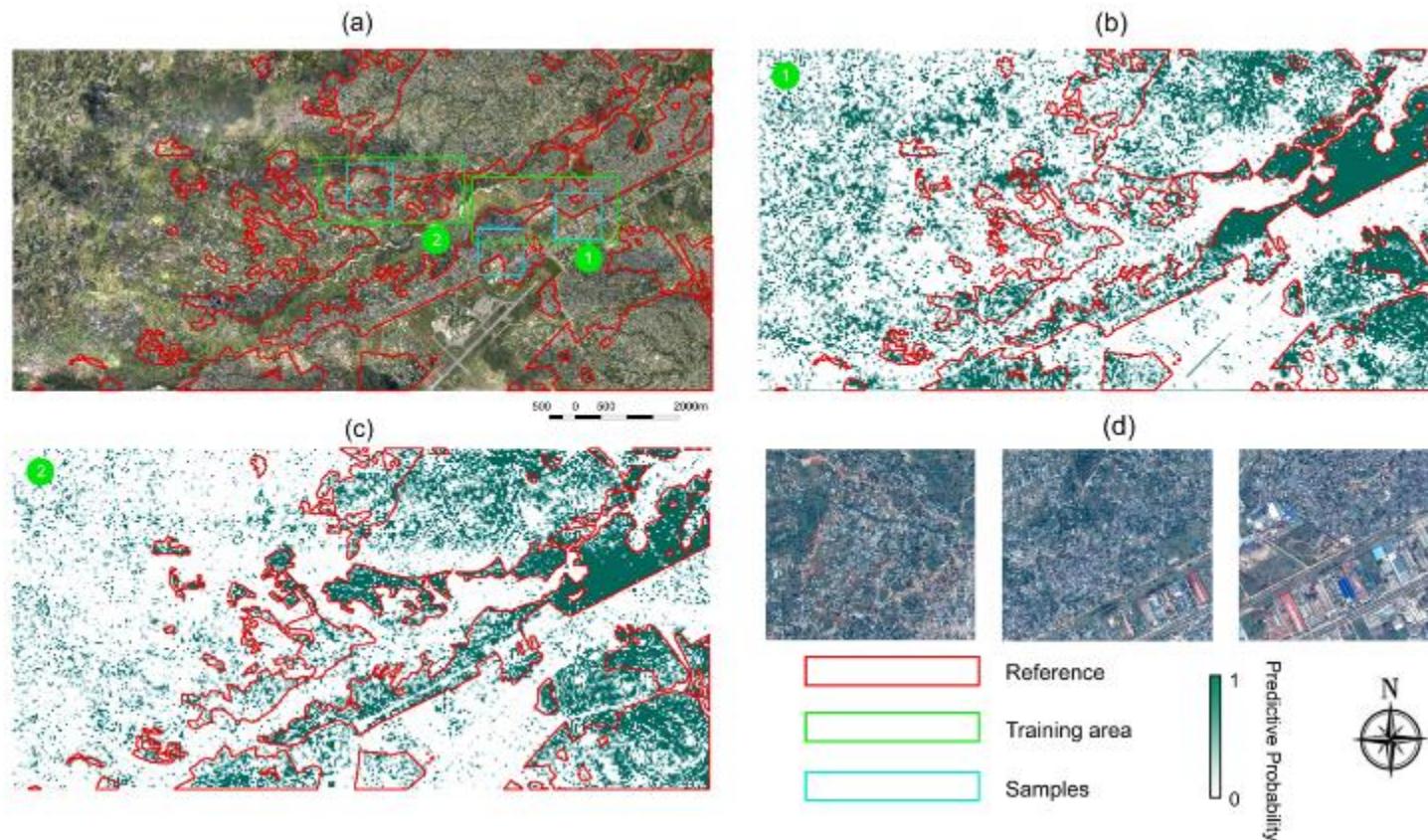
GOING AHEAD: CAN MAP SLUMS WITH CNNs BASED ON LIMITED TRAINING DATA

- Mapping small clusters of slums with training based on few large slums



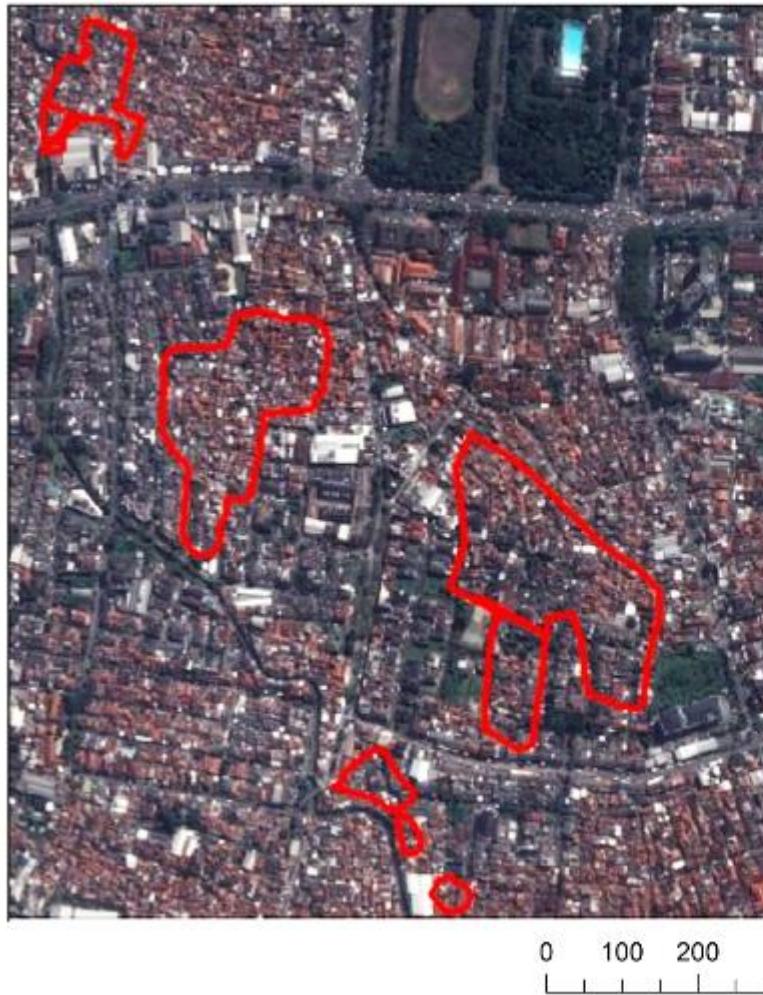
GOING AHEAD: CNNs NEED TO HAVE TRAINING DATA THAT INCLUDE THE VARIATIONS

- CNNs need to be trained based on the full variety of their morphologies



