

Collaboration, Innovation and Resilience: Championing a Digital Generation

Big Data and Smart Cities

Trends and Challenges

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Collaboration, Innovation and Resilience: Championing a Digital Generation



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Big Data and Smart Cities Trends and Cha

- 2D Smart City to 3D Smart City
- Sensors and Big Data
- Digital Twins and Al



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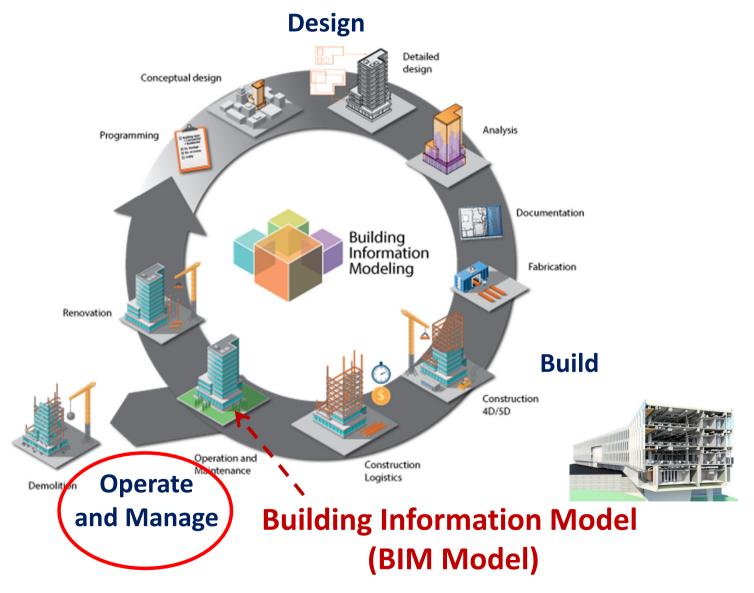


Backbone of Smart Cities = GIS + Sensors + ICT + WiFi

City is 3D – GIS should be 3D



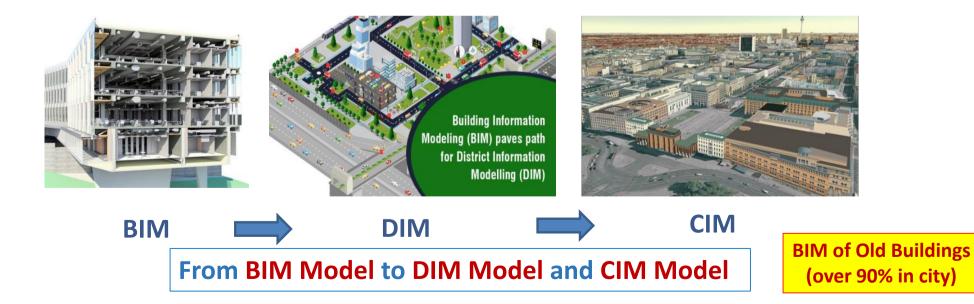
Building Information Modeling (BIM)

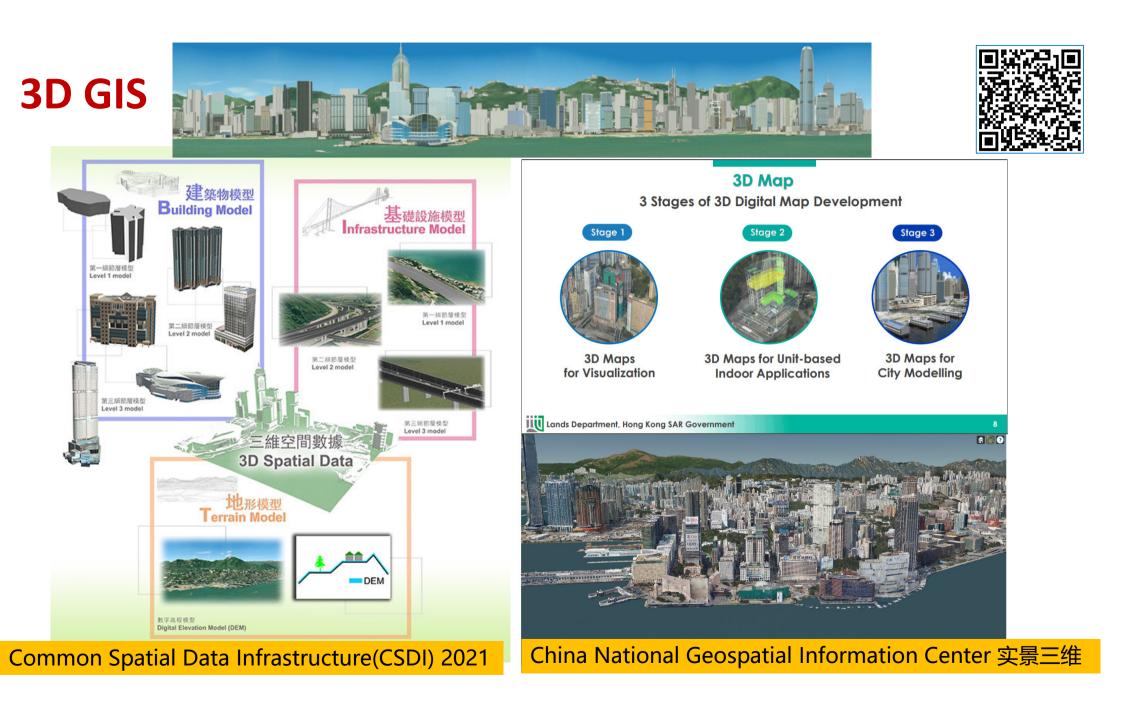


BIM technology is evolving from **District Information Modeling (DIM)** to **City Information Modeling (CIM).** DIM and CIM are similar to how we modeling buildings and infrastructure.

