

# **Urban Atlas - Building Block Height Model 2012 - Product User Manual (PUM)**

Copernicus Land Monitoring Service



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**Abbreviations & acronyms**

AoI	Area of Interest
BBHM	Building Block height Model
BIE	Bisector Elevation Angle
CLMS	Copernicus Land Monitoring Service
CNN	Convolutional Neural Networks
CZ	Coastal Zones
DHM	Digital Height Model
DSM	Digital Surface Model



DTM	Digital Terrain Model
EEA	European Environment Agency
EEA38	EU27, UK, EFTA, West Balkan, and Turkey
EFTA	Iceland, Liechtenstein, Norway and Switzerland
EPSG	European Petroleum Survey Group Geodesy
ESA	European Space Agency
EU27	European Union Countries
FUA	Functional Urban Area
GDAL	Geospatial Data Abstraction Library
GEOSS	Global Earth Observation System of Systems
INSPIRE	Infrastructure for Spatial Information in Europe
JRC	European Commission DG Joint Research Centre
LC/LU	Land Cover / Land Use
LiDAR	Light Detection and Ranging, or Laser Imaging Detection and Ranging
LZW	Lempel-Ziv-Welch (universal lossless data compression algorithm)
N2K	Natura 2000
RPC	Rational Polynomial Correspondence
RZ	Riparian Zones
SEIS	Shared Environmental Information System
UA	Urban Atlas
VHR	Very High-Resolution

## 1. Executive summary

Copernicus is the European Union's Earth Observation (EO) Programme. It offers information services based on satellite earth observation and in situ (non-space) data and is an integrated part of EEA's strategy to improve environmental information. These information services are **freely** and **openly** accessible to its users through six thematic Copernicus services (atmosphere monitoring, marine environment monitoring, land monitoring, climate change, emergency management and security).

The Copernicus Land Monitoring Service (CLMS) provides geographical information on land cover and its changes, land use, vegetation state, water cycle and earth surface energy variables to a broad range of users in Europe and across the world in the field of environmental terrestrial applications. CLMS information is based on space data combined with other sources. It addresses a wide range of policies such as environment, agriculture, regional development, transport and energy at EU level, and European commitments to International Conventions.

CLMS is jointly implemented by the European Environment Agency (EEA) and the European Commission DG Joint Research Centre (JRC).

The local component is coordinated by the EEA as part of CLMS and aims to provide specific and more detailed information that is complementary to the information obtained through the pan-European component. The local component focuses on



different “hotspots”, i.e., areas that are prone to specific environmental challenges and problems. It is based on very high-resolution (VHR) imagery in combination with other available datasets (high and medium resolution images) over the pan-European area.

The **Urban Atlas (UA) suite of products of CLMS** was the first in a series of land monitoring services on so called hotspots addressing urban areas. It was the first service to create harmonised Land Cover and Land Use (LC/LU) maps over several hundred cities and their surroundings, i.e., within Functional Urban Areas (FUA) together with an additional layer mapping street trees. The Urban Atlas goes hand in hand with the Urban Audit, in which the European Commission’s Directorate-General Eurostat collects a wide range of social and economic indicators. The Urban Atlas adds a spatial component to the statistical data, which enables comparison of urban spatial patterns across Europe.

The Building Block Height Model (BBHM) is an additional information layer available in the UA suite of products and will be referred to UA BBHM from now onwards. Building heights within cities are important hot spots as the information can be used to make simple 3D-visualisations of buildings and structures and can be used to assist a range of analytical applications.

One of the main purposes of the UA BBHM product is to provide VHR information (building block footprint displayed as a regular grid) containing building height in selected European cities in the EEA38 area, to obtain a better insight into measuring urban density.

The first release of the UA BBHM dataset covered 31 capital cities. The second release added 7 new capital cities (from Turkey and the West Balkans) and included 6 cities already produced and published that were extended in area. This added up to a total of 38 capital cities derived from IRS-P5 stereo images of the reference year 2012.

For this new release, the UA BBHM product has been produced for 845 cities and urban centres across EEA38 which defines the project’s working area. Within the 845 cities there are 13 capital cities that were remapped due to, again, extension in area. The selection of cities to map has been based on the definition of a city and urban centre (also known as High Density Clusters), see chapter 6.1 for details. To the 845 selected cities and remapped capital areas the remaining 25 capitals cities that were not remapped were added. The UA BBHM thus covers a total of 870 cities and urban centres.

The building height information of the new release is derived from VHR false pair stereo images of, or close to, the reference year 2012, and contains only the height of the building block itself (i.e. vegetation, water and roads are masked out).

The product covers approximately 75 000 km<sup>2</sup> with a MMU of 10 x 10 m.

## 2. Guide for the reader

### 2.1 Who is this guide for?

This Product User Manual is the primary document that users are recommended to read before using the UA BBHM product. It provides an overview of the product characteristics, product methodology and workflows, user requirements and example/potential use cases, information about the Quality Assessment (QA) checks and their results as well as product technical support.

### 2.2 Content and structure

The document is structured as follows:

- Chapter 1 introduces the executive summary.
- Chapter 2 presents the reader's guide.
- Chapter 3 recalls the user requirements.
- Chapter 4 presents potential product application areas and/or use cases examples.
- Chapter 5 presents the product description (product content and features, file naming convention and format(s)).
- Chapter 6 provides a description of the product methodology and workflows and highlights the important issues to users.
- Chapter 7 summarizes the quality assessment and/or validation procedure and results
- Chapter 8 provides information on the terms of access and use of the product, as well as on the technical support for the product.
- Chapter 9 provides information on references to the cited literature
- Annexes

## 3. Review of user requirements

The European Commission has been systematically collecting user needs for Copernicus products and services, expressed in different policy areas. Information on specific aspects of land cover (i.e., building blocks) such as provided by this UA BBHM product are an essential prerequisite to support many policies, reporting and monitoring obligations (European Commission, 2019).

Building height is a characteristic that can enhance disaggregation processes of population and employment data in urban areas. Height or volume information will reduce the error margin of the disaggregation results.

3-D information for disaggregation purposes was the major user requirement that led to the production of the UA BBHM dataset.

## 4. Product application areas and/or examples of use cases

The UA BBHM can help inform urban planning efforts in terms of, for example, monitoring urban climate and potential areas at risk for urban heat island effects, which affect human wellbeing. The distribution of population in an urban area may also facilitate the understanding of the risk of spreading infectious diseases (David Frantz, 2021), the energy use and in effect CO<sub>2</sub>-emissions of different parts of the city (transport hotspots, heating/cooling), pollution dispersion and air quality, as well as guide rapid decision making in emergency response and disaster risk mitigation.

### 4.1 Use case: Building height and sustainability (energy use and CO<sub>2</sub>-emissions)

Studies show that the most efficient way to impact urban form and increase the built-up density, is by increasing the building height (Eirik Resch, 2016) (Yu, et al., 2022). The urban form and built-up density can be used as an indicator of building energy use. Increased density in urban areas can result in significant energy savings due to less energy demand for heating and cooling (Burak Güneralp, 2017). Building more compact cities has also been identified by the IPCC as an important mitigation measure for climate change, as it is related to reduced energy use per capita, primarily because of a decrease in energy used for transportation when inhabitants are able to walk or cycle to a larger extent, but also due to reduced energy use for heating and/or cooling (Burak Güneralp, 2017) (Eirik Resch, 2016) (Jiayu Li, 2022).

There is a relation between built-up density, building height and building energy use, and subsequently CO<sub>2</sub>-emissions due to for example heating, cooling, ventilation, and artificial lighting. Taller buildings with compact urban form suffer less heat loss than lower buildings in colder climates and high-rise buildings in warmer climates decrease the solar exposure on surrounding buildings, in suitably dense built-up areas, which may reduce the cost of cooling. (Eirik Resch, 2016) (Burak Güneralp, 2017)

### 4.2 Use case: Informing sustainable and climate resilient cities

Built-up area and population density can be used for creating indicators to measure progress towards different international framework agreements and governance processes, aiming to build more resilient and sustainable societies (D. Ehrlich, 2018). These indicators could for example be measuring urbanisation, land consumption, and percentage of population residing in urban areas. Population change over time could be used to model migration trends. Data on built-up area and population density can also give an indication of sustainability in the form of



estimated amount of waste created and environmental ability to manage it, as well as the access to and availability of resources at the local level. (D. Ehrlich, 2018)

To facilitate an understanding for how to build a resilient response to the impacts caused by climate change on communities, as well as how to reassess the exposure and resulting risk of people to hazards in a changing urban landscape, built-up area and population density can provide useful tools and input data. (D. Ehrlich, 2018)

## 4.3 Use case: Informing urban planning and risks to human wellbeing

Urbanisation and dense urban living could impact spreading of infectious diseases (Wu, 2017), building height could prove an indicator of this risk to human wellbeing, supporting plans and policies to mitigate such hazards.

The urban form influences the urban environment and socioeconomic activities which in turn have various effects on the urban environment in terms of for example changing land cover and urban functions and structures, which can impact for example human wellbeing. The urban form can also affect urban heat islands (UHI), greenhouse gas emissions, energy consumption, urban ecosystems, and public health. (Yu, et al., 2022) (Tian, 2019)

Building height information in cities influence metrics such as the arise of UHI, and pedestrian ventilation and pollution dispersion. Knowledge about the building height and 3D-model of cities could prove a powerful tool to inform urban planning efforts aiming to improve human wellbeing, as well as inform studies on population distribution and CO<sub>2</sub>-emissions. (Yu, et al., 2022)

Indoor environment can also be impacted by surrounding building height, causing acoustic and illumination effects that impact the health of building occupants, thus building height (and density) information can aid in neighborhood and building design to alleviate health problems caused by acoustic or visual discomfort. (Isabelle Y.S. Chan, 2018)

Building height has been shown to affect Land Surface Temperature (LST) and air temperature ( $T_{air}$ ) and varying building heights can contribute to better ventilation and pollution dispersion in densely built areas in cities (Bálint Papp, 2021) (Zander S. Venter, 2020). Both are indicators of urban heat and can indicate the cooling requirements of urban infrastructure, and thus the energy-use and carbon emissions (Zander S. Venter, 2020), as presented in section 4.1.

LST is important for quantifying urban heat islands (UHI) (Zander S. Venter, 2020), and has been shown having a lower correlation (than building cover and vegetation cover and height) contributing to a higher LST in relatively low buildings (up to 9 m) and cooler climates (Alexander, 2021), but having a stronger correlation (than building density and vegetation coverage) and contributing to a lower LST in high rise buildings and warmer climates (Zhong Zheng, 2019).

$T_{air}$  is largely influenced by urbanisation (Zander S. Venter, 2020) and LST (Alexander, 2021), and can be predicted with building height. Both LST and  $T_{air}$  can be used as an indicator for human heat exposure and UHI, but  $T_{air}$  may be the better



indicator for this specific use and can additionally be used as a determinant of urban cooling, (Zander S. Venter, 2020). This information can be used to establish policy instruments and/or national plans to prepare for and adapt to extreme heat waves and the health risks and impact this may have on the urban population. The information on building height and the relation to  $T_{air}$  can also be used for climate change adaptation measures in cities, to quantify the value of eco-systems services and inform nature-based solutions.

Air temperature,  $T_{air}$ , is linearly connected with varying building height, and building height is one variable to predict  $T_{air}$ , and can significantly reduce city ventilation and thus can indirectly affect pollution dispersion (Zander S. Venter, 2020), (Bálint Papp, 2021). High-rise buildings tend to decrease natural ventilation (Burak Güneralp, 2017). Building height can also be used as an indicator of populations density and quantification of emissions (D. Ehrlich, 2018). The temperature of block surfaces, affected by building heights and density mainly when these factors combined contribute to a decreased solar radiation reaching the block surface, also affect the temperature experienced by pedestrians, wind patterns and heat loss (Yunhao Chen, 2020).

These indicators and information indicate the importance of open-source building height modelling for input into urban planning for human health and wellbeing and to facilitate nature-based solutions to climate change adaptation.

## 4.4 Use case: Modeling hazard impact and emergency response

Population density and built-up area can be used to model impacts from hazards, as population location is an important indicator of exposure and vulnerability in disaster risk analysis (Xu M, 2020). The measured impact of fast-onset hazards on the built environment is largely depending on the built-up area as the main indicator of impact is measured in casualties (D. Ehrlich, 2018). The more precise the measurement of disaster impact is, the better the estimations of aid and emergency services can be.

Emergency response to man-made, meteorological or geophysical hazards leading to unfolding disasters or humanitarian emergencies could also largely benefit from population density data. The information can aid in assessment of damage severity and number of people affected, for example with the Copernicus Emergency Management Service (Copernicus EMS) or similar services. Assessment of damage can also provide useful and updated information to resource allocation, specific and exposed assets or population groups such as transport networks and route design, utilities and industry as well as settlements. (D. Ehrlich, 2018) (Xu M, 2020)

## 4.5 Building height and building stock

Building height could prove an indicator of the building stock of a city, and thus an indicator of consumption of resources, and future source of supply of metallic and mineral resources (Kleemann, 2017).



Building stock is a variable associated to the energy performance of buildings and has led the EC to establish an [observatory](#) aiming to provide a better understanding of the energy performance of the building sector through reliable, consistent and comparable data.

Building height can provide key characteristics of the building stock in both large urban agglomerations and small-scale rural settlements as it enables previously impossible environmental, socioeconomic, and climatological studies (Thomas Esch, 2022).

This is especially true for the fields of spatial planning, sustainable development, urban climate, urban economics, or disaster and risk management. More precise data on building density is also essential for an improved modeling and assessment of population distribution, air pollution and emissions, carbon footprint, energy demand, and traffic patterns (Thomas Esch, 2022).

## 5. Product description

### 5.1 Product overview

Each Urban Atlas product is generated over the city and its commuting zone, according to the EU-OECD definition of a Functional Urban Area (FUA). The EU-OECD have developed a methodology to define high-density clusters (also called urban centres), cities and FUAs in a consistent way across countries. Using population density and travel-to-work flows as key information, a FUA consists of a densely inhabited city and a surrounding area (commuting zone) whose labour market is highly integrated with the city (OECD 2012). For the detailed definition of urban centres, cities and FUAs please refer to "The EU-OECD definition of a functional urban area" (Dijkstra, 2019).

The UA BBHM product is a set of layers in GeoTIFF raster format containing building height information for 870 European cities and urban centres, located in Functional Urban Areas (FUAs) across the EEA38 area, covering an approximate area of around 75 000 km<sup>2</sup>.

The UA BBHM is based on VHR imagery for the defined areas of interest (AoI) from the reference year 2012, or as close as possible to this year, see chapter 6.2.1 for details. From these a Digital Surface Model (DSM), a Digital Terrain Model (DTM) and Digital Height Model (DHM) are created, with which the BBHM product can finally be derived.

The UA BBHM product has a MMU of 10 x 10 m and is projected to the European reference system EPSG:3035. All information that does not correspond to a building height value is classified as NoData, with a value in the data matrix of 65535.

### 5.2 Technical specification

The table below summarises the technical specification of the UA BBHM product.