UNCERTAINTIES ON SLUM BOUNDARIES: BANDUNG, INDONESIA

Local authorities

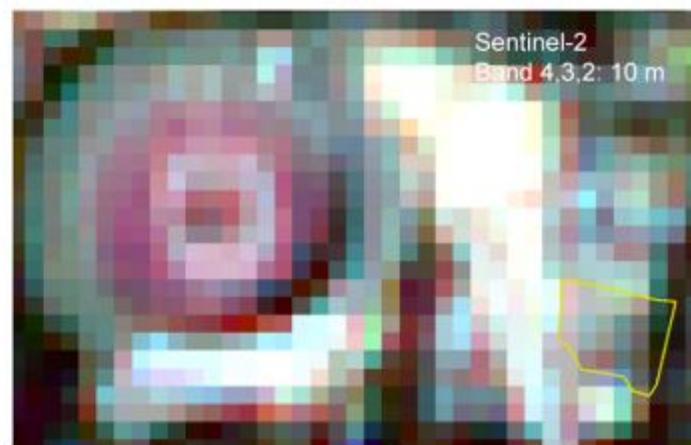


Ground-truth delineation including image



MOST SUITABLE SPATIAL RESOLUTION OF IMAGES

Benefits versus image and computational costs



HOW CAN WE SHOW THE FUZZINESS IN MAPS

- Is the highest detail necessary?
- Ethical considerations (not) making data on slums publically available?