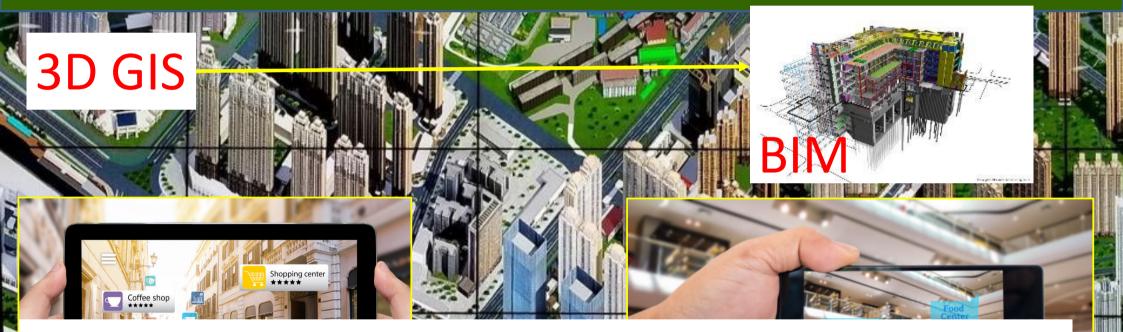
DIM is the 3D modeling at the **District level**; while **CIM** aims to model the smart city at the **City level**.

A **CIM** model could enable city-wide simulation (for architects and planners) of various aspects such as traffic, congestion, energy, impact of natural disasters such as earthquakes or hurricanes, flood control, etc.



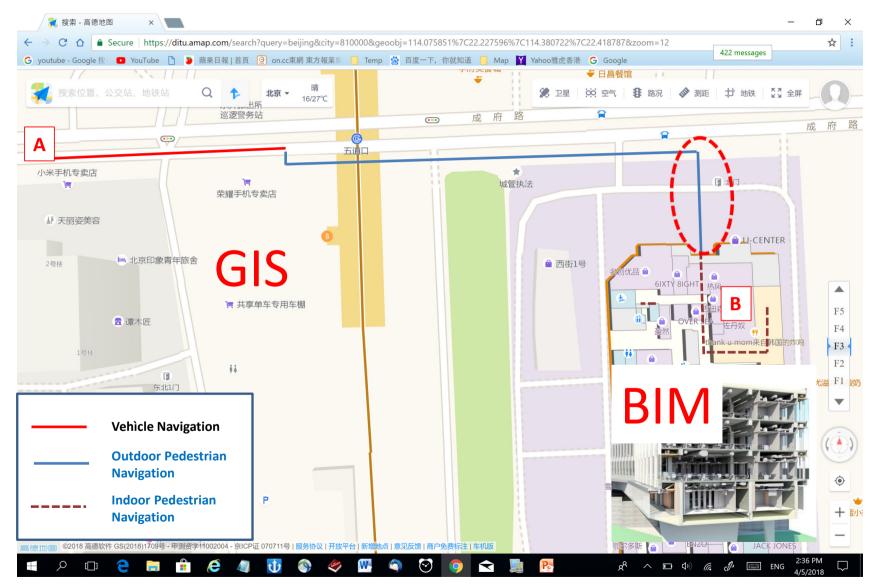


Integration of 3D GIS and BIM (Virtual Reality & Augmented Reality)



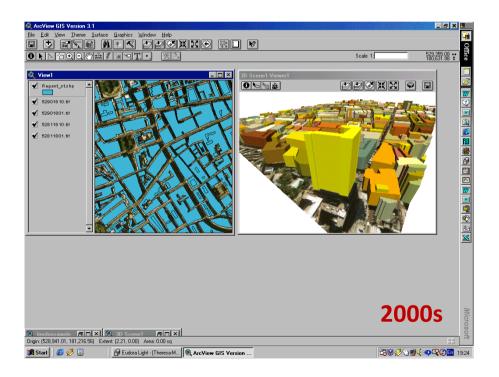
The development of 3D-GIS is more complicated than 2D-GIS The work has just started, we still have a long way to go

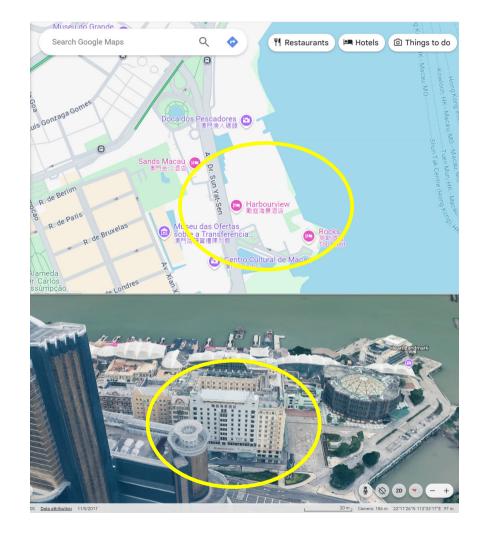
Integration of 2D GIS and BIM in Pedestrian Navigation



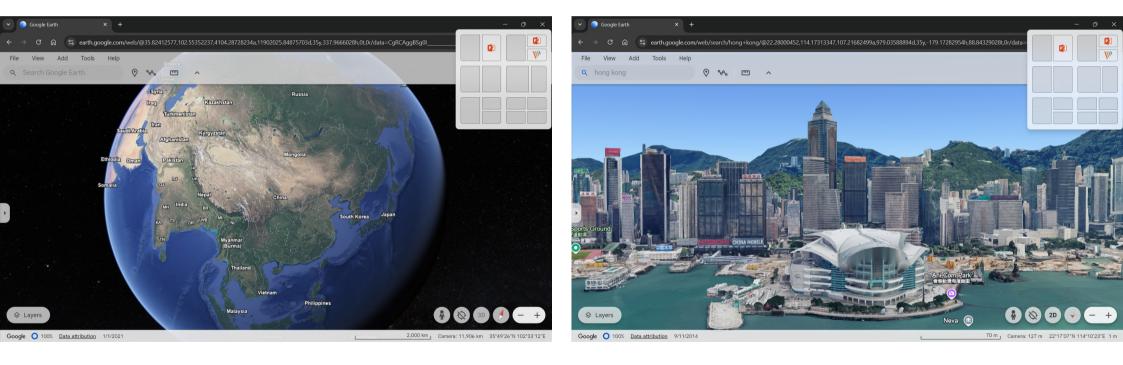
Applications of 3D GIS in Smart City Yesterday's Technology, Today's Applications