Specification	UA BBHM
Geographic Coverage	About 75 000 km <sup>2</sup> across 870 cities and urban centers in EEA38
Temporal description (VHR coverage reference dates)	2012 (primarily, depending on data availability), see chapter 6.2.1 for details
Minimum Mapping Unit (MMU)	10 x 10 m
Projection	ETRS89 Lambert Azimuthal Equal Area (LAEA) (EPSG 3035)
Format	Raster GeoTIFF 16-bit
Metadata	INSPIRE Metadata Implementing Rules: Technical Guidelines based on EN ISO 19115 and EN ISO 19119
Ancillary information layer	Pixel Based Information
Positional accuracy	Horizontal: half a pixel (5 meters) Vertical: 3 meters

The content of a downloaded package is explained below.

For each city or urban centre, the downloaded .zip-file contains four folders, see Figure 5-1:

- Dataset folder containingin the Digital Height Model in GeoTIFF format;
- Metadata folder containing an INSPIRE Compliant XML-file;
- Pixel Based Info folder containing the pixel based information shapefile containing the satellite / LiDAR footprints;
- QC folder containing the Quality Control (QC) report in PDF format.

#### NAMING CONVENTION:

| COUNTRY CODE NUMERIC CITY IDENTIFIER | CITY NAME | PRODUCT NAME | LAYER

The nomenclature of each layer consists of four parts, written in capital letters and separated by an underscore. The first part is an alphanumeric code referring to the country in which it is located, followed by a unique numeric identifier for each city, the second part is the name of the city, the third part is a suffix indicating the name of the product suite and the fourth and last part indicates the layer. An example of nomenclature for the GeoTIFF file layer is as follows: ES009\_VALLADOLID\_UA2012\_DHM (see Table 5-1 for more examples).

Figure 5-1 shows how the product is structured, as well as additional nomenclature example.

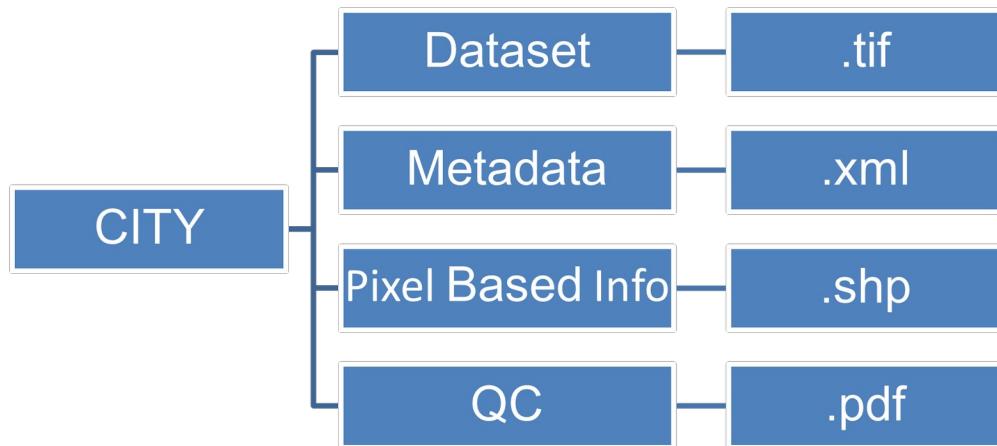


Figure 5-1 Structure of the product folders

**Table 5-1 Structure of the BBHM product**

	Type
CITY	Zipfile
Dataset	FOLDER
<i>Example: ES009_VALLADOLID_UA2012_DHM_v010.tif</i>	File
Metadata	FOLDER
<i>Example: ES009_VALLADOLID_UA2012_DHM_v010.xml</i>	File
Pixel Based Info	FOLDER
<i>Example: ES009_VALLADOLID_UA2012_DHM_v010.shp</i>	File
QC	FOLDER
<i>Example: ES009_VALLADOLID_UA2012_DHM_v010.pdf</i>	File

The following subchapters give a description of the content of each of the zipped folders.

### 5.2.1 Previous vs. current release folder naming

In previous releases of the Urban Atlas products, the delineations of the FUAs have been updated continuously from one release to another, as the population densities or commuting patterns of the city areas have changed. These updated FUAs have been labelled “L”+number to indicate which version of FUA delineation was used (e.g. L1, L2, etc.). In the previous release of the UA BBHM this labelling was kept, but for this new UA BBHM release it has been removed. Thus, for a user who has downloaded UA BBHM files for any of the capital cities from the previous release, the files are named as for the example of Vienna (WIEN) in Figure 5-2. As can be seen from the example, this was the second FUA version of Vienna, indicated by the “L2”. For this new release the downloaded folder names will not have this LX, where X indicates a number, see Figure 5-3.

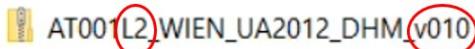


Figure 5-2 Example of naming of previous releases’ .zip-file

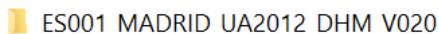


Figure 5-3 Example of naming of current releases

Regarding versioning there are a few aspects to take into consideration: In the first release there were two versions: version “v010” (original version) and version “v020” if the city had been re-mapped due to some area extension, see Figure 5-4. There were six cities with version “v020”. In this new release these cities will display version “v030”.

Show 20 entries		Search:		
<input type="checkbox"/> Name	Country	Type	Version	Size
<input type="checkbox"/> DE001L1_BERLIN	Germany	Raster	v020	3.1 MB
<input type="checkbox"/> DK001L2_KOBENHAVN	Denmark	Raster	v020	1.6 MB
<input type="checkbox"/> EL001L1_ATHINA	Greece	Raster	v020	1.9 MB
<input type="checkbox"/> ES001L2_MADRID	Spain	Raster	v020	1.8 MB
<input type="checkbox"/> NL002L2_AMSTERDAM	Netherlands	Raster	v020	1.2 MB
<input type="checkbox"/> UK001L2_LONDON	United Kingdom	Raster	v020	4.5 MB
<input type="checkbox"/> AL001L1_TIRANA	Albania	Raster	v010	1.4 MB
<input type="checkbox"/> AT001L2_WIEN	Austria	Raster	v010	2.0 MB
<input type="checkbox"/> BA001L1_SARAJEVO	Bosnia and Herzegovina	Raster	v010	1.0 MB

Figure 5-4 Example of cities with areas extension, and the subsequent version “v020”

The remaining 25 capital cities that were a part of the previous release have not been updated for building height values but were subjected to a remapping of “0”



values into “NoData”, to be aligned with this new release. In this case these cities will display version “v020”.

All newly added cities for this release will display version “v010”.

### **5.2.2 Dataset**

In this folder, a raster file in GeoTIFF format is stored. This is a layer that shows the heights of the buildings, with a data type of 16-bit unsigned integer (UInt16) with LZW compression, since a negative height value on buildings is not expected. It presents a MMU of 10 x 10 meters, and it is projected to the European terrestrial reference system “ETRS89 Lambert Azimuthal Equal Area (LAEA)” with EPSG:3035. All information that does not correspond to a building height value is classified as NoData, with a value in the data matrix of 65535. These characteristics allow the GeoTIFF file to occupy a very small size (as compared to the first release that also included “0” values). It was decided to remove the zeros because we could not distinguish a “0” that was a measured value or a “0” that referred to vegetation, water or roads (masked out).

When the user opens the GeoTIFF file in a GIS software, it will be displayed using a default grayscale ramp, as the product does not have a default color ramp or style defined.

### **5.2.3 Metadata**

A metadata file is created for each BBHM file, containing the information of the product. The information refers to data identification, classification of spatial data, geographic reference, temporal reference, quality and validity, conformity, constraints related to access and use and responsible organisation. These files were created according to the EEA metadata INSPIRE compliant requirements guideline.

### **5.2.4 Pixel Based Information**

The Pixel Based information data includes an ESRI shapefile layer, together with its associated files, showing the information and spatial distribution of the VHR satellite, or LiDAR, image pairs used for the construction of the Digital Height Model (DHM).

Each file details the images conforming each false stereo pair. The information is subdivided into polygonal entities. When one of the polygonal entities is selected the accompanying attribute table can be shown, see example in Figure 5-5. The attribute table shows the properties of the images that compose the image pair used to evaluate the building height in that specific area; like the image ID, the date of image acquisition, sensor used (vehicle, see Table 5-2), percentage of cloud cover, the value of nadir angle, sun elevation, and value of azimuthal angle. These values are essential for the calculation of the DSM using optical satellite photogrammetry technology, see chapter 6.3.

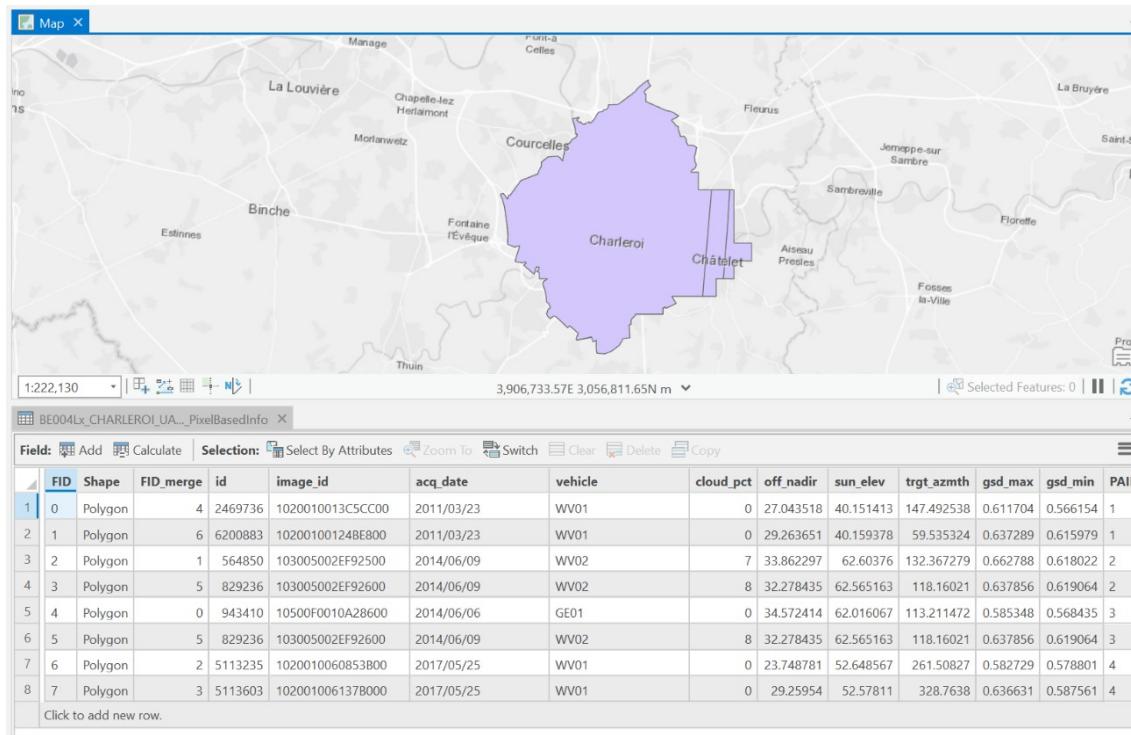


Figure 5-5 Shapefile attribute table with the main characteristics of sensor images used

The previous example is true for any of the 845 new cities added or capital cities remapped in this release, i.e. the 25 capital cities which were not remapped from the previous release will not have a folder called Pixel Based Info in the downloaded .zip-file. These 25 capital cities from the previous release have been included without remapping, however they have undergone a reprocessing as described in chapter 6.6.

Table 5-2 shows the name and description of the different sensors used.

**Table 5-2 List of sensors used**

Sensor	Description	Resolution	Spectral Bands
WV-01	Worldview 1	0.5	PAN
WV-02	Worldview 2	0.5	PAN + RGB NIR
WV-03	Worldview 3	0.5	PAN + RGB NIR
GE-01	GeoEye 01	0.5	PAN + RGB NIR
QB-02	Quickbird 02	0.65	PAN + RGB NIR
IK	Ikonos	0.5	PAN + RGB NIR

## 5.2.5 QC

The QC report is a report in PDF format with the results obtained after the execution of the quality control. This report is produced for each of the processed cities.



The first part of the report presents a summary table identifying the city and the name of the layer to which the report refers, the country to which it belongs, and the area assessed, as well as the organization that performs the functions of data provider and quality control, followed by a rating of the quality level, see Figure 5-6.

## 1. SUMMARY

<b>Country</b>	Italy
<b>City</b>	Milano
<b>Product Name</b>	IT002Lx_MILANO_UA2012_DHM
<b>Data Provider</b>	COTESA
<b>Data Source</b>	Satellite Imagery
<b>Area</b>	647.08 sq. km
<b>QC Date</b>	22/09/2021
<b>External QC Provider</b>	KPGeo
<b>QC Reviewer</b>	Kamil Grudzień

This document summarizes the results of Quality Control (QC) performed by KPGeo for the Digital Height Model (DHM) created by COTESA.

Overall product acceptance: **Yes**

Product quality level: **Very High**

Figure 5-6 First part of the QC Report with a summary that has been extracted

This is followed by format consistency table, see Figure 5-7, in which the conformity of the GeoTIFF file with the properties and technical characteristics relevant to data type, coordinate system, MMU etc., are evaluated.



## 2. FORMAT CONSISTENCY

<b>File format</b>	Raster GeoTIFF 16-bit
<b>Coordinate Reference System (CRS)</b>	ETRS 1989 LAEA (Lambert Azimuthal Equal Area) EPSG 3035
<b>Minimum Mapping Unit (MMU)</b>	10 x 10 m
<b>Mapping area</b>	Compliant
<b>Height value range</b>	3.0 – 125.0 m
<b>Metadata</b>	Compliant
<b>No Data value</b>	65535

Figure 5-7 Format consistency evaluation of the dataset

Finally, the methodology followed during the QC process is described, accompanied by an accuracy assessment table with the results obtained for each control point. The control points are distributed throughout the AoI and associated QC blocks, see Figure 5-8, and represented on an overview map.