1

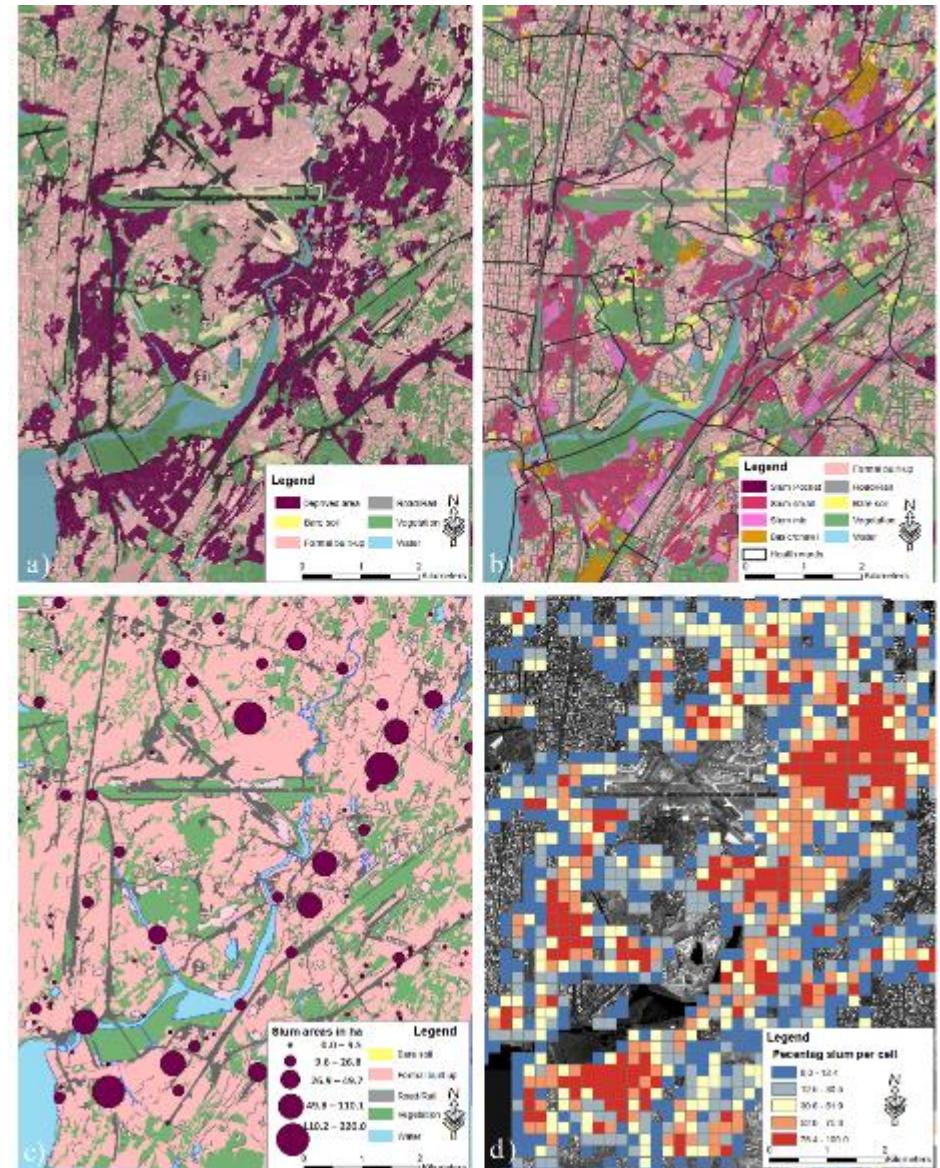


Kampungs with basic facilities, amenities, durable housing materials, cars

2

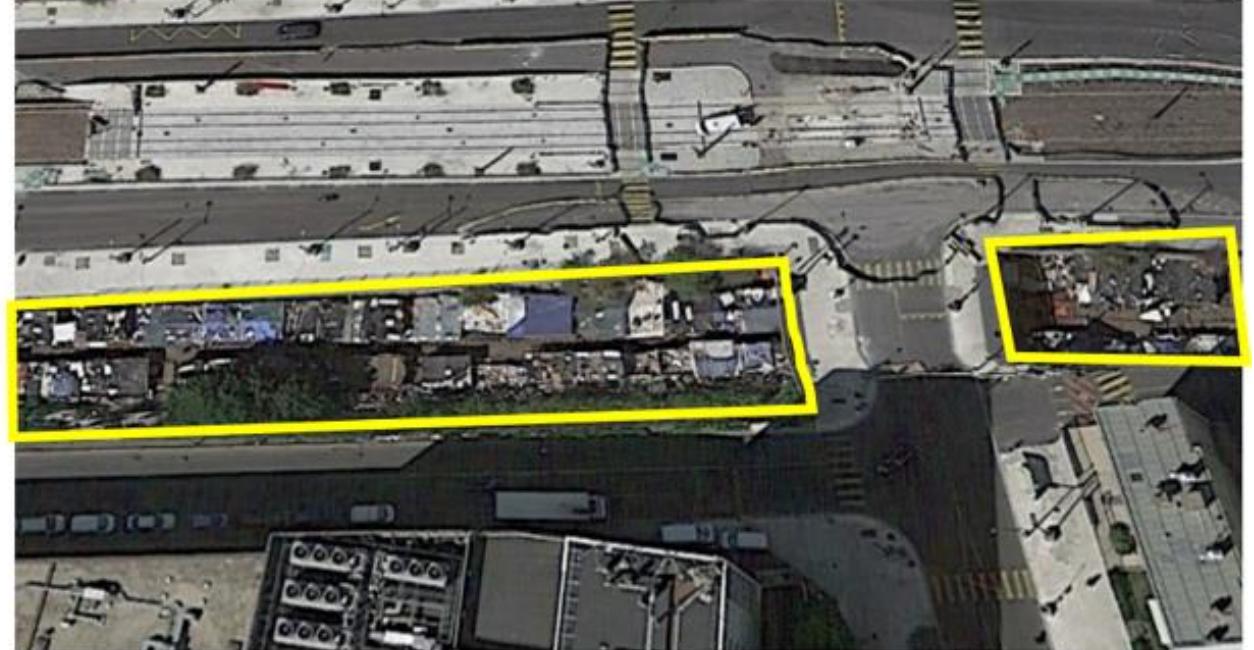


Kampungs without basic facilities, poor housing materials, poor households



SLUMS IN EUROPE?

Immigrants in France. (L) Eviction from Calais, (R) New settlements in Paris



INFORMATION NEEDS AND ETHIC CONSIDERATIONS



Shall we make slum maps and images publically available ????

<https://www.sicherheitspolitik-blog.de/2018/07/11/the-digitalization-of-the-globe-machine-learning-about-population-in-need-of-support/>