• 3D Map and Visualization

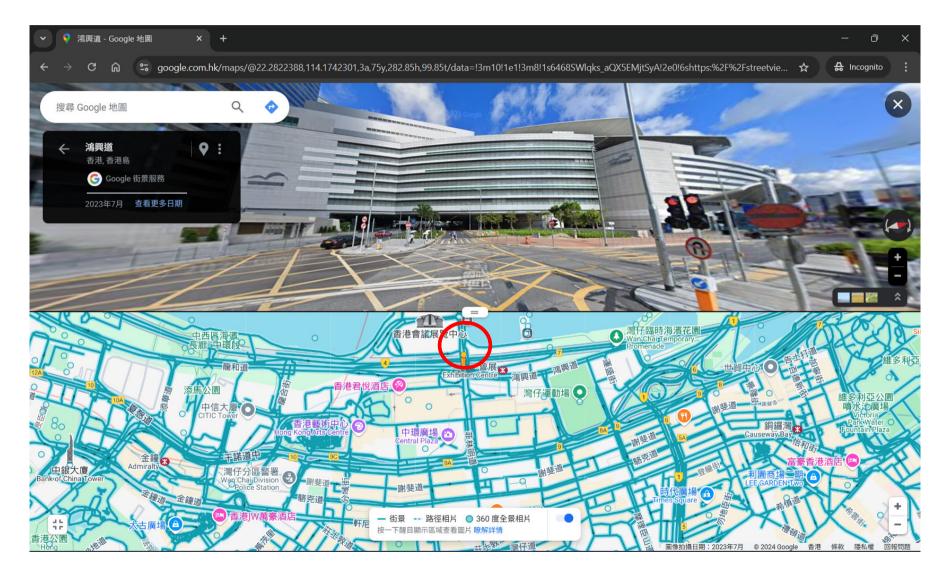




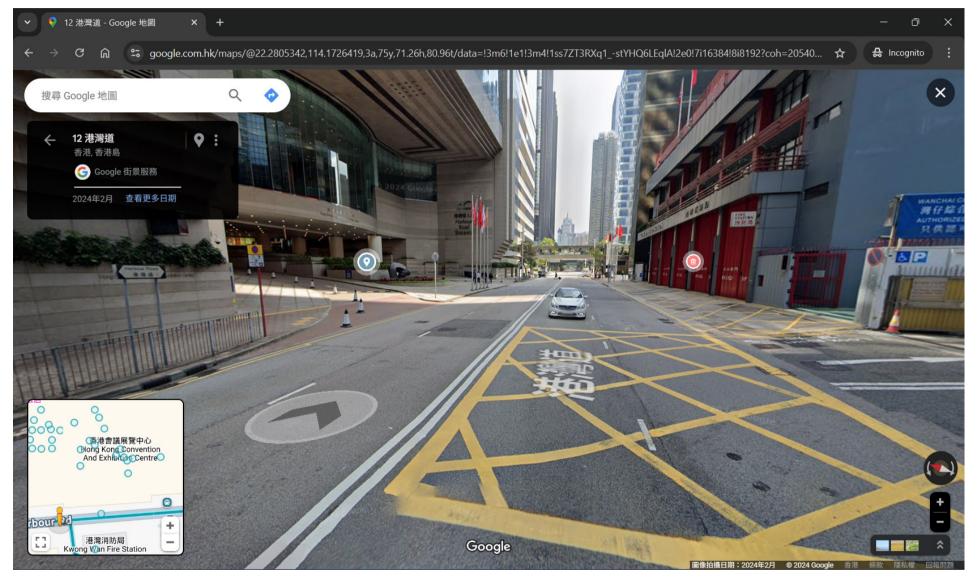
Google Earth 2005



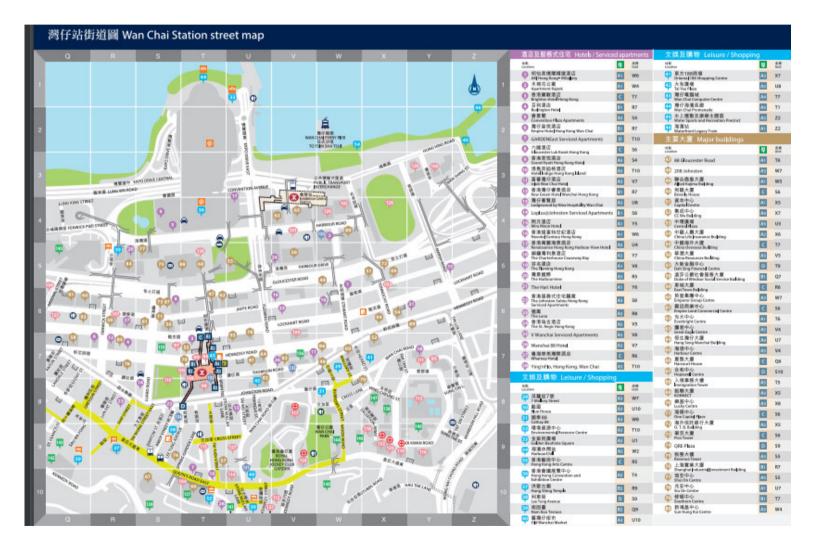
Google Street View 2007



Visual 3D Street View Navigation



MTR Location Map (2D)

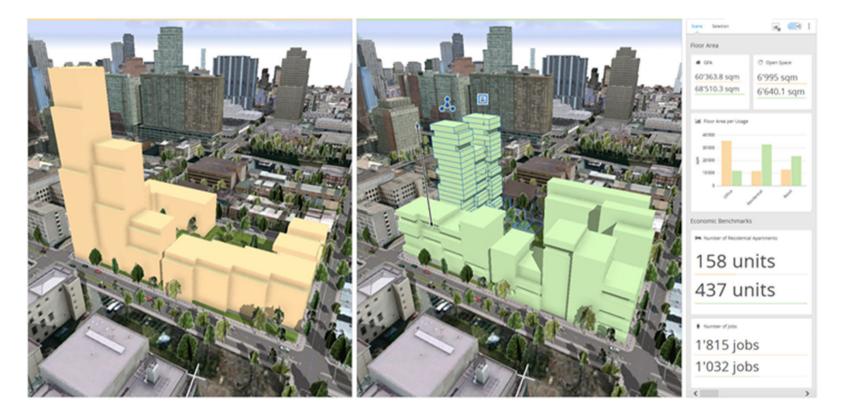


Dongguan MTR Location Map (2023 – 2.5D)