## 5. ACCURACY ASSESSMENT AT BUILDINGS LEVEL

Point ID	WGS84 Coordinates	QC Location	DHM Height Value [m]	QC Height Value [m]	Quality Level
IT002Lx_MILANO_UA2012_DHM_001	45.674252N 8.865336E	Correct	3	3	Good
IT002Lx_MILANO_UA2012_DHM_002	45.664282N 8.878322E	Correct	5	8	Good
IT002Lx_MILANO_UA2012_DHM_003	45.649777N 8.877143E	Correct	6	9	Good
IT002Lx_MILANO_UA2012_DHM_004	45.630659N 8.871141E	Correct	5	6	Good
IT002Lx_MILANO_UA2012_DHM_005	45.605736N 8.875140E	Correct	12	13	Good
IT002Lx_MILANO_UA2012_DHM_006	45.597042N 8.883423E	Correct	24	25	Good
IT002Lx_MILANO_UA2012_DHM_007	45.653980N 8.890687E	Correct	6	7	Good
IT002Lx_MILANO_UA2012_DHM_008	45.642591N 8.903388E	Correct	3	3	Good
IT002Lx_MILANO_UA2012_DHM_009	45.619612N 8.890420E	Correct	10	12	Good
IT002Lx_MILANO_UA2012_DHM_010	45.614711N 8.904792E	Correct	22	25	Good
IT002Lx_MILANO_UA2012_DHM_011	45.585631N 8.896835E	Correct	5	8	Good
IT002Lx_MILANO_UA2012_DHM_012	45.573747N 8.907940E	Correct	6	7	Good
IT002Lx_MILANO_UA2012_DHM_013	45.648944N 8.910633E	Acceptable	3	6	Good
IT002Lx_MILANO_UA2012_DHM_014	45.617012N 8.909387E	Acceptable	7	9	Good
IT002Lx_MILANO_UA2012_DHM_015	45.603847N 8.917826E	Correct	7	6	Good
IT002Lx_MILANO_UA2012_DHM_016	45.592406N 8.922753E	Correct	5	7	Good
IT002Lx_MILANO_UA2012_DHM_017	45.565164N 8.915275E	Acceptable	3	3	Good
IT002Lx_MILANO_UA2012_DHM_018	45.550345N 8.930232E	Correct	5	8	Good
IT002Lx_MILANO_UA2012_DHM_019	45.602921N 8.943002E	Correct	8	9	Good
IT002Lx_MILANO_UA2012_DHM_020	45.581263N 8.951090E	Correct	4	3	Good
IT002Lx_MILANO_UA2012_DHM_021	45.577173N	Correct	3	5	Good

Annex 1 - Overview Map

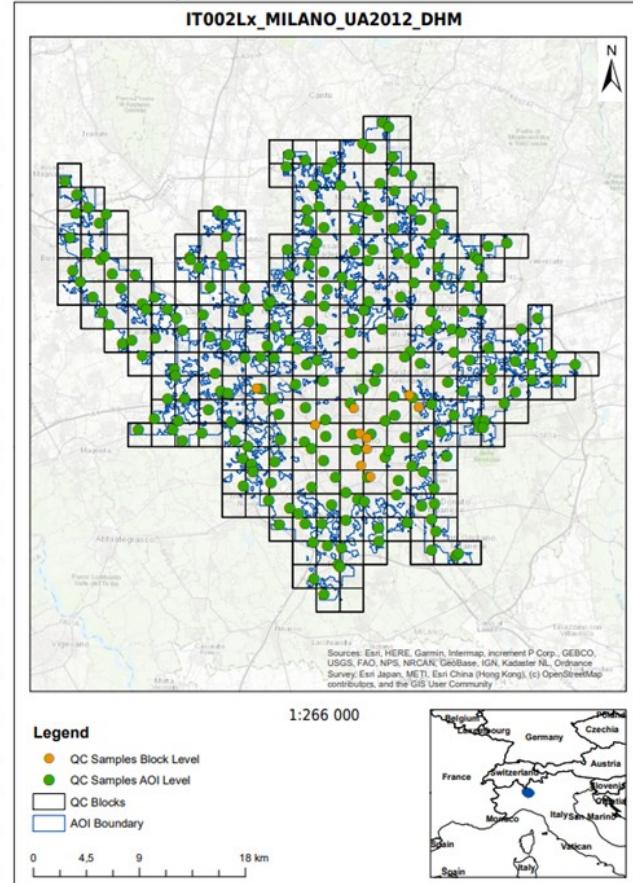


Figure 5-8 Accuracy assessment at building block level

The quality control reports contain the following information:

- File naming
- Minimum Mapping Unit value
- Reference system
- Raster data type
- Raster NoData value
- Building heights range
- DHM coverage correctness
- Building heights correctness

### 5.2.6 Symbology

Symbology files in two formats, xml and lyr are also available for download.



## 6. Product methodology and workflow

This chapter is a description of the methodological development of the UA BBHM, a brief introduction is given below.

The spatial coverage of the UA BBHM product is defined by the Area of Interest (AoI), which comprises 75.000 km<sup>2</sup>.

The proposed dataset with building height information covering European cities and urban centres is based on satellite images from sensors WV-01, WV-02, WV-03, GE-01 and IK-02, constructing false stereo pairs of images acquired as close as possible to the defined reference year, 2012. Based on these false stereo pair images, a digital surface model (DSM) was generated. Afterwards a digital terrain model (DTM) was derived from the DSM with different filter algorithms and the assistance of Urban Atlas 2012 datasets. The calculation of the normalized DSM was done by a simple subtraction of the DTM from the DSM. The final product was then clipped based on the AoI and quality controlled.

The following steps were undertaken, detailed in Figure 6-1.

- Region Studies:
  - Defining and refining of the AoI by country, city and areas with buildings
  - Selection of the appropriate image(s) to be used in the algorithm for each AoI
- Building Block Height Model (BBHM) data extraction:
  - Automatic extraction of the DSM, DTM and DHM
  - Building mask/footprint generation and building block extraction

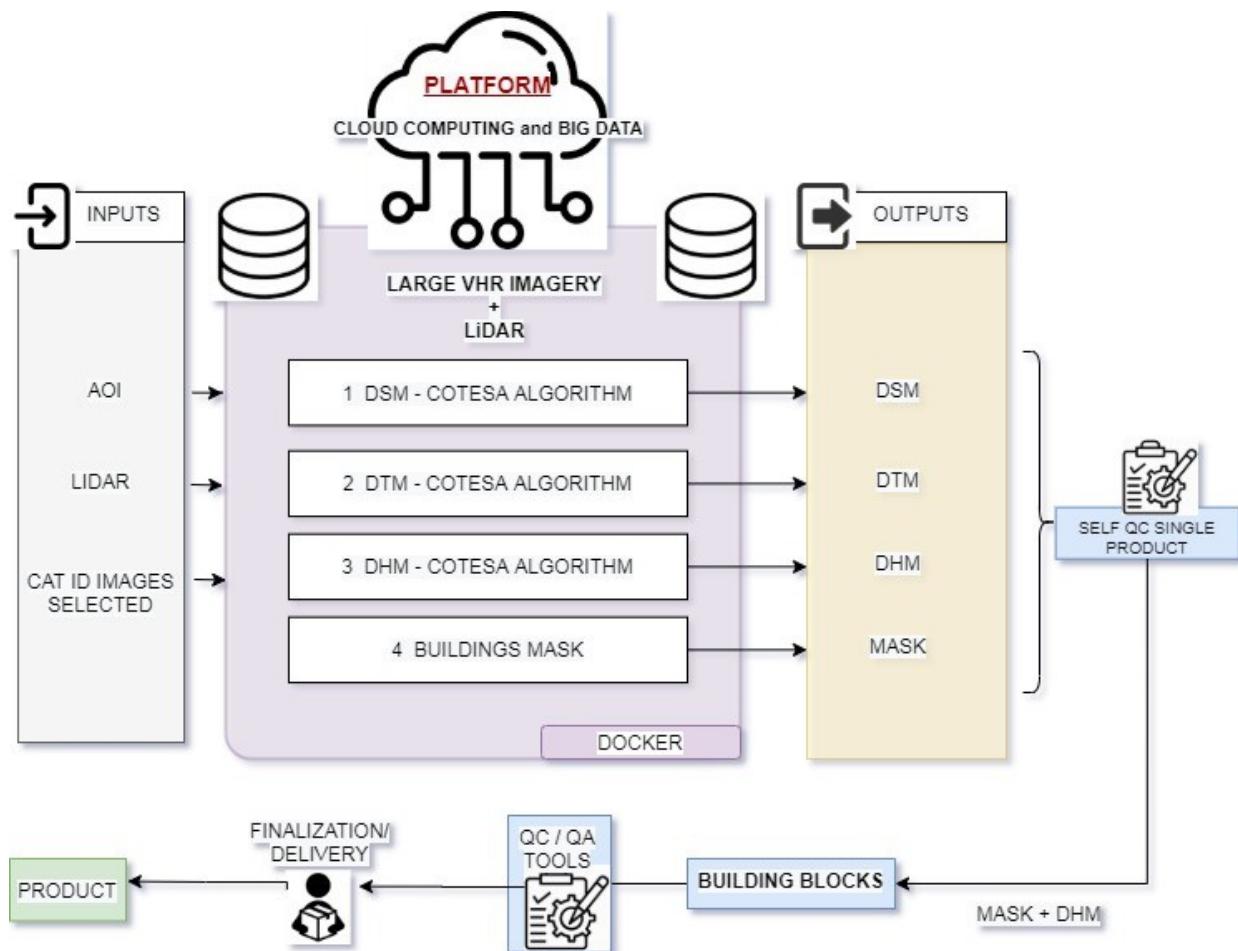


Figure 6-1 Workflow of the the creation of the BBHM

## 6.1 Defining and refining the Aols

The AoI (Area of Interest) for the BBHM represents the urban areas of each city or urban centre. It is the combination of Urban Audit cities from 2018<sup>1</sup> and urban centres from 2011<sup>2</sup> (also called high-density clusters) available at GISCO<sup>3</sup> servers.

The two files were merged and added to the high-density clusters of West Balkans and Turkey delineated in the framework of a previous project since West Balkans and Turkey are not part of GISCO. It represents approximately an area of 198.000km<sup>2</sup>. For simplicity reasons it was decided to call it the City Perimeter described in more detail in 6.1.1.

<sup>1</sup> <https://ec.europa.eu/eurostat/web/gisco/geodata/reference-data/administrative-units-statistical-units/urban-audit#ua18>

<sup>2</sup> <https://ec.europa.eu/eurostat/web/gisco/geodata/reference-data/population-distribution-demography/clusters>

<sup>3</sup> [Overview - GISCO - Eurostat \(europa.eu\)](https://ec.europa.eu/eurostat/)

This initial AoI had to be refined. The process of refining the initial AoI is described below and illustrated in Figure 6-2.

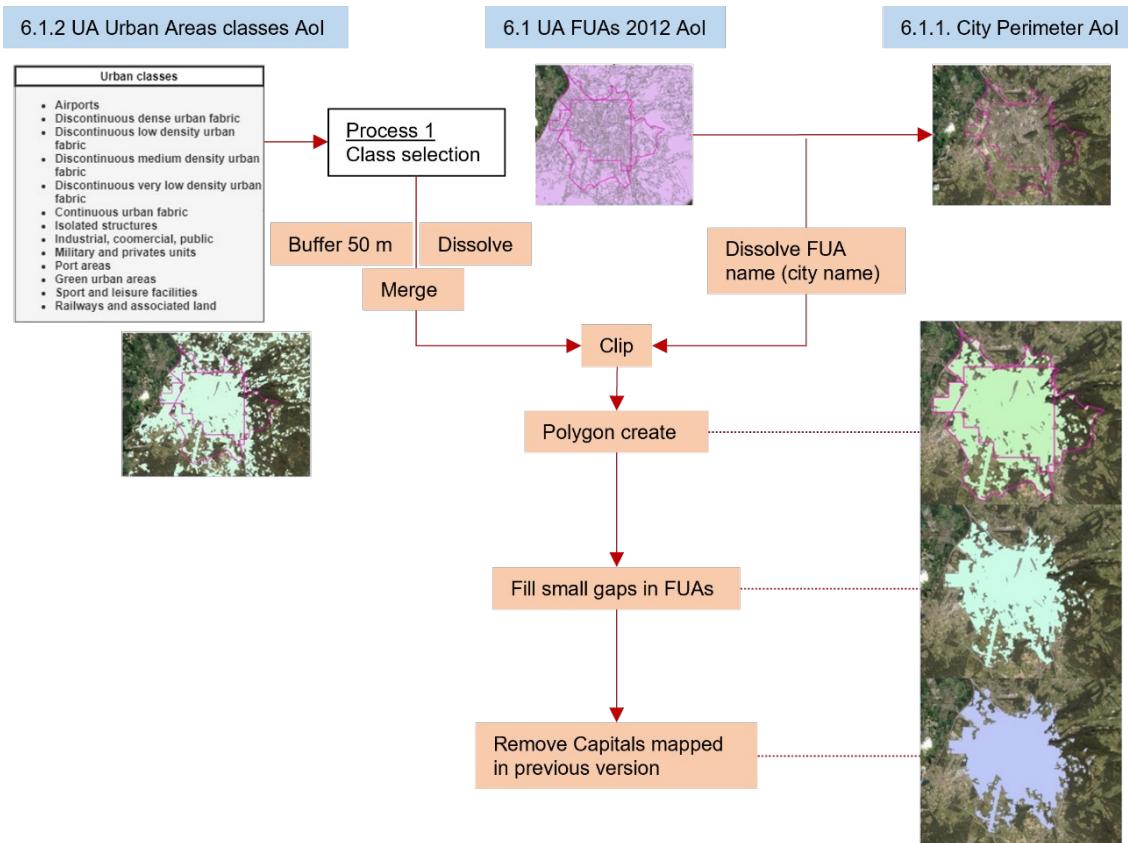


Figure 6-2 Flowchart of the AoI refinement process

### **6.1.1 City Perimeter (City + High density clusters polygon)**

The City Perimeter is the outer delineation of the merging of the Urban Audit cities for 2018 and “high-density clusters” (also called “urban centres”) for 2011.

All these files originate from different dates but were the best available at the time. The Urban Audit cities are from 2018, the High Density Custers from 2011, while the UA underlying LC/LU data is from 2012 and the satellite information are based on 2012 data, or as close as possible to 2012, see further in chapter 6.2.1.

This AoI was used during the searching of satellite images for the DSMs generation.

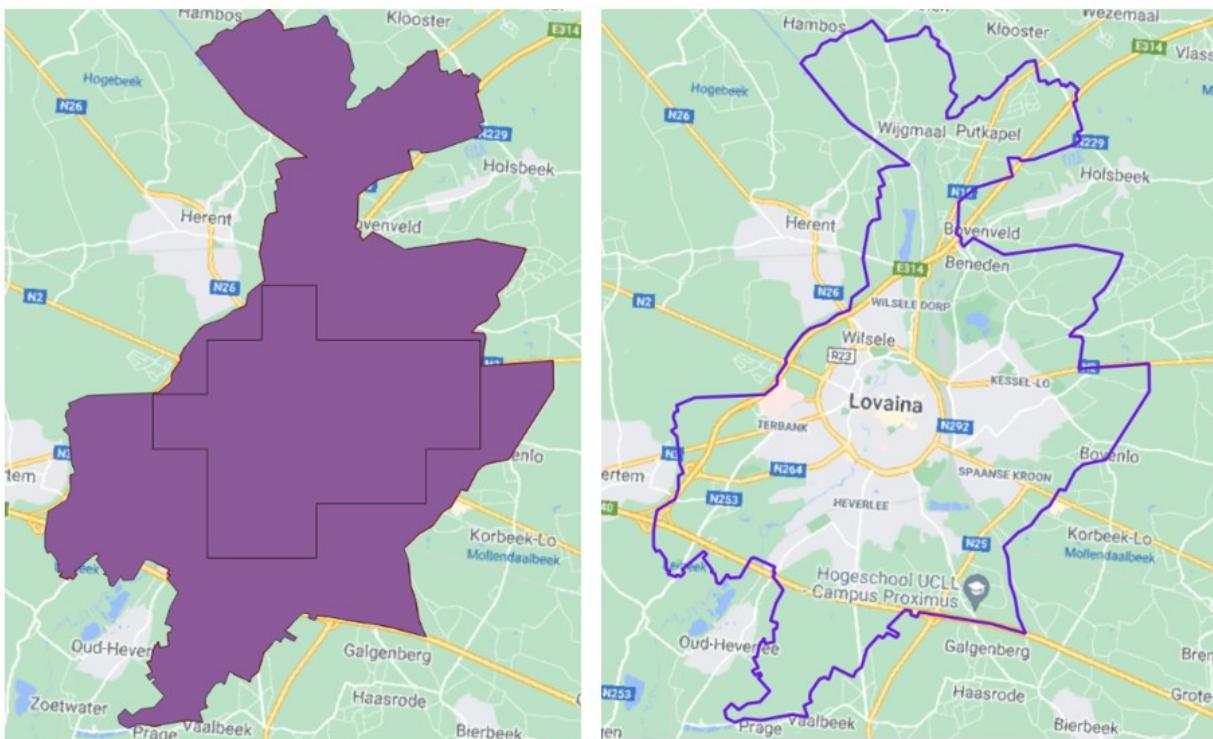


Figure 6-3 Left: City administrative area (purple) and high density cluster (black line inside purple field) of Leuven. Right: City administrative area and high-density cluster merged into City Perimeter layer

### 6.1.2 UA Urban Areas Classes

The second step in refining the AoI for the BBHM product is to select the relevant LC/LU classes from Urban Atlas where buildings are expected to be found, i.e., any classes belonging to codes 1XXXX, but not for example forests or pastures or any of the classes with codes 2XXXX-5XXXX. For reference, the nomenclature of Urban Atlas LC/LU product can be found in pages 19-21 of the Nomenclature guideline<sup>4</sup>.