POSSIBLE ETHICAL CONCERNS IN SLUM MAPPING

- Who decides and who owns the process?
- Who is eligible for compensation and resettlement?
- Who pays?
- Issues of possible eviction or economic displacement – gentrification?
-

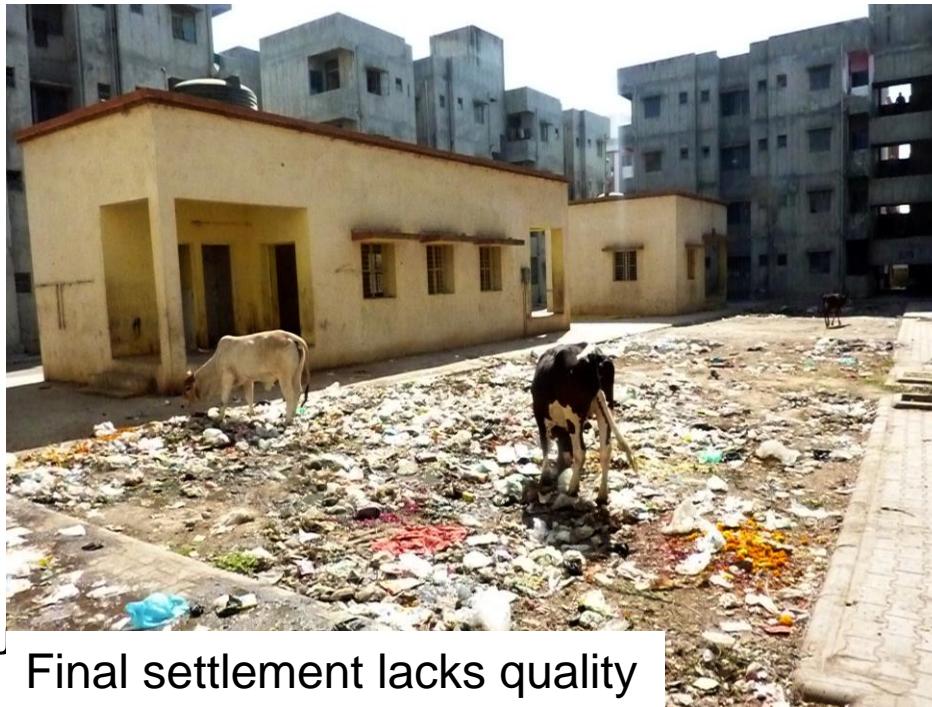
SLUM EVICTION IN AHMEDABAD INDIA

LEADS TO FURTHER DEPRIVATION RELATED MOSTLY TO SERVICE LEVELS AND LOCATION OF NEW SITES

Patel, S., Sliuzas, R., & Mathur, N. (2015).
<http://doi.org/10.1177/0956247815569128>