Urban Design and Planning

BIM-based Analysis, Visualization, Planning and Management:



https://www.shelidon.it/?p=8298



Before Redevelopment



Old Buildings Demolished



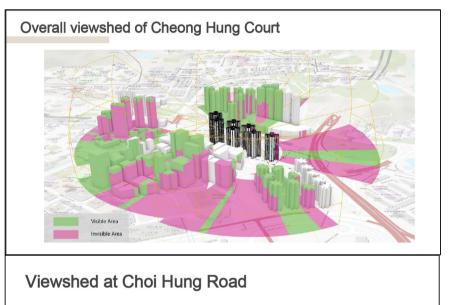
Proposed Development

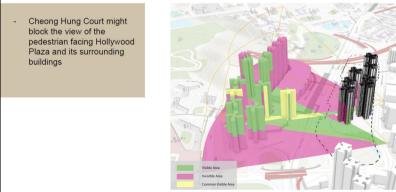


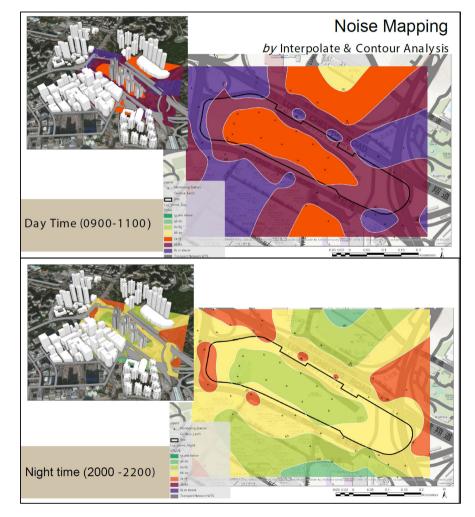
Urban Design and Planning

Viewshed

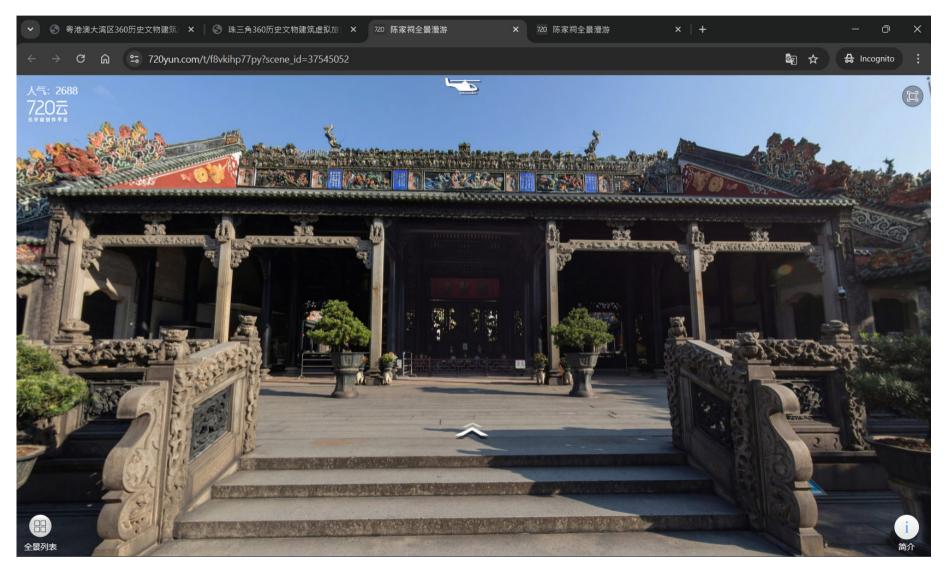
Noise







Virtual Tour of Chen Clan Ancestral Hall, Guangzhou



3D Photorealistic Multimedia Games





3D-GIS Resolution







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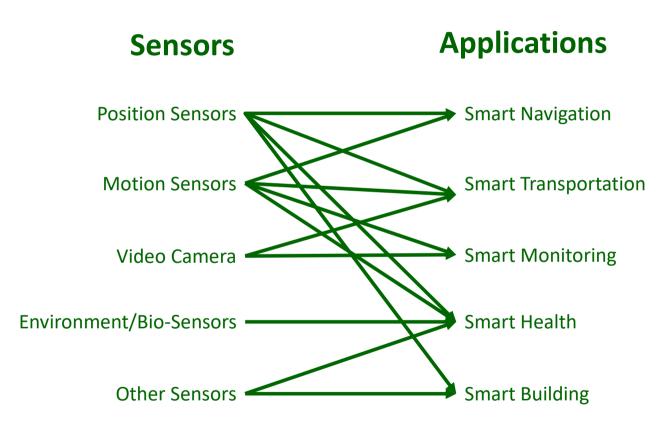






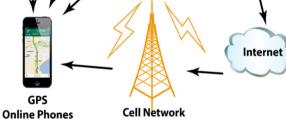
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Sensors, Information and Related IT and Applications is the Heart of Smart Cities



Sensors in Smart Mobility







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11 min (8.7 km)

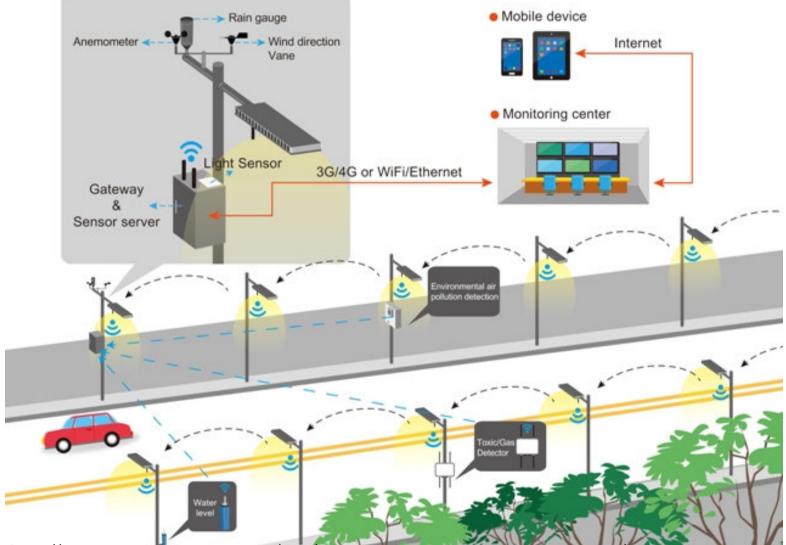
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astest mute now due to traffic conditions

Ο

Global Navigation Satellite Systems (GNSS)

Smart Street Lamp Posts



Source: https://www.environment-monitor.com/item/5g-acceleration-smart-street-lights-promising/

Car Camera

1111 Car Camera. Your Reliable Witness.

Sensors in Smart Mobility

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Global Positioning Systems (GPS): Locate the vehicle by using satellites to triangulate its position. Although GPS has improved since the 2000s, it is only accurate within several meters.