The selected classes from the Urban Atlas representing urban areas can be seen below:

- Airports
- Discontinuous dense urban fabric (S.L.: 50% - 80%)
- Discontinuous low density urban fabric (S.L.: 10% - 30%)
- Discontinuous medium density urban fabric (S.L.: 30% - 50%)
- Discontinuous very low density urban fabric (S.L.: < 10%)

---

<sup>4</sup>

[https://land.copernicus.eu/user-corner/technical-library/urban\\_atlas\\_2012\\_2018\\_mapping\\_guide](https://land.copernicus.eu/user-corner/technical-library/urban_atlas_2012_2018_mapping_guide)

- Continuous urban fabric (S.L.: > 80%)
- Isolated structures
- Industrial, commercial, public, military and private units
- Port areas
- Green urban areas
- Railways
- Sports and leisure facilities

This AoI (UA Urban Areas Classes) is used to clip the City Perimeter AoI, to represents areas where buildings are expected within the defined City Perimeter AoI.



Figure 6-4 Comparison between City Perimeter (purple) and UA Urban Areas Classes layer (red)

A buffer of 50 m is added around the borders to avoid border errors, see Figure 6-5.



Figure 6-5 Selected UA Urban Areas classes, city of Leuven (Belgium). Left: Final result of selected classes. Right: Example of a buffer area applied to expand the original margin.

The result is the UA Urban Areas classes + 50 m mapped within the City Perimeter boundaries. The polygons within this shapefile are then checked for small gaps, which are filled.

Whenever UA LC/LU classes were missing within the City Perimeter of a selected city, these missing polygons were manually created and added. In some cases, the same UA Urban Areas classes were included in two separate cities. Depending on factors like distance between the cities, FUA codes and city codes in UA, these cities have either been considered one and the same (when close together for example) or separate cities (for example when there is a big distance between them, or the cities have different city codes in the UA).

The previous release of the BBHM included 38 European capital cities. These areas have not been remapped in this release, except for in cases where new areas have been added to the city's FUA (a total of 13).

The total number of cities mapped in this release of the BBHM is 845.

### 6.1.3 UA roads, water, and wetland

An additional shapefile extracted from UA is created including the UA classes of: Other roads and associated land, Water or Wetlands, see Figure 6-6. This layer is useful to eliminate pixels that do not represent buildings.



Figure 6-6 Urban roads, city of Leuven (Belgium). Left: Land uses (orange), UA Urban Areas Classes. Right: Example of class selection. (red).

The following example shows the different Aols applied for the city of Münster, located in the west of Germany. Figure 6-6 shows the three types of Aols used in the refinement process Anything in the City Perimeter AoI which does not correspond to buildings is shown in red (will be given the value NoData in the final dataset); the UA Urban Areas Classes AoI are shown in orange (numerical values in dataset if building is present), and the UA roads, water, and wetland AoI is shown in pink.

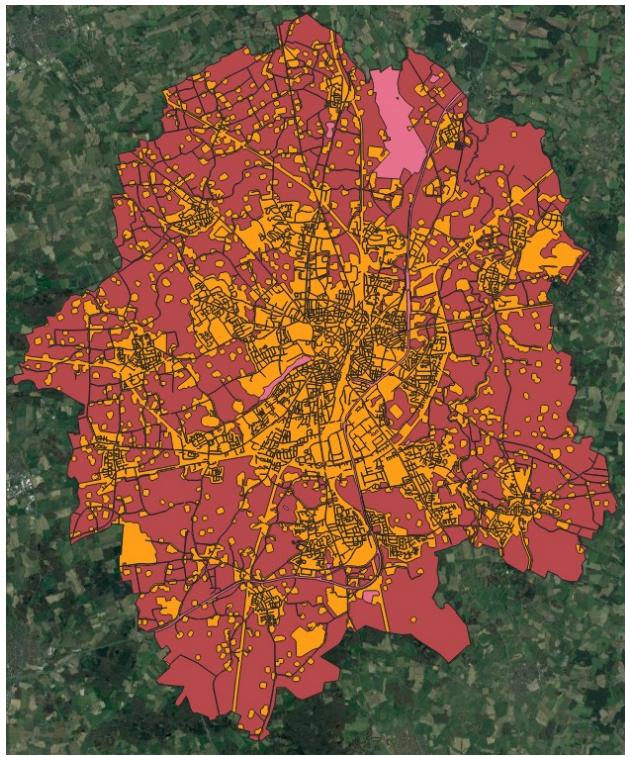


Figure 6-7 Examples of the Aols for Münster, Germany. NoData (red), UA Urban Areas Classes (orange) and Urban Road (pink)

#### **6.1.4 Building Footprints**

Building Footprints represent the individual constructions that were present in 2012 within the Urban Atlas. The Building Footprints layer is obtained by applying an elaborated Python algorithm to the selected high resolution satellite images, which leverages a semantic segmentation model to extract and separate individual buildings in densely compacted areas. The algorithm considers the pixel information of the multispectral and panchromatic images.

Convolutional neural networks (CNN) architecture is used for semantic segmentation of building footprints from multispectral images, and to label every pixel in the images. After each convolution layer an extra batch normalization layer will be added to avoid any extreme values, reducing sensitivity towards initial weights and reduce overfitting of the model.

Basic morphological operation on binary data will be used for noise removal. Thereafter a Local Maxima approach will be used to separate connected buildings, by allocating local maxima and higher weight to the larger area out of building and connection. The Local Maxima is used as input source/sink in a watershed algorithm used for image segmentation. Each building obtained from this process is given a unique index for further processing. The result is a raster.

For areas with dense vegetation, the algorithm might generate some confusing pixels. Therefore, a review of the Building Footprints layer has been carried out for

areas with dense vegetation. Hereafter a quality control is conducted to verify that no buildings or constructions were excluded, and to discard false positives.

The raster is then vectorized. Spatial references and coordinate systems are preserved at every step and are transferred to vector file for correct overlaying. While converting from raster to vector, the output has a lot of vertices and noises. The vector is then simplified using Douglas-Peucker algorithm (David Douglas, 1973) to help preserve the overall geometry of the shape while simplifying number of vertices. Basic attributes like area, perimeter and elevation of the buildings were automatically added. Finally, minimum bounding box was used and saved.

The resulting Building Footprints was later used in the BBHM production as a mask to separate areas with height data values from areas with no height data values, see chapter 6.6. If there is building footprints data available in the area, this will be considered for refining the AoI of each city.



Figure 6-8 Building footprints for the city of Leuven (Belgium)

## 6.2 Data pre-processing

The main input data is comprised of VHR satellite images from 2012, on which the different algorithms for producing the DSM and DTM will be executed, see chapter 6.3 and 6.4. The different sensors used for the VHR image collection are, as shown in Table 5-2, Worldview 1, 2, 3, Quickbird, Ikonos and Geoeye. The aim is to acquire stereoscopic pair images, overlapping imagery acquired from different locations, to produce a 3D model for the refined Aols.

The worldwide coverage of stereoscopic pair images is very limited, as it is not the most common and easy data to capture. Satellite data might therefore, despite the vast number of images, not be sufficient to create stereo pair images of all the refined Aols in the mapped FUAs. Instead, false stereo pair images can be used. False stereo pair images make use of two images taken from different angles, and/or not on the same date or in a simultaneous capture. The images that form the false stereo pair should however be taken as close as possible to each other in time



to improve the correlation between them, which increases precision. This approach guarantees coverage of all the refined Aols with images from different dates. To create false stereo pairs the VHR satellite images from MAXAR products are supported with Copernicus satellite images through Copernicus Reference Access Data (CORDA), and as a further option LiDAR information.

In cases where no satellite image close to year 2012 is available, and also no images from CORDA, the search date range is expanded, primarily around 2012 (Q4 for 2011 or Q1 for 2013) and secondarily for later years (2014 onwards) as later years have more available satellites (WV3 etc.) from which data imagery can be gathered.

The process of selecting suitable VHR, CORDA and LiDAR imagery was based on the following key factors:

- Imagery availability from the MAXAR Catalogue, CORDA
- Date of capture, see detailed description in chapter 6.2.1
- Off-Nadir Angle and Cloud Cover, see detailed description in chapter 6.2.2

### **6.2.1 Date of capture**

The following are the relevant scenarios for fulfilling the “Date of capture”-criteria for the selection of VHR and LiDAR images:

- **Scenario 1:** The best option for selection of images is the use of LiDAR information from dates as close as possible to 2012, when available. This scenario shows better results since there is more density and precision.
- **Scenario 2:** The second best option is images from 2012 taken in the same month and in the same day. Part of the area has been reviewed and is covered with images from 2012. Where images are not available, LiDAR information from 2012, or on a date as close as possible to 2012, will be prioritized for generating the BBHM.
- **Scenario 3:** The third best option is to use images from 2012 captured in the same month but on different days.
- **Scenario 4:** The fourth best option is using images from second half of 2011 or first half of 2013 in the same month and in the same day.
- **Scenario 5:** The fifth best option is images from 2012 with a difference in date of capture of approximately 1 month.
- **Scenario 6:** Scenario 2 but in the first part of 2011 or in the second part of 2013
- **Scenario 7 (extra):** Images from 2014, 2015 and 2016... until today of similar dates (i.e. captured on the same day or week). It has been shown that images of the same season and similar date give good results for obtaining elevation models even though the dates differ by one year. The only thing that would affect this scenario is in terms of large changes in apples that



were not in one of the years being analyzed. In this scenario, an analysis of change between the two dates will be made. Thanks to the refinement of the AOL the commission and omission error will be minimized because the Refined AOL is focused on the Urban Areas.

The images are selected and downloaded from the MAXAR library and organized in pairs of images for each city. The images are separated by CAT IDs, which is a unique code for each image. Images were then quality controlled for errors, reviewed and validated for use, as well as merged to ease the further processing of the images. This phase generates as a result a list of selected images to be used in the next analysis, where the algorithms for creating the DSM and DTM are applied.

## **6.2.2 Off-Nadir Angle and Cloud Cover**

The most suitable satellite images were selected, in the case of Off-Nadir angle, based on base-to-height ratio, Convergence, BIE and Asymmetry Angle. If these values, when estimated, are not in the correct range, the DSMs are not correctly built.

### *6.2.2.1 Off-Nadir Angle*

The best is to have images with an Off-Nadir angle of 20 degrees. Pairs formed with images that have less than 5 degrees or more than 45 degrees Off-Nadir angle give bad results, and are discarded. As far as possible, images with Off-Nadir angles of more than 40 degrees have been avoided. From these criteria two scenarios are made:

- Scenario 1: Pair of satellite images with Off-Nadir angles of between 10 and 20 degrees
- Scenario 2: Satellite images with Off-Nadir angles of between 5 and 30 degrees, preferably between 5 and 20 degrees

For urban areas, large angles of convergence would lead to ground occlusions, as well as a loss of similarity between the two images that form the “stereo pair” and, consequently, a worsening of the process of searching for homologous points by automatic correlation (matching). For these reasons, small Base/Height (B/H) ratios are preferred in urban areas (Eckert and Hollands, 2010). Using an algorithm that considers the maximum and minimum incidence angle, the time difference of the images, the angle between the two images and the region of interest, a Pearson’s correlation matrix and a B/H coefficient as a function of the orbital geometric parameters of the image, as well as temporal parameters are obtained, thanks to which it is possible to find which “false stereo pair” image is the most suitable to use in the generation of the DSM, see chapter 6.3.

The selected false stereo pair images have an angle difference of 20 degrees and none of them beyond 35 degrees.

### *6.2.2.2 Cloud cover*

The aim is to generate cloud-free results, for this reason all scenes were selected, regardless of cloud cover. This approach maximized the results, facilitating the



selection of the best scenarios. In case some scattered cloudiness was found in an image, a proprietary cloud masking algorithm based on multilayer neural networks was applied. If clouds covered buildings in an image, this image was discarded. In case some scattered cloudiness could be found, a proprietary cloud masking algorithm based on multilayer neural networks was applied.

## 6.3 Generating the DSM

Based on the (false) stereo pair images, a digital surface model (DSM) is generated, which represents the height of all objects above the earth's surface, i.e. plant/forest canopy or the top of artificial constructions such as buildings.

The major steps in creating the DSM are:

- Selection of suitable (false) stereo pair images
- Pixel-wise stereo correspondence
- Void filling
- Automated outlier detection and removal
- 3D point cloud creation, including RPC improvement
- Projection, alignment and homogenisation of the 3D point cloud
- Process QA

The available false stereo pair images for a certain AoI, obtained through the process described in chapter 6.2, are further processed to generate the DSM. The method for generating the DSM is based on taking elevations from the terrain using the false stereo pair satellite images, and by choosing the most suitable false stereo pair images. The images chosen are panchromatic with high spatial resolution of 30, 40 or 50 cm, and RPC information.

Thereafter a stereo correspondence between the two chosen images that make up each false stereo image pair is conducted, using a patented-modified version of the Semi-Global Matching (SGM) algorithm. The algorithm is based on global energy minimization that comprises a cost of equalization in pixels and smoothness terms in disparity mapping. The algorithm then applies a 3-kernel median filter to fill in small holes in null match points in the disparity map. By implementing a version of the classic RANdom SAmple Consensus (RANSAC) based on reverse methodology, outliers are eliminated.

Lastly, a 3D point cloud is obtained by triangulating the stereo correspondences between the two panchromatic VHR images, taking into consideration the rational polynomial coefficients (RPC) camera model provided for each image. The RCPs will allow triangulating the position of a 3D point identified on an image pair. This elaboration of the Digital Surface Models (DSM) is carried out using an AI-EOSURFACE algorithm on Python and C++ (and compatible with C heads) based on the epipolar geometry of a given pair of satellite images, see Figure 6-9.