Many residents do not qualify



Final settlement lacks quality



Very poor temporary resettlement



SOME KEY ISSUES AND QUESTIONS

- Definitions: do we really have a global definition?
 - Slums are often not binary (slum vs non-slum)
 - How do we bring in hazards in an effective manner (also non-binary, dynamic and related to
- Diversity
 - Should we differentiate at regional, country or city level?
 - At least we will need to include training sets that reflect diversity
 - What do slum development processes imply for training samples and processing?
 - *This is shown on the slides 4&5 – we need to include different development stages – but this makes the analysis complex!*
- How to connect local actors and communities (SDI etc.)?
 - In data collection efforts for sample generation?
 - In validation of slum classification maps?
 - As users of the data in daily management and upgrading, etc.?



SOME KEY ISSUES AND QUESTIONS

- Uses and users:
 - Which potential uses have priority and for whom?
 - Will political and other actors be prepared to accept and use such data sets derived from advanced image analysis?
 - What are the margins for error and will these be context dependent?
- Technical
 - Image availability and sensor types
 - Computational power – which processing facilities can support the level of computation required for this task and can these be accessed as and when needed?
 - How to best connect to socio-economic datasets (Census, DHS, MICS etc.)
- Ethical and privacy issues
- Social-technical: what would an inclusive global slum mapping infrastructure look like and how to build and maintain it?



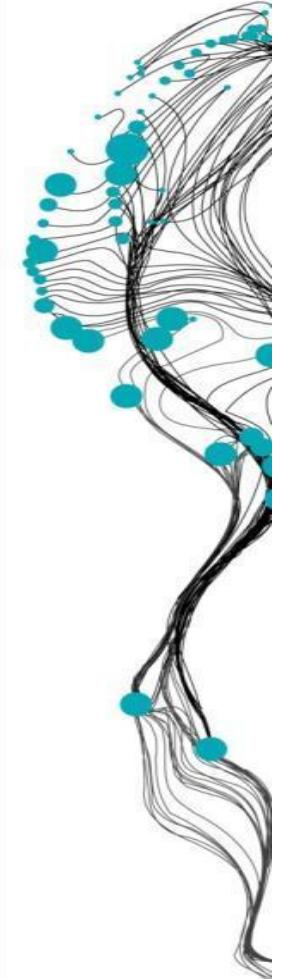
NEW INITIATIVES

Opportunities

- SDG process
- MAUPP partners and follow-up projects
- Group on Earth Observation – Human Planet Initiative
(<https://www.earthobservations.org/index.php> <https://www.itc.nl/hpi-forum/>)
- Global programmes related to hazards and climate change (UNISDR, UNFCC)
- UN-HABITAT - GUO, Slum Upgrading Programme, Climate Change Unit: Building the Climate Resilience of the Urban Poor Initiative for UN Summit 2019

**SLUMS ARE NOT JUST
THERE ‘LIKE THAT’**

Source: Ralf Graf. RxAxLxF Informal City



SOME USEFUL REFERENCES

- Gevaert, C. M., Sliuzas, R., Persello, C., & Vosselman, G. (2018). Evaluating the societal impact of using drones to support urban upgrading projects. *ISPRS International Journal of Geo-Information*, 7(3). <https://doi.org/10.3390/ijgi7030091>
- Kuffer, M., Wang, J., Nagenborg, M., Pfeffer, K., Kohli, D., Sliuz, ... Persello, C. (2018). The Scope of Earth-Observation to improve the consistency of the SDG slum indicator. *International Journal of Geo-Information*, 7(428), 1–28. <https://doi.org/10.3390/ijgi7110428>
- Kuffer, M., Pfeffer, K., & Sliuzas, R. (2016). Slums from space-15 years of slum mapping using remote sensing. *Remote Sensing*. <https://doi.org/10.3390/rs8060455>
- Kohli, D., Sliuzas, R., Kerle, N., & Stein, A. (2012). An ontology of slums for image-based classification. *Computers, Environment and Urban Systems*, 36(2), 154–163.
- Mahabir, R., Croitoru, A., Crooks, A. T., Agouris, P., & Stefanidis, A. (2018). A Critical Review of High and Very High-Resolution Remote Sensing Approaches for Detecting and Mapping Slums: Trends, Challenges and Emerging Opportunities. *Urban Science*, 2(8), 1–38.
<https://doi.org/10.3390/urbansci2010008>