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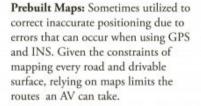
Ultrasonic sensors: Provide short distance data that are typically used in parking assistance systems and backup warning systems. sensor that uses light beams to determine the distance between obstacles and the sensor.

Light Detection and Ranging (LIDAR): A 360-degree

0

Cameras: Frequently used inexpensive technology, however, complex algorithms are necessary to interpret the image data collected.

Radio Detection and Ranging (RADAR): A sensor that uses radio waves to determine the distance between obstacles and the sensor.



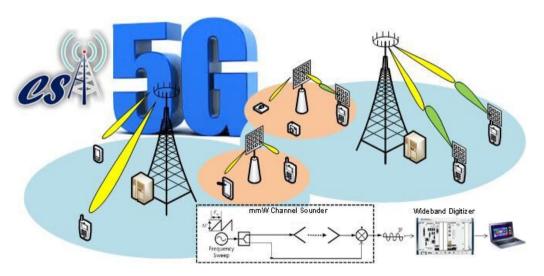
Dedicated Short-Range Communication (DSRC): Used in Vehicle to Vehicle (V2V) and Vehicle to Infrastructure (V2I) systems to send and receive critical data such as road conditions, congestion, crashes, and possible rerouting. DSRC enables platooning, a train of vehicles that collectively travel together.

Inertial Navigation Systems (INS): Typically used in combination with GPS to improve accuracy. INS uses gyroscopes and accelerometers to determine vehicle position, orientation, and velocity. Infrared Sensors: Allow for the detection of lane markings, pedestrians, and bicycles that are hard for other sensors to detect in low lighting and certain environmental conditions.

Autonomous Vehicle

5G Positioning

- 5G Cellular Grid + Mobile Phone
- Density of Base Station
- Accuracy : cm to m





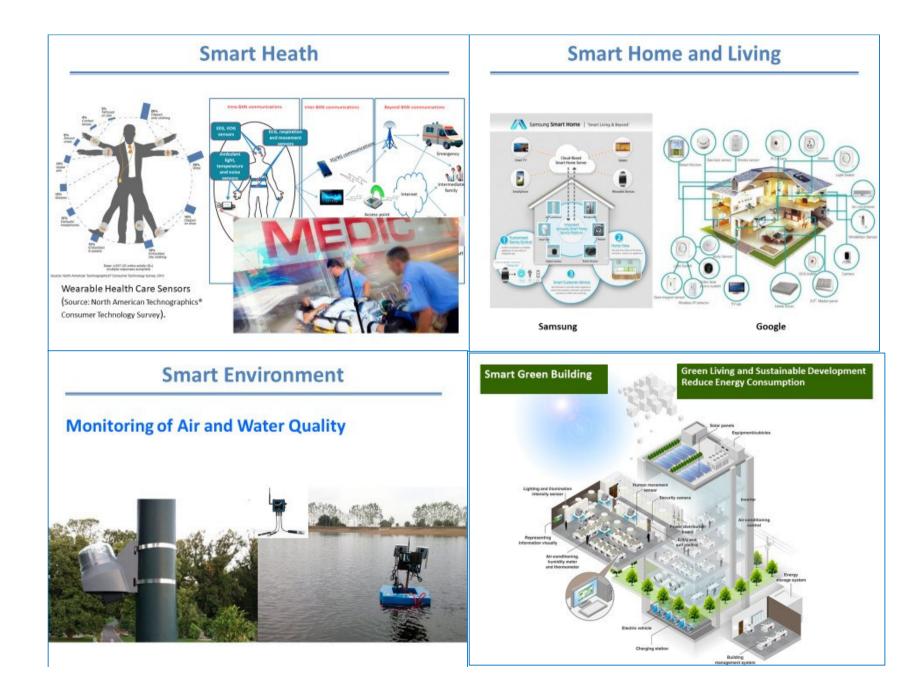
Source: ieeemy.org

Satellite



UAV Unmanned Aerial Vehicle (Drone)





Mobile Phone Data

Mobile data have been widely applied in human mobility pattern and behavioral research.

Smart Card Data/Smart Pay

Smart card data/smart pay have been used to identify commuting pattern. This data can also be used to study city spatial structure.



Smart Business

E-Pay

- Cashless
- QR Code
- Face Recognition



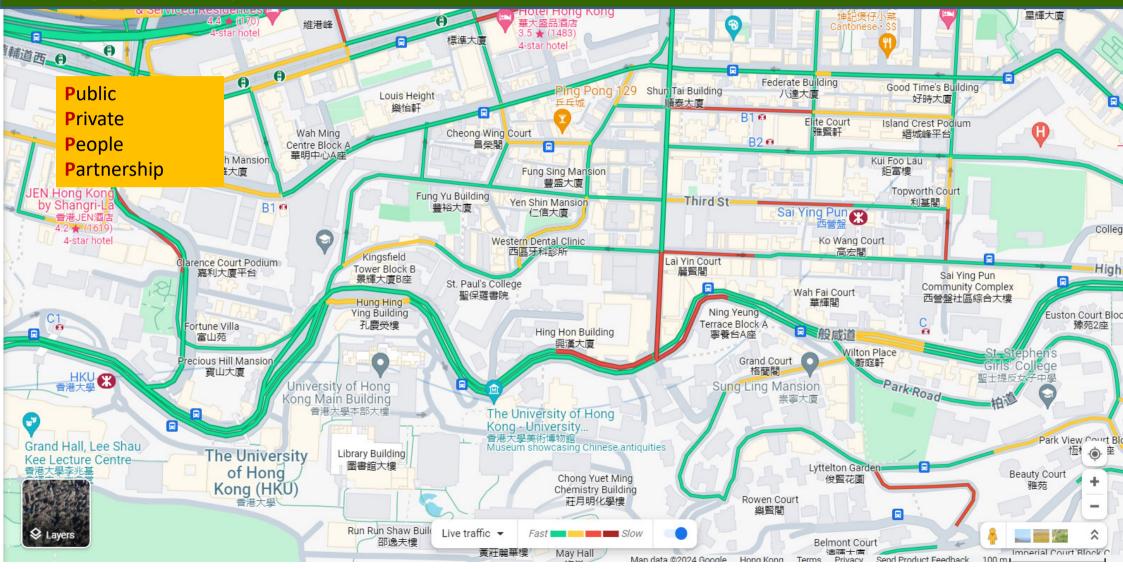
E-Shopping

- E-shopping
- Unmanned Shops : Amazon Go、 BingoBox、 Taobao



User Supplied (Crowd Sourcing) Traffic Information

(Google/Gaode Map)



Smart Government – Crowd Sourcing (Participation in Urban Management)

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- **Report Location:** xxDistrict, yy Street, No. zz
- **Report Problem:** rubbish in public area
- Work Done: Urban Service Department has cleared the rubbish on DD/MM/YEAR at xx time.