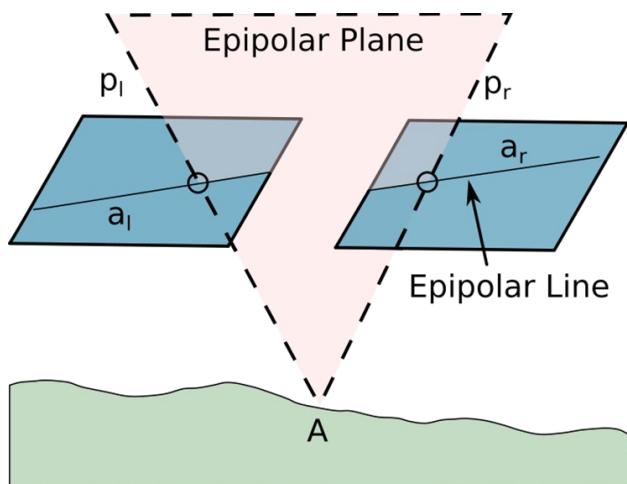


Figure 6-9 Schematic representation of the epipolar geometry for a given pair of satellite images, on which the AI-EOSURFACE algorithm is based to obtain elevations

This algorithm has been developed for the DSM generation based on the bidimensional couple of points of a pair for the later three-dimensional densification. The dataset includes images from different platforms, all of them with a spatial resolution of 0.5 m. The DSM generation only requires the panchromatic band, however RGB and NIR bands were included as these are required for other processes of the BBHM creation.

The independently calculated 3D point cloud is then projected, aligned and homogenized, and the DSM is generated with a spatial resolution of 50 cm, and all DSMs are transformed to an orthometric height because to transform height data into meters of above sea level.

## 6.4 Generating the DTM

Digital terrain models (DTMs) represent the height of the bare earth's surface. In this case, the DTMs are generated from the DSM. Non-ground features such as buildings, trees, bushes and vehicles have to be identified, and their height measurements removed from the data before interpolation. Once the DSM has been generated, it is filtered using a building emptying algorithm and removing objects that are not bare ground. Thereafter a pixel level prediction and clustering approach is used on a high-resolution multispectral image to detect the bare terrain.

The DTM is carried out using an AI-EOTERRAIN algorithm on Python to process each row of the given DSM separately in all directions in a window with a given extent, considering the minimum value of each row as the position of the ground. Therefore, if the pixel being evaluated has a large difference with the minimum value, the pixel is considered as not corresponding to ground, as well as when the slope between the evaluated pixel and the next one is positive and above a predefined threshold. If the slope is positive but less than the threshold, the evaluated pixel gets the same label as the pixel evaluated in the previous iteration.

If the slope is less or equal, the distance to the nearest ground point is used to decide whether the evaluated pixel is classified as ground or not.

The building extraction process was successful with most buildings, however, in buildings that occupy a large area, the algorithm interprets the innermost areas of the roofs as ground values, due to the homogeneity of the slopes. To ensure that all values of the emptied DSM correspond to ground values, a second filtering is done using the buildings footprints layers, see chapter 6.1.4.

*DSM algorithm | Building footprints | DSM filtered*

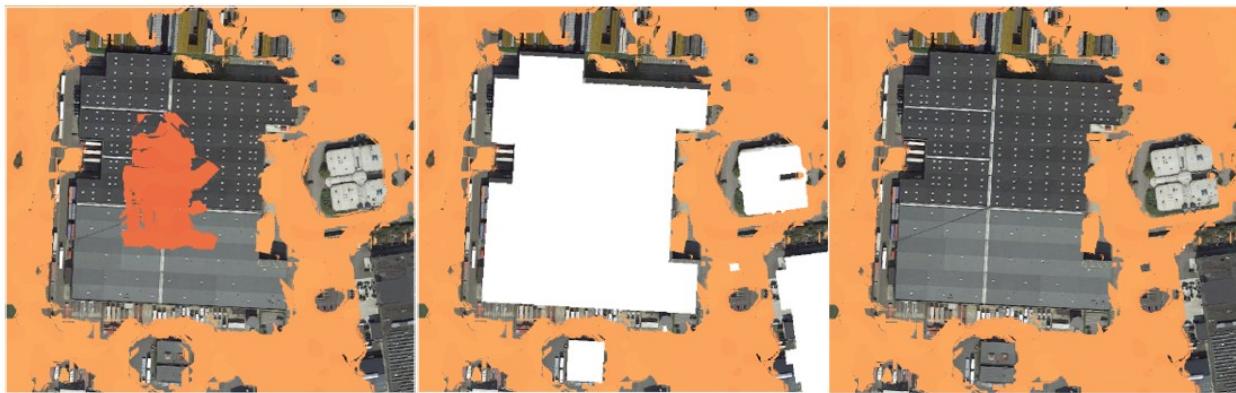


Figure 6-10 DSM filtering to remove large buildings and obtain the DTM

Thereafter the remaining bare ground is filtered by a cut off range for objects to remain in the DTM. This is based on the gradient difference, the size of the object to be removed and the terrain type, this can be either flat or hilly. A “bump and pit”-filter is applied to smooth the DTM by removing small pits and bumps in the landscape; a median filter which is part of the final smoothing of the DTM; and finally the clamp filter, which stabilizes portions of the elevation model by raising and lowering pixels within a defined threshold.

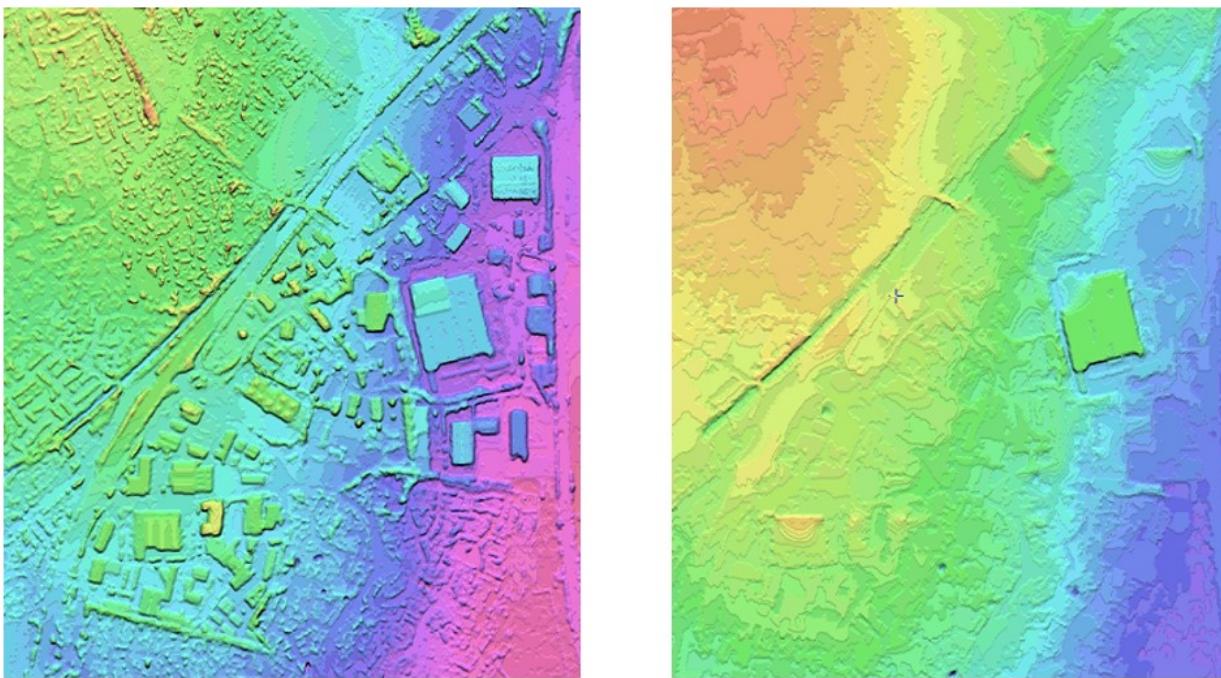


Figure 6-11 Left: DSM for an area in London. Right: DTM for an area in London

The generated file retains the same properties as the DSM in terms of extension and spatial resolution.

## 6.5 Generating the DHM

Through a subtraction between the DSM and DTM by means of raster file algebra, and nearest value resampling for the Digital Height Model (DHM) file is obtained.

The result of the operation of generating the DHM is a GeoTIFF raster file that represents the height values of the entire area encompassed by the AoI, layers that are not of interest are given a value of 0 for now. Figure 6-12 shows the resulting DHM derived from this process.

DSM | DTM | DHM



Figure 6-12 DHM production

The result of the operation is one DHM file per city and per image pair, in Int32 format.

## 6.6 Generating the BBHM

Once the DHM of the city is available, it is necessary to extract only the height values that correspond to buildings. To do this, first all the buildings are detected from the available multispectral VHR images, using machine learning techniques. Then a building footprint extraction operation is performed, obtaining a GeoTIFF raster layer with the height values of the buildings, and with a NoData value for all those elements which are not buildings.

To extract the values that correspond to buildings, the vector files of the building footprints and the DHM files are required as input.

Since the building footprints layer are in vector format with complex geometries, these are transformed into Boolean raster format, where the pixels with zero values represent the areas that do not include buildings.

Once the DHM file and its associated building footprints file are both available in raster format and with the same extension, a multiplication is performed, resulting in a DHM with height values only in the areas occupied by buildings. Finally, all pixels with zero in height value, i.e. the height assigned to all elements other than buildings, are reclassified to a NoData value, as otherwise they could have a negative influence on subsequent processes. This is an important difference to note between the 2006 and 2012 edition of the BBHM. It was introduced in this release since in the previous release it was not possible to distinguish between satellite measured zero values, and zero values assigned because of the type of area. Instead, in this release, all areas not representing buildings have a NoData value.



Finally, the DHM has undergone resampling through a number of filters; reprojection to EPSG3035 coordinate system, removal of pixels outside of the City Perimeter AoI, removal of pixels outside of the UA Urban Areas AoI, removal of pixels within the Urban roads AoI and finally resampling from a resolution of 0.5m to a resolution of 10 m using a majority filter.

This point explains why it is necessary to translate the 0 height data values of the input DHM to NoData, since otherwise the majority of values in a pixel corresponding to a building could be equal to 0 due to the influence of the surrounding street.

It is also important to point out the role played by vegetation in this process, since, as the DSM is derived from stereoscopic images; it is possible that trees near buildings influence the final value of certain pixels.

Also, a reclassification of all pixels with a value lower than 3 m is performed, since it is assumed that no buildings are lower than 3 m.

Finally, a final conversion of the DHM files is executed to adapt them to the requirements set according to CLMS Tools needs.

- EPSG:3035 coordinate reference system
- GeoTIFF data type UInt16.
- No data value equal to 65535.
- Compression type LZW
- Pixel size 10 m
- BLOCKSIZE 256x256 Specification
- TILED = YES

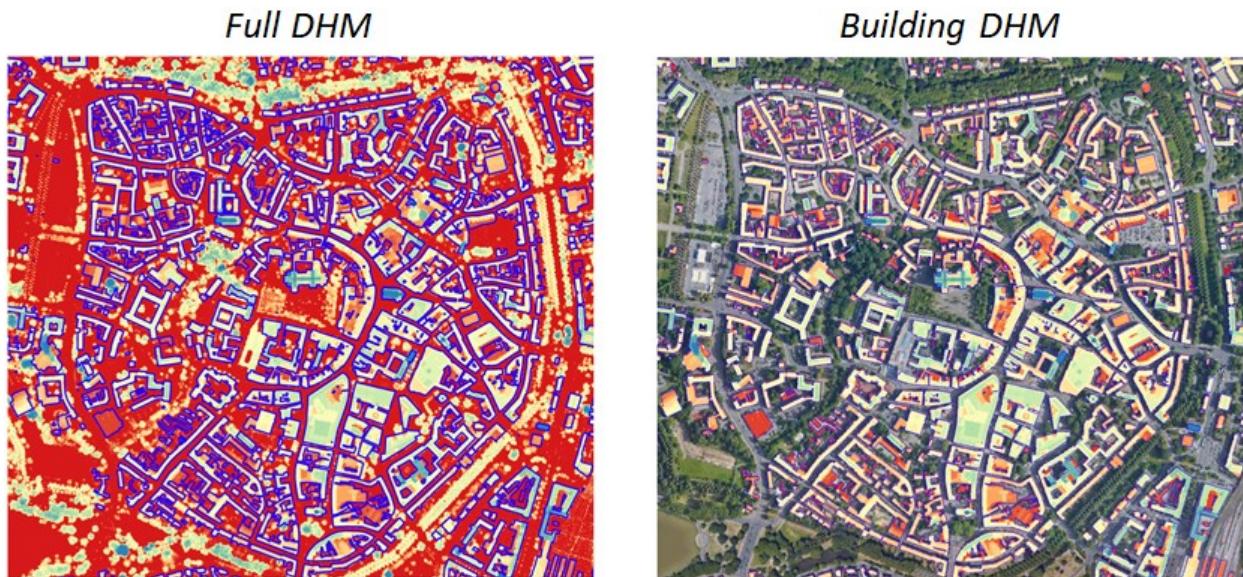


Figure 6-13 From a general DHM (left) to a BBHM (right) in Münster, Germany



Figure 6-14 Final BBHM at 10x10 meter resolution for Münster, Germany

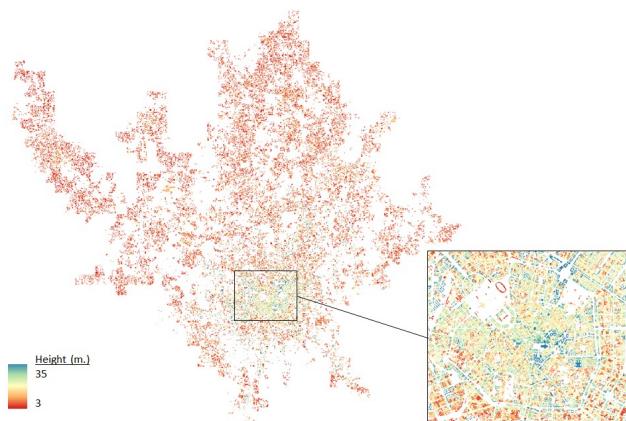


Figure 6-15 General overview of building height values on Milano and zoom to the city center

## 6.7 Characteristics and limitations of the product

This chapter describes the properties of the product and some methodological aspects that are important to understand the characteristics and limitations of the product.

It is important to note that because the DSMs are derived from LiDAR and VHR satellite information, the initially generated DHMs have a spatial resolution of between 0.5 and 2 meters. For the final product to comply with EEA technical requirements (spatial resolution of 10 meters), a resampling operation was performed to transform the DHM pixels from 0.5-2 m to 10 m.

For this transformation to be consistent with the first height values obtained, it was agreed to use a majority algorithm, i.e. the calculation will determine the most common value shown by the 0.5 - 2 m pixels contained within a new 10 m pixel, which will be assigned the most repeated value after the execution of this analysis.

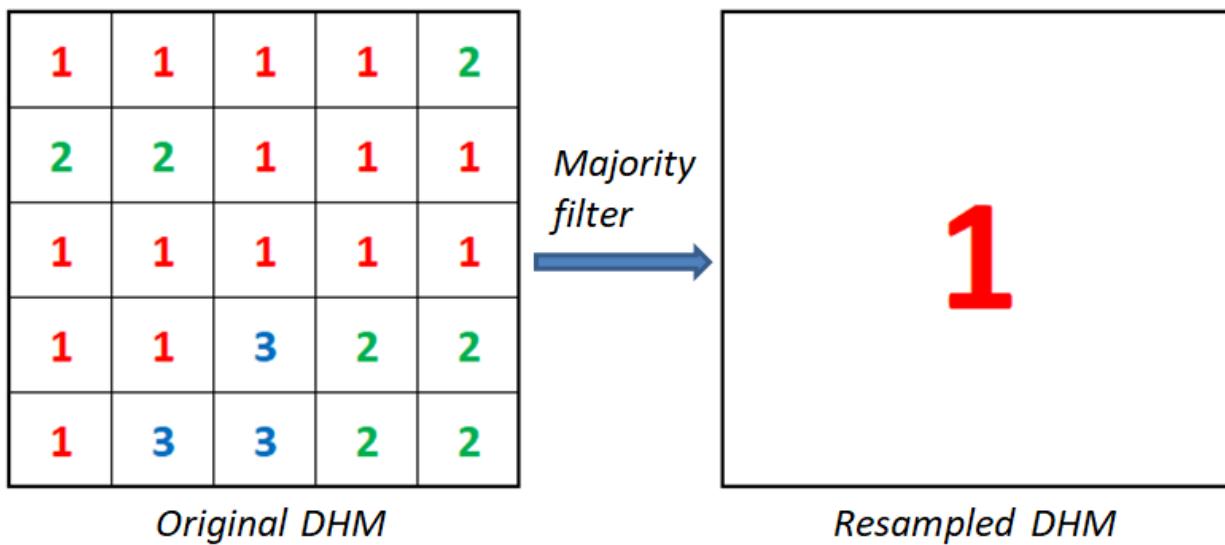


Figure 6-16 Methodological explanation about how the majority filter works when resampling

It is this simple resampling operation that poses the main limitations of the BBHM Urban Atlas product, which are further detailed below.

- **No height values in buildings that are smaller than the minimum mapping size**

Due to the spatial resolution of the pixel, there will be a NoData value in buildings smaller than 100 m<sup>2</sup>, such as single-family houses, buildings for public use etc., see example in Figure 6-17.



Figure 6-17 Small houses that have not been captured in the BBHM due to the minimum mapping size

- **Issues on building edges**

The influence of height values of elements or artifacts adjacent to the buildings can induce abnormally high or low values in the pixels of the edges of the buildings. In

case of influence of the ground abnormally low values would be presented, or abnormally high values if it is a tree or other vertical element that interferes.

This happens, in addition to the majority algorithm, when extracting the height values within the building footprints. This is because it is very unusual that the outline of the building represented in the DHM coincides exactly with its building footprint, which causes the edge pixels to be retained in the DHM prior to the resampling process and, therefore, may interfere with the final result.

The example in Figure 6-18 shows the height values obtained for a random building, in this case obtained by means of LIDAR technology, with a spatial resolution of 2 m.

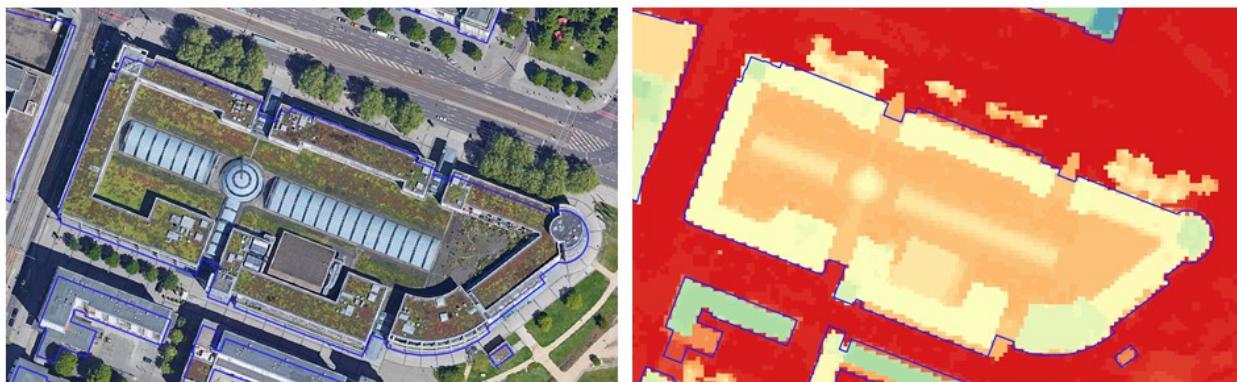


Figure 6-18 Satellite imagery (left) and DSM at half meter resolution (right)

Figure 6-19 shows how some pixels, with ground elevation information, are represented in red tones after extraction of the DHM values in relation to the building footprints.

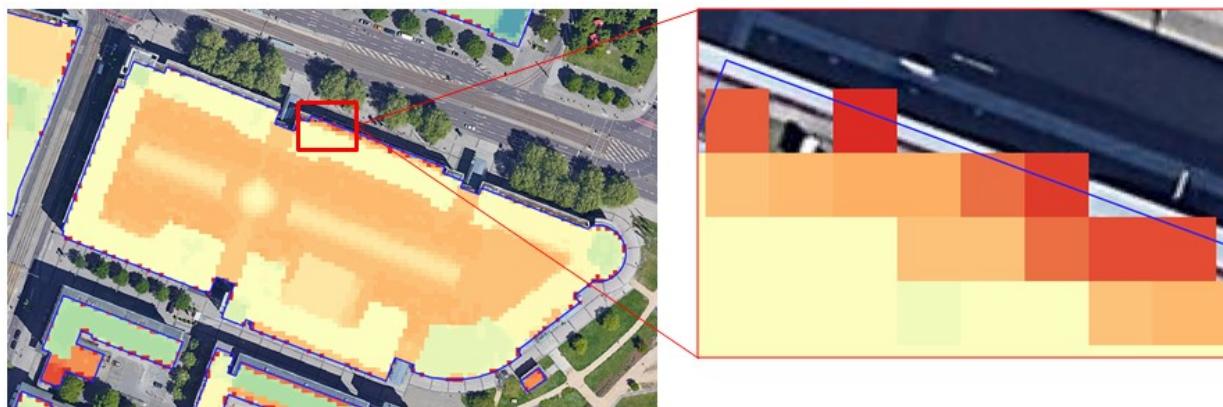


Figure 6-19 Difference between DSM at half meter resolution and final BBHM (right)

When the resampling operation is performed and the final result is obtained, this ground elevation information is transferred to the pixels that make up the building boundary, causing the aforementioned alterations in the value of these pixels.

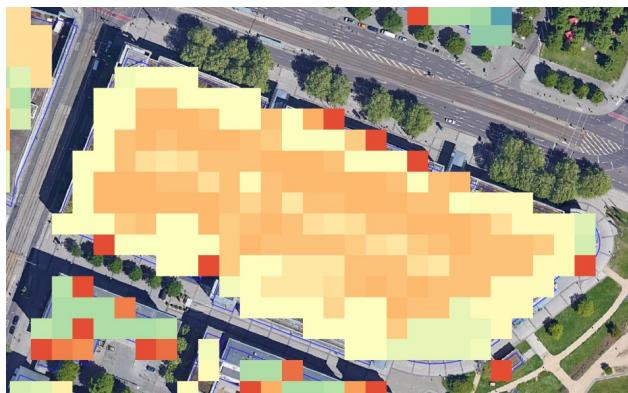


Figure 6-20 Overview of building edges in the BBHM

Without a detailed case-by-case study it is not possible to conclude whether these height deviations are caused by elements of the building itself or by influences from the ground or other artifacts. Thus, these pixels have not been corrected beyond those clearly identified during QC, which is the main limitation of the project.

The influence of edge elements becomes more noticeable in cities whose DHM values have been calculated from a DSM derived from VHR satellite imagery. This is because optical satellite photogrammetry technology does not reach the same levels of accuracy as achieved by LIDAR, and its results depend heavily on the acquisition properties of the images. Especially important are the angles and the correlation calculated between the two images, giving the results a more continuous and smooth appearance, i.e. there is no perfect detection of abrupt elevation changes intrinsic to the height difference between a building and the ground.

This behavior is especially noticeable in cities with a high density of buildings, close together, separated by narrow streets, typical in urban centers of historic cities. This combined with the resampling algorithm can result in areas with apparently great variability of building heights. This phenomenon can be observed in Figure 6-21.

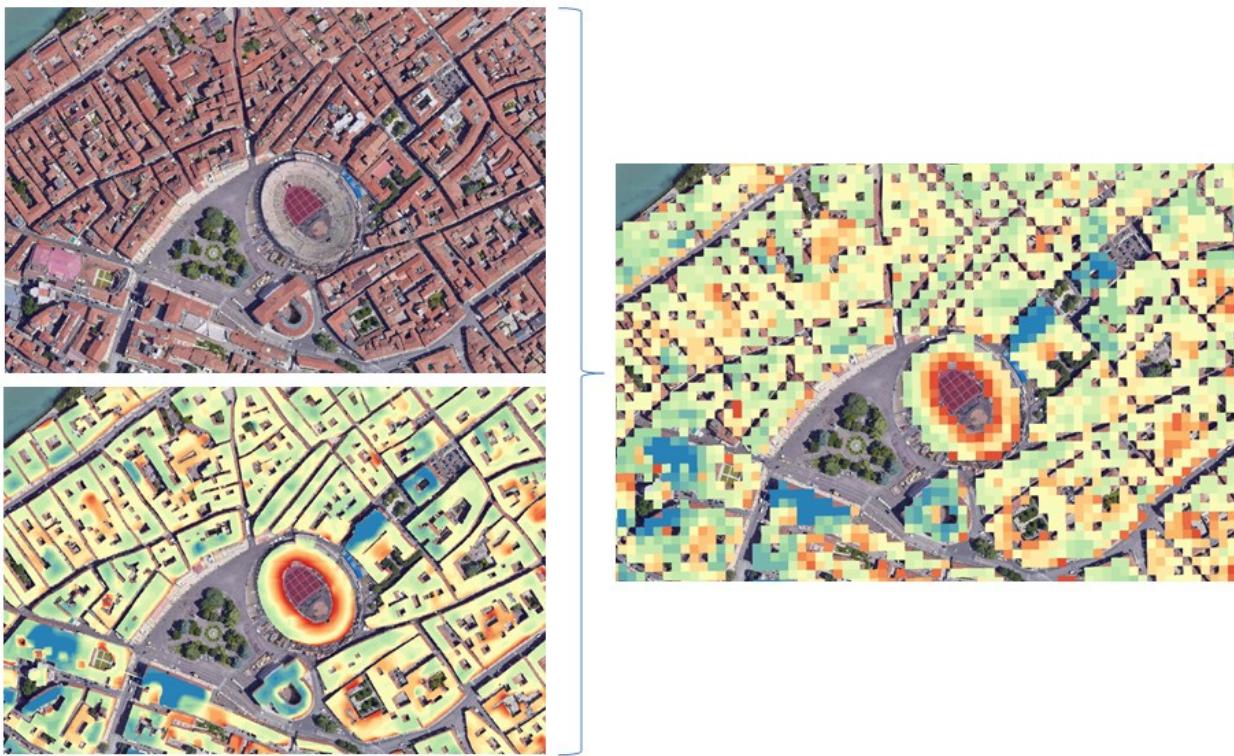


Figure 6-21 Higher heterogeneity in BBHM dataset (right) than in a half meter resolution BBHM (bottom left)

- **Roof artifacts**

Another limitation of the product lies in the poor response to the LIDAR pulses of some roofs, especially noticeable in the metal cladding that covers some industrial buildings. These errors do not occur when using satellite photogrammetry and thus do not pose an issue in most of the processed cities. To correct the errors, the erroneous pixels are manually reclassified with the correct elevation value.

- **Limitation of building footprints detection in scattered areas**

The building footprints identification algorithm based on Machine Learning techniques has certain limitations when detecting isolated buildings, especially in rural environments, since the algorithm considers certain properties of coherence and spatial context for the identification of buildings. In the same way, the algorithm may present commission errors, which have been eliminated whenever they have been detected during QC.

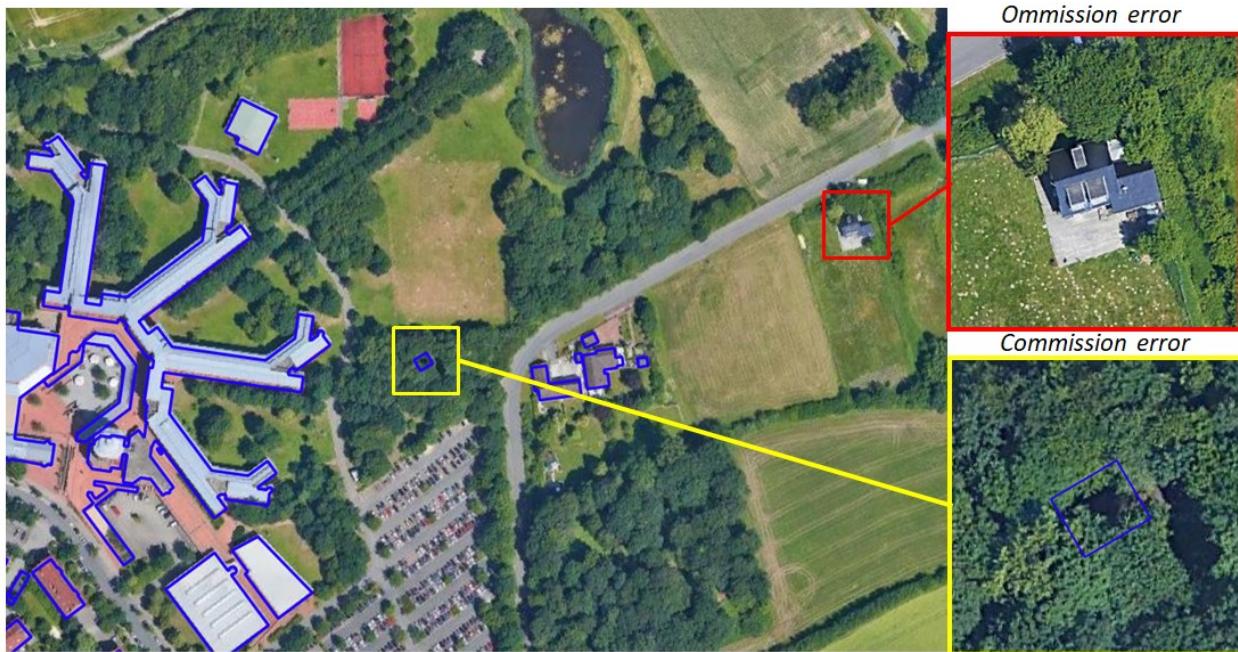


Figure 6-22 Limitations on building footprint detection especially; comision and omission

- **Urban Areas with no values outside the AoI**

As stated previously in the document, building height values are only gathered inside the AoI (see definition in chapter 6.1). In some cases, it is not possible to produce building height data for the entire delineation of the AoI, i.e. when part of the defined high-density cluster delineation falls outside the area that has been mapped by Urban Atlas. For example, the AoI for Malinas is a high-density cluster but is not mapped by Urban Atlas and does not have a FUA (no Urban Classes). Thus, it is key to consider that in these cases some urban areas in the city surroundings are not captured, so there is no height value.