Before Reporting



After Reporting

Traditional Big Data Sources

- **Remote Sensing** land use, land use changes
- Census Data social area analysis, residential mobility, target marketing
- Telephone Directories location and spatial distribution, firm births and closures
- Credit Card Data consumer behaviour, target marketing, consumer profiling, human mobility

Smart Cities – New Data Sources

New Big Data

- **Spatial-Temporal Data**: GPS trajectory, mobile phone data, Smart Card etc.
- Smart Grid and Sensor Data: car sensor, oil pipes, Internet of Things (IoT)
- Social Media Data: Facebook, WeChat, Twitter, Instagram, etc.
- Web Data: page views, searches, purchasing, etc.
- Text Data: email, news, etc.

Urban Informatics \rightarrow **Urban Analytics**

The Urban Book Series

Wenzhong Shi • Michael F. Goodchild • Michael Batty • Mei-Po Kwan • Anshu Zhang *Editors*

Urban Informatics



OPEN ACCESS



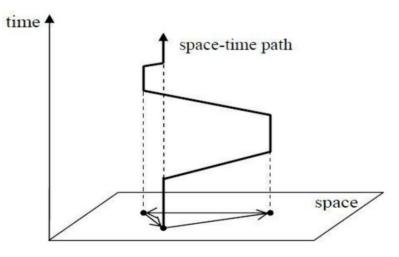
https://link.springer.com/book/10.1007/978-981-15-8983-6



Spatial-Temporal Big Data

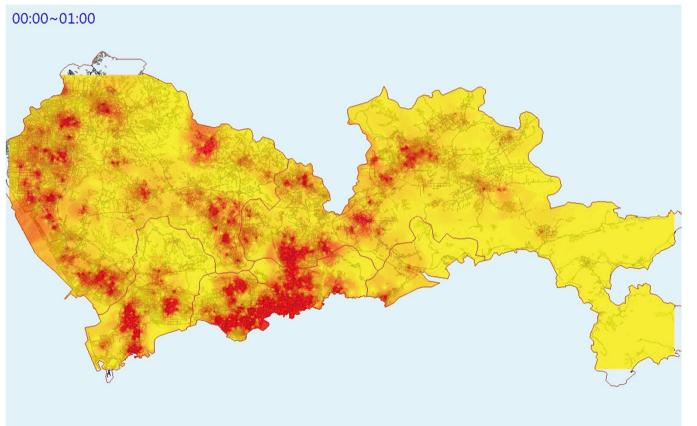
- Higher Resolution (Spatial)
- Higher Granularity (Time)
- Real Time

Space-Time Prism



Swedish Torsten Hagerstrand's Time Geography (1970)

Mobile Phone Location Data – Big Data



Activity intensity in Shenzhen within 24 hours - 2011

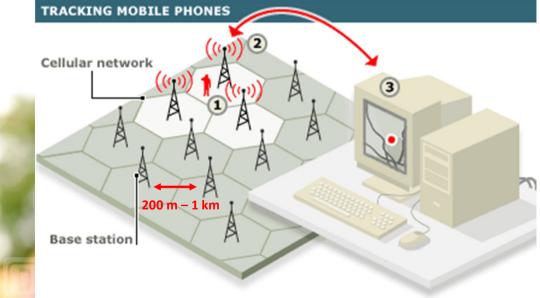
Mobile Phone Data

Mobile phone data have been widely applied in human mobility pattern and behavioral research.

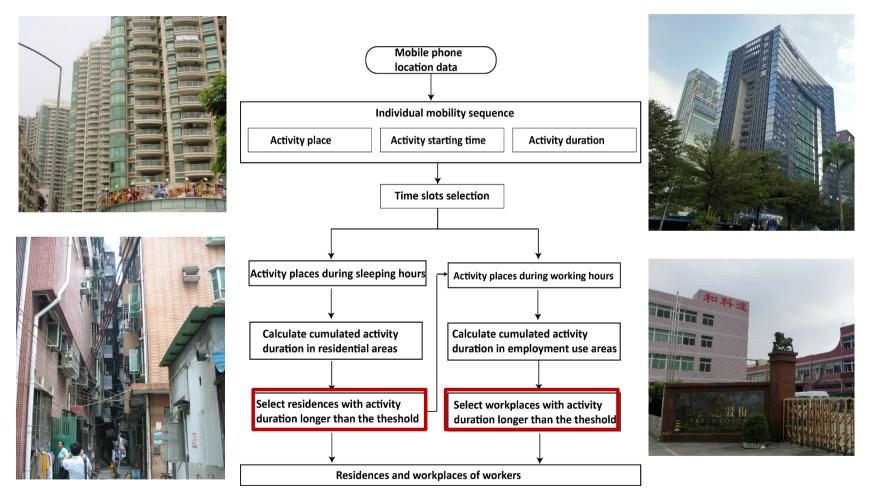
Mobile Phone User ID	Date	Time	Latitude	Longitude
А	2011/8/1	20:34:33	22.53133	114.0658
В	2011/8/1	20:34:59	22.53133	114.0658
С	2011/8/1	20:46:16	22.53133	114.0658
D	2011/8/1	14:21:38	22.56144	114.0902
E	2011/8/1	14:21:30	22.56144	114.0902



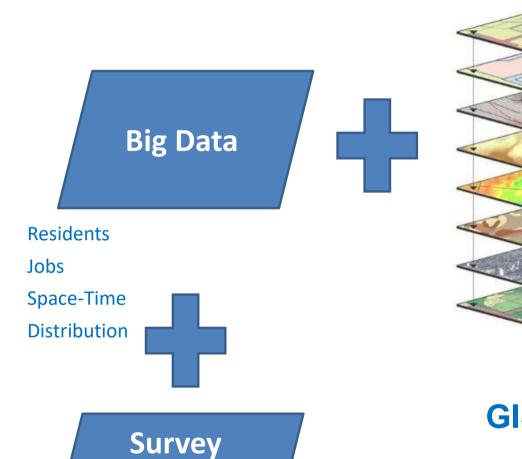
Mobile Phone Tracking



Big Data processing with GIS

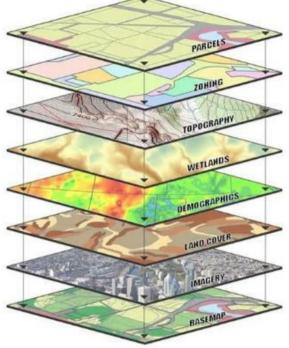


Identification of mobile phone users' residences and workplaces



Small Data

Field Work



Adminstrative Boubdaries Census and Economic Data POI Services and Facilities Land Use Transport Network Public Space etc.