Figure 6-23 General view of Malinas, capturing industrial areas with no height values outside the AoI

- **Limitations of DSM generation**

DSM generation from false pair stereo of satellite images is always challenging, since it depends on several variables: off-nadir angle, difference in days between images, cloud cover, etc. These values vary from city to city, but in general terms these have been the main issues:

- DSM orthorectification
- Building shadows
- Small houses' height
- Outlier removal
- Water-topography influence

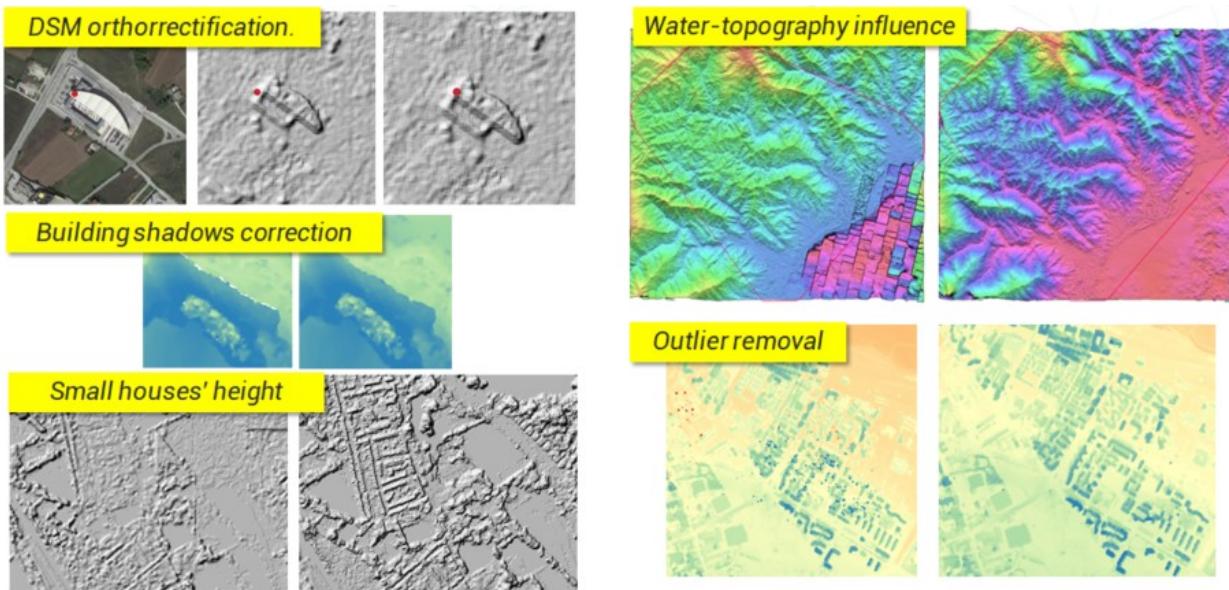


Figure 6-24 Examples of the limitations of the DSM production

- **Limitations on DTM generation**

When generating a DTM from DSM, the reliability on the accuracy of the DSM falls, e.g. a DSM derived from a high-density point cloud by using LiDAR information is more reliable than one built from satellite imagery. Also, an automatic algorithm pursues a constant accuracy across different realities, setting parameters to fix values that better capture all. However, larger buildings (e.g. industrial buildings) may not be removed completely, so a manual editing has been carried out in all the cities.

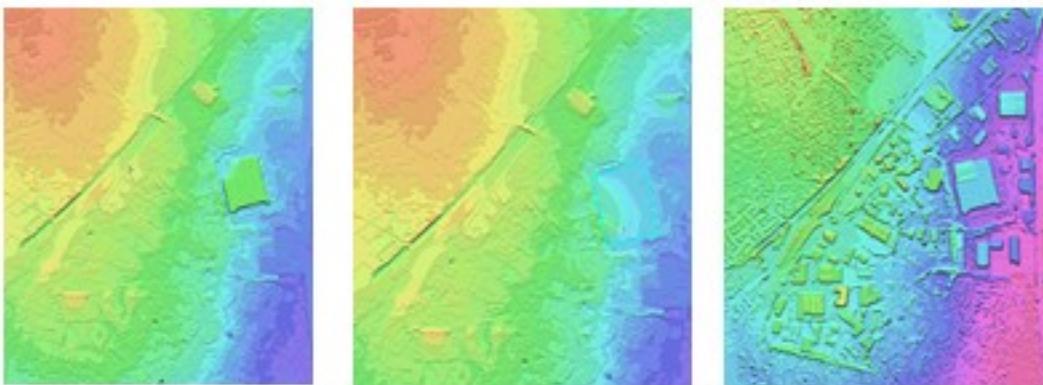


Figure 6-25 Example of DTM refinement



## 7. Quality Assessment

The UA BBHM product has been produced by service providers with accredited ISO 9001:2015 quality standard management systems. Quality control of the final product level has been divided into the following three stages:

- Data completeness check and data compliance with the client guidelines.
- Data quality control at building level.
- Data quality control at block level.

### 7.1 QC DSM

Before passing the DSM on to the DTM construction the following checks and corrections are made:

- Orthorectification, if erroneous this is corrected through georeferencing in QGIS with Google Satellite image for reference
- Height conversion (ellipsoidal-orthometric)
- Height review between DSM and DTM
- Outlier removal, corrected by selecting smaller AoI and generating new DSM without outliers
- Limitations (small buildings), if not detected in areas with a lot of vegetation this is corrected by changing the parameters of the smoothing filter in the affected areas
- Gap filling, very tall buildings' shadows cause no data areas and holes in the data mesh between the building and adjacent ground, this is corrected by a data filling algorithm that interpolates the closest values in the building and ground levels to cover these gaps

### 7.2 QC DTM

Before passing the DTM on to the DHM construction checks and corrections are made through refinement by adding different filters.

### 7.3 QC DHM

After finalisation each generated DHM file is manually reviewed, verifying that all the constructions identified in the building footprints have an assigned height value consistent with reality. During this manual review, it is also verified that there are no pixels with abnormal height values, relying on the altimetry information provided by Google Earth and Street View.



## 7.4 Quality Assurance (QA) and Quality Control (QC) measures

Subsequently a third-party quality control is carried out in three stages

1. Data completeness check and data compliance with the client's guidelines check
2. Data quality control at building level
3. Data quality control at block level

After acceptance of the data by the third party, the data is checked additionally using the CLMS QC Tool, which checks for discrepancies in:

- Final delivery files naming
- Correctness of metadata with the INSPIRE specification
- Raster EPSG code
- Raster pixel size
- Position of the upper left corner of raster bounding box
- Raster bit depth
- Raster compression format
- Raster value range
- Raster tiles
- NoData value

The final data were considered valid if no issues were reported from all of the quality control tasks, including both the third party and the CLMS QC Tool controls. The results are shown in the QC report, see chapter 5.2.5.

## 8. Terms of use and product technical support

### 8.1 Terms of use

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"© European Union, Copernicus Land Monitoring Service <year>, European Environment Agency (EEA)"

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# 10. Annexes

## 10.1 Annex 1 - List of cities mapped

Country code	City name
AL	ELBASAN
AL	SHKODER
AL	VLORE
AL	TIRANA
AT	GRAZ
AT	LINZ
AT	SALZBURG
AT	INNSBRUCK
AT	KLAGENFURT
AT	BREGENZ
AT	WIEN
BA	BANJA LUKA
BA	MOSTAR
BA	TUZLA
BA	ZENICA
BA	SARAJEVO
BE	ANTWERPEN
BE	GENT
BE	CHARLEROI
BE	LIÈGE
BE	BRUGGE
BE	NAMUR
BE	LEUVEN
BE	MONS
BE	KORTRIJK
BE	OOSTENDE
BE	MALINAS
BE	LA LOUVIÈRE
BE	VERVIERS
BE	BRUXELLES BRUSSEL
BG	PLOVDIV
BG	VARNA
BG	BURGAS
BG	PLEVEN
BG	RUSE
BG	VIDIN
BG	SLIVEN
BG	DOBRICH



Country code	City name
BG	SHUMEN
BG	YAMBOL
BG	HASKOVO
BG	PAZARDZHIK
BG	BLAGOEVGRAD
BG	VRATSA
BG	SOFIA
BG	STARA ZAGORA
BG	VELIKO TARNOVO
CH	ST GALLEN
CH	ZÜRICH
CH	GENÈVE
CH	BASEL
CH	LAUSANNE
CH	WINTERTHUR
CH	LUZERN
CH	LUGANO
CH	BIEL BIENNE
CH	THUN
CH	FRIBOURG
CH	BERN
CY	LARNACA
CY	LEMESOS
CY	LEFKOSIA
CZ	PRAHA
CZ	BRNO
CZ	OSTRAVA
CZ	PLZEŇ
CZ	HRADEC KRÁLOVÉ
CZ	OLOMOUC
CZ	LIBEREC
CZ	KARLOVY VARY
CZ	CHOMUTOV JIRKOV
CZ	PARDUBICE
CZ	ZLÍN
CZ	JIHLAVA
CZ	MOST
CZ	ÚSTÍ NAD LABEM
CZ	ČESKÉ BUDĚJOVICE
DE	FRANKFURT AM MAIN
DE	OLDENBURG
DE	HAMBURG



<b>Country code</b>	<b>City name</b>
DE	MÜNCHEN
DE	KÖLN
DE	HALLE AN DER SAALE
DE	STUTTGART
DE	LEIPZIG
DE	DRESDEN
DE	BREMEN
DE	DÜSSELDORF
DE	HANNOVER
DE	NÜRNBERG
DE	BIELEFELD
DE	FREIBURG IM BREISGAU
DE	MAGDEBURG
DE	WIESBADEN
DE	GÖTTINGEN
DE	DARMSTADT
DE	TRIER
DE	FRANKFURT ODER
DE	REGENSBURG
DE	WEIMAR
DE	SCHWERIN
DE	ERFURT
DE	AUGSBURG
DE	BONN
DE	KARLSRUHE
DE	MÖNCHENGLADBACH
DE	MAINZ
DE	RUHRGEBIET
DE	KIEL
DE	SAARBRÜCKEN
DE	KOBLENZ
DE	ROSTOCK
DE	KAIERSLAUTERN
DE	ISERLOHN
DE	WILHELMSHAVEN
DE	TÜBINGEN
DE	FLENSBURG
DE	MARBURG
DE	KONSTANZ
DE	NEUMÜNSTER
DE	DESSAU ROSSLAU
DE	GIESSEN



Country code	City name
DE	LÜNEBURG
DE	BAYREUTH
DE	CELLE
DE	ASCHAFFENBURG
DE	BAMBERG
DE	PLAUE
DE	NEUBRANDENBURG
DE	FULDA
DE	BOCHOLT STADT
DE	LANDSHUT
DE	ROSENHEIM
DE	STRALSUND
DE	FRIEDRICHSHAFEN
DE	OFFENBURG
DE	GÖRLITZ
DE	SCHWEINFURT
DE	GREIFSWALD
DE	WETZLAR
DE	PASSAU
DE	BRAUNSCHWEIG
DE	MANNHEIM
DE	MÜNSTER
DE	CHEMNITZ
DE	AACHEN
DE	KREFELD
DE	LÜBECK
DE	KASSEL
DE	SOLINGEN
DE	OSNABRÜCK
DE	HEIDELBERG
DE	PADERBORN
DE	WÜRZBURG
DE	BREMERHAVEN
DE	HEILBRONN
DE	REMSCHEID
DE	ULM
DE	PFORZHEIM
DE	INGOLSTADT
DE	GERA
DE	REUTLINGEN
DE	COTTBUS
DE	SIEGEN



Country code	City name
DE	HILDESHEIM
DE	ZWICKAU
DE	WUPPERTAL
DE	JENA
DE	GUTERSLOH
DE	BERLIN
DE	VILLINGEN SCHWENNINGEN
DE	BRANDENBURG AN DER HAVEL
DE	DÜREN STADT
DE	KEMPTEN ALLGÄU
DK	ÅRHUS
DK	ODENSE
DK	AALBORG
DK	KØBENHAVN
EE	TARTU
EE	NARVA
EE	TALLINN
EL	THESSALONIKI
EL	PÁTRA
EL	IRAKLEIO
EL	LARISA
EL	VOLOS
EL	IOANNINA
EL	KAVALA
EL	KALAMATA
EL	CHANIA
EL	XANTHI
EL	KATERINI
EL	SERRES
EL	TRIKALA
EL	ATHINA
ES	LAS PALMAS
ES	PALMA DE MALLORCA
ES	SANTIAGO DE COMPOSTELA
ES	TALAVERA DE LA REINA
ES	SANTA CRUZ DE TENERIFE
ES	BARCELONA
ES	VALENCIA
ES	SEVILLA
ES	ZARAGOZA
ES	MÁLAGA
ES	MURCIA



<b>Country code</b>	<b>City name</b>
ES	VALLADOLID
ES	JEREZ DE LA FRONTERA
ES	CHICLANA DE LA FRONTERA
ES	VITORIA GASTEIZ
ES	OVIEDO
ES	PAMPLONA IRUÑA
ES	SANTANDER
ES	TOLEDO
ES	BADAJOZ
ES	LOGROÑO
ES	BILBAO
ES	CÓRDOBA
ES	ALICANTE ALACANT
ES	VIGO
ES	GIJÓN
ES	PUERTO DE LA CRUZ
ES	REUS
ES	LUGO
ES	GIRONA
ES	TORREVIEJA
ES	CÁCERES
ES	SANLÚCAR DE BARRAMEDA
ES	AVILÉS
ES	CIUDAD REAL
ES	PALENCIA
ES	FERROL
ES	PONTEVEDRA
ES	CEUTA
ES	GANDIA
ES	GUADALAJARA
ES	MANRESA
ES	BENIDORM
ES	MELILLA
ES	PONFERRADA
ES	ZAMORA
ES	SAN SEBASTIÁN DONOSTIA
ES	SANTA LUCÍA DE TIRAJANA
ES	IRUN
ES	ARRECIFE
ES	ELDA
ES	GRANADA
ES	ELCHE ELX



<b>Country code</b>	<b>City name</b>
ES	CARTAGENA
ES	ALMERÍA
ES	BURGOS
ES	SALAMANCA
ES	ALBACETE
ES	HUELVA
ES	CÁDIZ
ES	TARRAGONA
ES	LEÓN
ES	JAÉN
ES	LLEIDA
ES	OURENSE
ES	ALGECIRAS
ES	MARBELLA
ES	ALCOY
ES	ÁVILA
ES	CUENCA
ES	EIVISSA
ES	LINARES
ES	LORCA
ES	MÉRIDA
ES	SAGUNTO
ES	IGUALADA
ES	CASTELLÓN DE LA PLANA CASTELLÓ DE LA PLANA
ES	A CORUÑA
ES	PUERTO DE SANTA MARÍA EL
ES	LA LÍNEA DE LA CONCEPCIÓN
ES	MADRID
FI	TAMPERE TAMMERFORS
FI	TURKU ÅBO
FI	OULU ULEÅBORG
FI	LAHTI LAHTIS
FI	KUOPIO
FI	JYVÄSKYLÄ
FI	HELSINKI
FR	PARIS
FR	LYON
FR	TOULOUSE
FR	BORDEAUX
FR	NANTES
FR	LILLE
FR	MONTPELLIER



Country code	City name
FR	SAINT ETIENNE
FR	LE HAVRE
FR	RENNES
FR	AMIENS
FR	NANCY
FR	METZ
FR	REIMS
FR	ORLEANS
FR	DIJON
FR	POITIERS
FR	CLERMONT FERRAND
FR	CAEN
FR	LIMOGES
FR	BESANCON
FR	GRENOBLE
FR	TOULON
FR	VALENCIENNES
FR	TOURS
FR	ANGERS
FR	BREST
FR	LE MANS
FR	AVIGNON
FR	MULHOUSE
FR	DUNKERQUE
FR	PERPIGNAN
FR	NIMES
FR	PAU
FR	BAYONNE
FR	LORIENT
FR	MONTBELIARD
FR	TROYES
FR	SAINT NAZAIRE
FR	LA ROCHELLE
FR	ANGOULEME
FR	BOULOGNE SUR MER
FR	CHAMBERY
FR	CHALON SUR SAONE
FR	CHARTRES
FR	NIORT
FR	CALAIS
FR	BEZIERS
FR	ARRAS