GIS - Geographical Context





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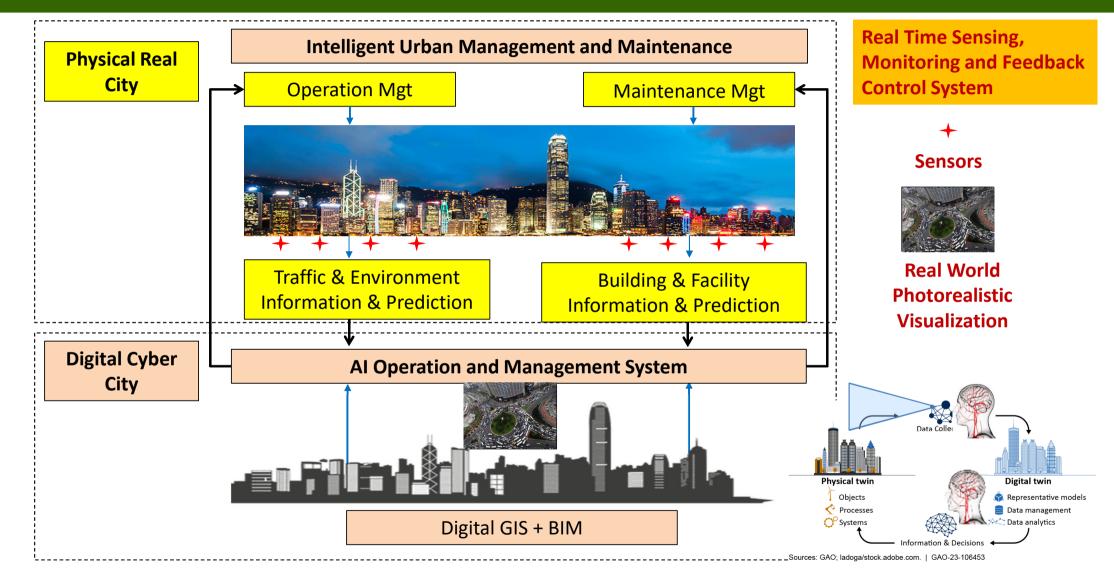


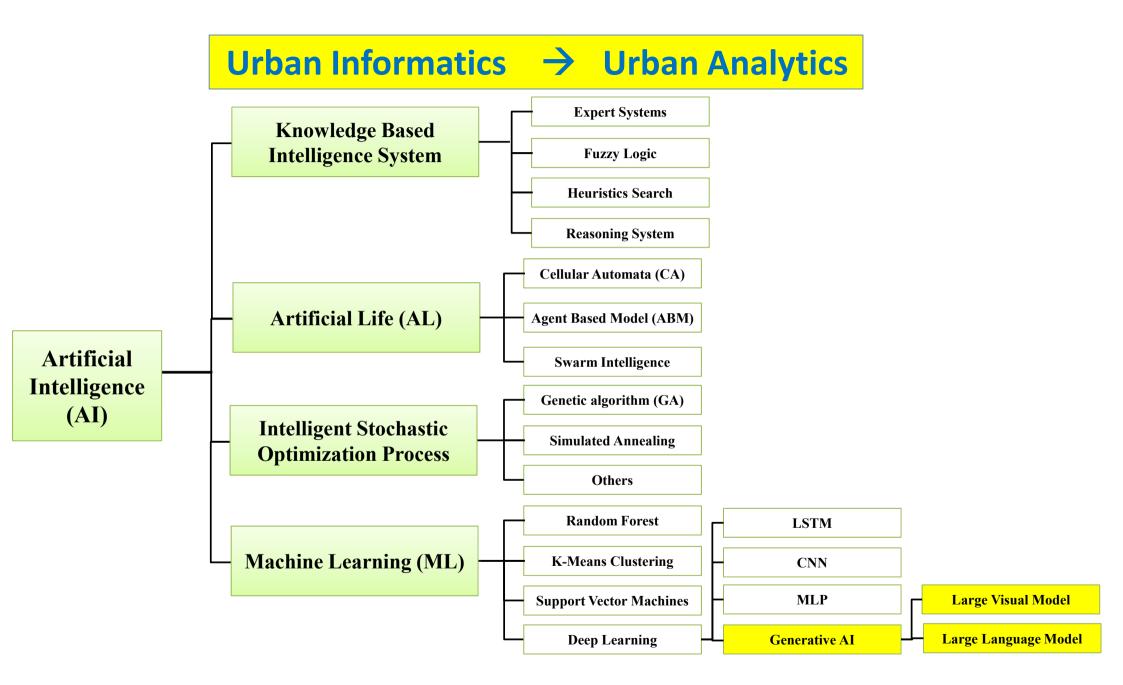


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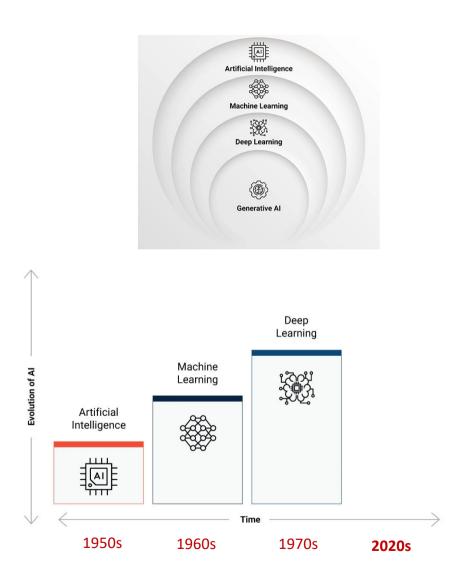
Digital Twin City