<b>Country code</b>	<b>City name</b>
FR	BOURGES
FR	SAINT BRIEUC
FR	QUIMPER
FR	VANNES
FR	CHERBOURG EN COTENTIN
FR	TARBES
FR	BELFORT
FR	ROANNE
FR	BEAUVAIS
FR	CREIL
FR	EVREUX
FR	CHATEAUROUX
FR	BRIVE LA GAILLARDE
FR	ALBI
FR	FREJUS
FR	CHALONS EN CHAMPAGNE
FR	NICE
FR	LENS
FR	HENIN BEAUMONT
FR	DOUAI
FR	VALENCE
FR	ROUEN
FR	MELUN
FR	MARTIGUES
FR	CHARLEVILLE MEZIERES
FR	COLMAR
FR	CANNES
FR	STRASBOURG
FR	AJACCIO
FR	ANNEMASSE
FR	ANNECY
FR	COMPIEGNE
FR	SAINT QUENTIN
FR	MARSEILLE
HR	RIJEKA
HR	OSIJEK
HR	SPLIT
HR	SLAVONSKI BROD
HR	ZADAR
HR	PULA - POLA
HR	GRAD ZAGREB
HU	BUDAPEST



<b>Country code</b>	<b>City name</b>
HU	MISKOLC
HU	NYÍREGYHÁZA
HU	PÉCS
HU	DEBRECEN
HU	SZEGED
HU	GYŐR
HU	KECSKEMÉT
HU	SZÉKESFEHÉRVÁR
HU	SZOMBATHELY
HU	SZOLNOK
HU	TATABÁNYA
HU	VESZPRÉM
HU	BÉKÉSCSABA
HU	KAPOSVÁR
HU	TEGER
HU	DUNAÚJVÁROS
HU	ZALAEGERSZEG
HU	SOPRON
IE	CORK
IE	LIMERICK
IE	GALWAY
IE	WATERFORD
IE	DUBLIN
IS	REYKJAVIK
IT	LA SPEZIA
IT	REGGIO DI CALABRIA
IT	ROMA
IT	MILANO
IT	NAPOLI
IT	TORINO
IT	PALERMO
IT	GENOVA
IT	FIRENZE
IT	BARI
IT	BOLOGNA
IT	CATANIA
IT	VENEZIA
IT	VERONA
IT	CREMONA
IT	TRENTO
IT	TRIESTE
IT	PERUGIA



Country code	City name
IT	ANCONA
IT	PESCARA
IT	CAMPOBASSO
IT	CASERTA
IT	TARANTO
IT	POTENZA
IT	CATANZARO
IT	BUSTO ARSIZIO
IT	SASSARI
IT	CAGLIARI
IT	PADOVA
IT	BRESCIA
IT	MODENA
IT	FOGGIA
IT	SALERNO
IT	PIACENZA
IT	BOLZANO
IT	UDINE
IT	LECCE
IT	BARLETTA
IT	PESARO
IT	COMO
IT	PISA
IT	TREVISO
IT	VARESE
IT	ASTI
IT	PAVIA
IT	MASSA
IT	COSENZA
IT	SAVONA
IT	MATERA
IT	ACIREALE
IT	AVELLINO
IT	PORDENONE
IT	LECCO
IT	ALTAMURA
IT	BATTIPAGLIA
IT	BISCEGLIE
IT	CARPI
IT	CERIGNOLA
IT	GALLARATE
IT	GELA



Country code	City name
IT	SASSUOLO
IT	MESSINA
IT	PRATO
IT	PARMA
IT	LIVORNO
IT	L AQUILA
IT	RAVENNA
IT	FERRARA
IT	RIMINI
IT	SIRACUSA
IT	BERGAMO
IT	FORLÌ
IT	LATINA
IT	VICENZA
IT	TERNI
IT	NOVARA
IT	ALESSANDRIA
IT	AREZZO
IT	GROSSETO
IT	BRINDISI
IT	TRAPANI
IT	RAGUSA
IT	ANDRIA
IT	TRANI
IT	MANFREDONIA
IT	SAN SEVERO
IT	REGGIO NELL EMILIA
LT	KAUNAS
LT	PANEVĖŽYS
LT	ALYTUS
LT	KLAIPĖDA
LT	ŠIAULIAI
LT	VILNIUS
LU	LUXEMBOURG
LV	JELGAVA
LV	LIEPĀJA
LV	DAUGAVPILS
LV	RIGA
ME	PODGORICA
MK	BITOLA
MK	TETOVO
MK	PRILEP



Country code	City name
MK	SKOPJE
MT	VALLETTA
NL	GREATER SOEST
NL	ALPHEN AAN DEN RIJN
NL	S HERTOGENBOSCH
NL	BERGEN OP ZOOM
NL	GREATER SITTARD GELEEN
NL	GREATER ROTTERDAM
NL	GREATER UTRECHT
NL	GREATER EINDHOVEN
NL	GREATER HEERLEN
NL	TILBURG
NL	GRONINGEN
NL	ENSCHEDE
NL	ARNHEM
NL	BREDA
NL	NIJMEGEN
NL	APELDOORN
NL	LEEUWARDEN
NL	GREATER MIDDELBURG
NL	DELFT
NL	HILVERSUM
NL	ROOSENDAAL
NL	GREATER ALKMAAR
NL	KATWIJK
NL	GOUDA
NL	GREATER LEIDEN
NL	AMERSFOORT
NL	MAASTRICHT
NL	DORDRECHT
NL	GREATER EDE
NL	ZWOLLE
NL	DEVENTER
NL	VENLO
NL	ALMELO
NL	LELYSTAD
NL	OSS
NL	ASSEN
NL	VEENENDAAL
NL	GREATER AMSTERDAM
NL	GREATER S GRAVENHAGE
NO	BERGEN



<b>Country code</b>	<b>City name</b>
NO	TRONDHEIM
NO	STAVANGER
NO	KRISTIANSAND
NO	TROMSØ
NO	OSLO
PL	STAŁOWA WOLA
PL	TOMASZÓW MAZOWIECKI
PL	GORZÓW WIELKOPOLSKI
PL	WARSZAWA
PL	ŁÓDŹ
PL	KRAKÓW
PL	WROCŁAW
PL	POZNAŃ
PL	GDAŃSK
PL	SZCZECIN
PL	BYDGOSZCZ
PL	LUBLIN
PL	KATOWICE
PL	BIAŁYSTOK
PL	KIELCE
PL	TORUŃ
PL	OLSZTYN
PL	RZESZÓW
PL	OPOLE
PL	KONIN
PL	JELENIA GÓRA
PL	NOWY SĄCZ
PL	SUWAŁKI
PL	CZĘSTOCHOWA
PL	RADOM
PL	PŁOCK
PL	KALISZ
PL	KOSZALIN
PL	SŁUPSK
PL	JASTRZĘBIE-ZDRÓJ
PL	SIEDLCE
PL	OSTRÓW WIELKOPOLSKI
PL	LUBIN
PL	GNIEZNO
PL	PIŁA
PL	INOWROCŁAW
PL	STARGARD SZCZECIŃSKI



Country code	City name
PL	PRZEMYŚL
PL	ZAMOŚĆ
PL	CHEŁM
PL	PABIANICE
PL	GŁOGÓW
PL	ŁOMŻA
PL	LESZNO
PL	ŚWIDNICA
PL	TCZEW
PL	EŁK
PL	BIELSKO-BIAŁA
PL	RYBNIK
PL	WAŁBRZYCH
PL	ELBLĄG
PL	WŁOCŁAWEK
PL	TARNÓW
PL	LEGNICA
PL	GRUDZIĄDZ
PL	MIELEC
PL	BELCHATOW
PL	ZIELONA GÓRA
PL	PIOTRKÓW TRYBUNALSKI
PL	OSTROWIEC ŚWIĘTOKRZYSKI
PT	PORTO
PT	BRAGA
PT	FUNCHAL
PT	COIMBRA
PT	AVEIRO
PT	FARO
PT	VISEU
PT	GUIMARÃES
PT	VIANA DO CASTELO
PT	PÓVOA DE VARZIM
PT	LISBOA
RO	CLUJ-NAPOCA
RO	ALBA IULIA
RO	TIMIȘOARA
RO	CRAIOVA
RO	BRĂILA
RO	ORADEA
RO	BACĂU
RO	ARAD



<b>Country code</b>	<b>City name</b>
RO	SIBIU
RO	PIATRA NEAMȚ
RO	CĂLĂRAȘI
RO	GIURGIU
RO	SATU MARE
RO	FOCŞANI
RO	TÂRGU MUREŞ
RO	TULCEA
RO	SLATINA
RO	TÂRGOVIŞTE
RO	BÂRLAD
RO	ROMAN
RO	BISTRIȚA
RO	CONSTANȚA
RO	IAŞI
RO	GALAȚI
RO	BRAŞOV
RO	PLOIEŞTI
RO	PITEŞTI
RO	BUZĂU
RO	BOTOŞANI
RO	SUCEAVA
RO	SFANTU GHEORGHE
RO	DEVA
RO	HUNEDOARA
RO	BAIA MARE
RO	DROBETA TURNU SEVERIN
RO	TÂRGU JIU
RO	RÂMNICU VÂLCEA
RO	BUCUREŞTI
RS	NOVI SAD
RS	NIS
RS	KRAGUJEVAC
RS	SUBOTICA
RS	NOVI PAZAR
RS	ZRENJANIN
RS	KRALJEVO
RS	CACAK
RS	KRUSEVAC
RS	LESKOVAC
RS	VALJEVO
RS	VRANJE



<b>Country code</b>	<b>City name</b>
RS	SMEDEREVO
RS	BEOGRAD
SE	GÖTEBORG
SE	UMEÅ
SE	UPPSALA
SE	VÄSTERÅS
SE	NORRKÖPING
SE	HELSINGBORG
SE	STOCKHOLM
SE	MALMÖ
SE	JÖNKÖPING
SE	LINKÖPING
SE	ÖREBRO
SE	BORÅS
SI	MARIBOR
SI	LJUBLJANA
SK	NITRA
SK	PREŠOV
SK	ŽILINA
SK	TRNAVA
SK	TRENČÍN
SK	MARTIN
SK	BRATISLAVA
SK	BANSKÁ BYSTRICA
SK	KOŠICE
TR	ADANA
TR	ANTALYA
TR	BALIKESIR
TR	DENIZLI
TR	DIYARBAKIR
TR	EDIRNE
TR	ERZURUM
TR	GAZIANTEP
TR	ANTAKYA
TR	IZMIR
TR	KARS
TR	KASTAMONU
TR	KAYSERI
TR	KONYA
TR	MALATYA
TR	NEVSEHIR
TR	SAMSUN



<b>Country code</b>	<b>City name</b>
TR	SIIRT
TR	TRABZON
TR	VAN
TR	ZONGULDAK
TR	ESKISEHIR
TR	SANLIURFA
TR	KAHRAMANMARAS
TR	BATMAN
TR	SIVAS
TR	ELAZIG
TR	ISPARTA
TR	CORUM
TR	OSMANIYE
TR	AKSARAY
TR	AYDIN
TR	SIVEREK
TR	ORDU
TR	AFYONKARAHISAR
TR	NIGDE
TR	USAK
TR	AGRI
TR	KARAMAN
TR	YUMURTALIK
TR	RIZE
TR	ERGANI
TR	KUTAHYA
TR	KADIRLI
TR	KARABUK
TR	CANAKKALE
TR	AKCAKALE
TR	ERCIS
TR	EREGLI
TR	ADIYAMAN
TR	VIRANSEHIR
TR	FETHIYE
TR	CEYLANPINAR
TR	TOKAT
TR	PATNOS
TR	ODEMIS
TR	BOLU
TR	BANDIRMA
TR	MUS



Country code	City name
TR	ELBISTAN
TR	NIZIP
TR	SURUC
TR	SALIHLI
TR	KILIS
TR	KIZILTEPE
TR	MIDYAT
TR	CIZRE
TR	CANKIRI
TR	BINGOL
TR	AKSEHIR
TR	POLATLI
TR	MANAVGAT
TR	YOZGAT
TR	ALASEHIR
TR	ISTANBUL
TR	ANKARA
UK	BLACKBURN WITH DARWEN
UK	BATH AND NORTH EAST SOMERSET
UK	KINGSTON UPON HULL
UK	GREAT YARMOUTH
UK	DUNDEE CITY
UK	BIRMINGHAM
UK	LEEDS
UK	SHEFFIELD
UK	WREXHAM
UK	EAST STAFFORDSHIRE
UK	NOTTINGHAM
UK	GUILDFORD
UK	DONCASTER
UK	MILTON KEYNES
UK	CARLISLE
UK	CRAWLEY
UK	LONDON
UK	STOKE ON TRENT
UK	BRIGHTON AND HOVE
UK	NORTH EAST LINCOLNSHIRE
UK	BASINGSTOKE AND DEANE
UK	TELFORD AND WREKIN
UK	TUNBRIDGE WELLS
UK	CHESHIRE WEST AND CHESTER
UK	GLASGOW



Country code	City name
UK	LIVERPOOL
UK	EDINBURGH
UK	MANCHESTER
UK	CARDIFF
UK	BRISTOL
UK	BELFAST
UK	LEICESTER
UK	DERRY STRABANE
UK	ABERDEEN
UK	CAMBRIDGE
UK	EXETER
UK	LINCOLN
UK	STEVENAGE
UK	PORTSMOUTH
UK	WORCESTER
UK	COVENTRY
UK	BRACKNELL FOREST
UK	NEWCASTLE UPON TYNE
UK	THANET
UK	WAVENEY
UK	ASHFORD
UK	DARLINGTON
UK	WORTHING
UK	MANSFIELD
UK	CHESTERFIELD
UK	BURNLEY
UK	EASTBOURNE
UK	HASTINGS
UK	REDDITCH
UK	SUNDERLAND
UK	MEDWAY
UK	PLYMOUTH
UK	SWANSEA
UK	DERBY
UK	SOUTHAMPTON
UK	NORTHAMPTON
UK	WARRINGTON
UK	LUTON
UK	YORK
UK	SWINDON
UK	BOURNEMOUTH
UK	WYCOMBE



<b>Country code</b>	<b>City name</b>
UK	PETERBOROUGH
UK	COLCHESTER
UK	BEDFORD
UK	FALKIRK
UK	READING
UK	BLACKPOOL
UK	MAIDSTONE
UK	DACORUM
UK	NEWPORT
UK	MIDDLESBROUGH
UK	OXFORD
UK	TORBAY
UK	PRESTON
UK	NORWICH
UK	IPSWICH
UK	CHELTENHAM
UK	GLOUCESTER
UK	RUSHMOOR
UK	CORBY
UK	KETTERING
UK	BOGNORREGIS
UK	LITTLEHAMPTON
UK	TAUNTON
UK	NEWBURY
UK	AYLESBURY
UK	HEREFORD
UK	KIDDERMINSTER
UK	SHREWSBURY
UK	STAFFORD
UK	SCUNTHORPE
UK	AYR
XK	MITROVICE
XK	PRISTINA
XK	PRIZEN