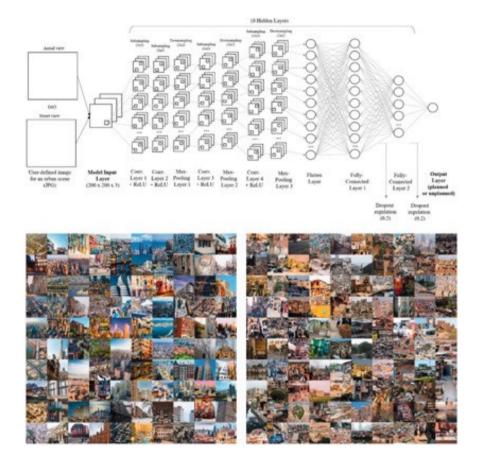
AI Definition

- Artificial Intelligence originated around **1950s**. It represents simulate intelligence in machines. It is a subset of data science. Its aim is to build machines which are capable of thinking like humans.
- Machine Learning originated around 1960s. It is the practice of getting machines to make decisions without machines to make decisions without being programmed. It is a subset of AI & Data Science. Its aim is to make machines learn through data so that they can solve problems.
- Deep Learning originated around 1970s. It is the process of using artificial neural networks to solve complex problems. It is a subset of Machine Learning, AI & Data Science. Its aim is to build neural networks that automatically discover patterns for feature detection.
- Generative AI surged around 2020s. It is advanced from using the Transformerbased deep neural networks. It is a subset of Deep Learning. Its aim is to generate different types of content – such as text, imagery, audio, video – based on what has learnt from existing content.



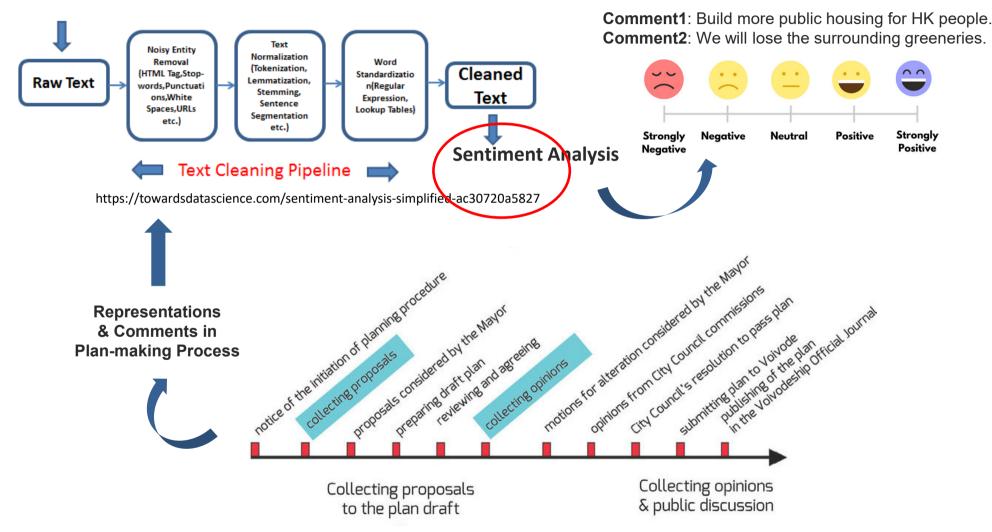
Mapping Slums and the Dynamics of the Deterioration in Cities

- With the growth of the fields of deep learning and computer vision, understanding cities through the eyes of a computer opens the door for analysing missing attributes of city dynamics.
- Large-scale analysis of digital images and patterns of captured features that may not be recognised of significance by human eyes can potentially enable various urban issues to be identified and collected.
- AI-based tool of deep convolutional neural networks could extract rich geospatial data such as slums, transport modes, and pedestrians in cities from street view images.



Ibrahim, M. R., Haworth, J., & Cheng, T. (2021). URBAN-i: From urban scenes to mapping slums, transport modes, and pedestrians in cities using deep learning and computer vision. *Environment and Planning B: Urban Analytics and City Science*, *48*(1), 76-93.

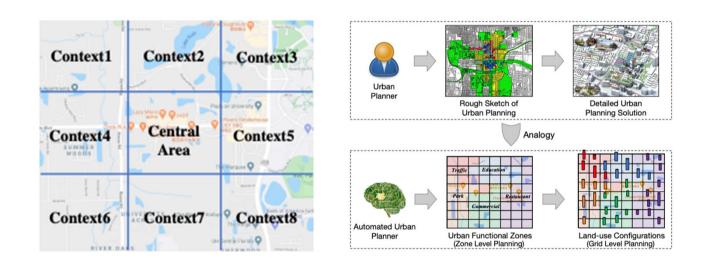
Generative AI – Large Language Model *Natural Language Processing for Public Participation*



Haklay, M., Jankowski, P., & Zwoliński, Z. (2018). Selected modern methods and tools for public participation in urban planning–a review. Quaestiones Geographicae, 37(3).

Generative AI – Large Visual Model 恢复草图 **Urban Design Rendering** 生成器 Recovered Sketch Generator (F) 渲染结果 原始草图 Original Sketch MasterplanGAN 生成器 Generator (G) Transferred Result 鉴别器 鉴别器 Discriminator (D_x) Discriminator (D,) Is it a 目标结果 Target master plan Is it a Sketch? Master Plan? i -Source: Ye, X., Du, J., & Ye, Y. (2022). MasterplanGAN: Facilitating the smart rendering of urban master plans via пг generative adversarial networks. Environment and Planning B: Urban Analytics and City Science, 49(3), 794-814.

Generative AI – Large Visual Model *Automated Urban Planning*



- The essential task of urban planning is to generate the optimal land-use configuration of a target area.
- This study proposes a landuse configuration generation framework, namely LUCGAN, which can generate a landuse configuration automatically for an empty geographical area based on surrounding contexts.

Wang, D., Fu, Y., Wang, P., Huang, B., & Lu, C. T. (2020, November). Reimagining city configuration: Automated urban planning via adversarial learning. In *Proceedings of the 28th international conference on advances in geographic information systems* (pp. 497-506).



Smart Technology, Big Data, AI and Urban Analytics

- Make cities more efficient and sustainable A Very Valuable Tool
 - New Smart Technology Better real time survey, information, and monitoring
 - **Crowd Sourcing** Public participation in sustainable urban management
 - Digital Twin with AI More efficient and timely urban management
- Changes in Surveying Education *have to learn more skills and technologies*

Thank You

• What new knowledge CAN we learn from Big Data, AI and Urban Analytics ?

