



The influence of Level of Details (LoD0-2) and buffer sizes on parameter effectiveness for fine dust distribution modelling

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Original paper: <https://www.isprs-ann-photogramm-remote-sens-spatial-inf-sci.net/IV-4-W6/41/2018/isprs-annals-IV-4-W6-41-2018.pdf>

Fine dust – an urban system failure



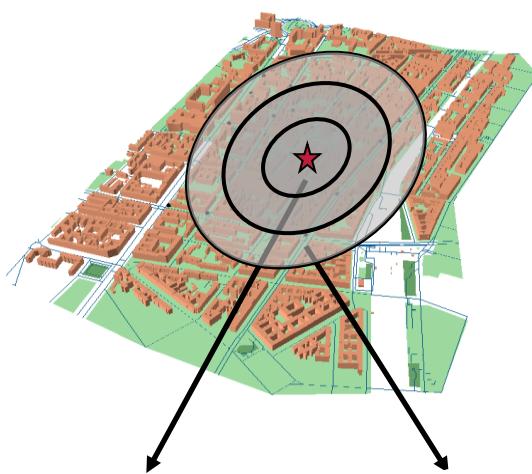
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Foto: imago/Christian Mang

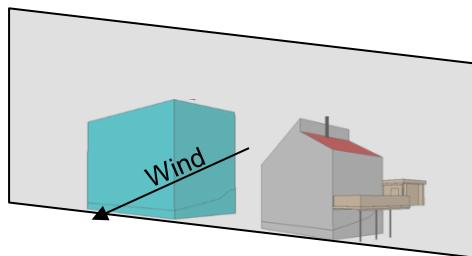
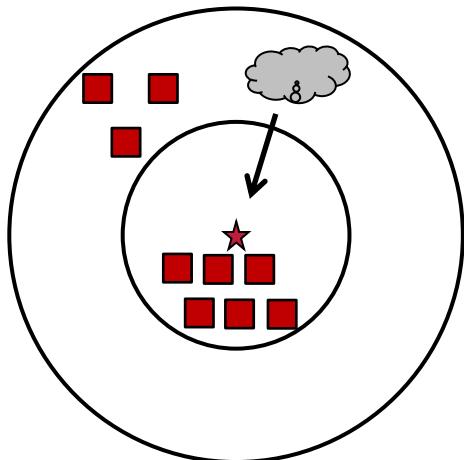
- Traffic generated fine and Ultra Fine Particles (UFP, 10nm - 420nm) significantly **increase health risks in urban areas** (WHO, 2013)
- brought to the focus by the Volkswagen emissions scandal (germ.: Dieselaffäre)
- **Fast** and preferably accurate **now-cast of UFP** is needed to plan, optimize, and evaluate **counter measures**
 - short term: smart traffic management system (diversion)
 - mid term: greenery
 - long term: urban planning
- Land Use Regression model (LUR)
 - geostatistical method
 - parameter extensive
 - easy to use
 - supports fast to instantaneous now-cast
 - useful for the implementation as a service

Modelling the urban fine dust distribution – the LUR approach



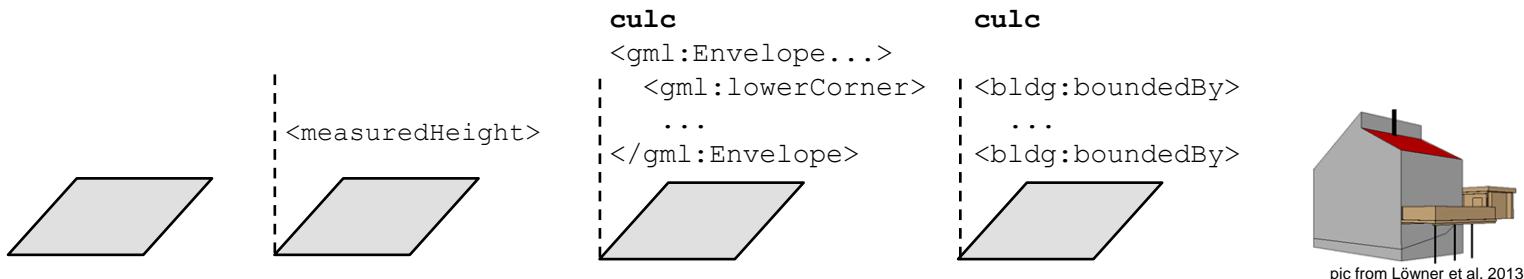
- Measuring the pollution concentration at a certain location
- Analyzing spatial parameters within certain buffers
- Determine the correlation between measured values and the parameters
- Performing an enhanced multilinear regression analysis (Henderson 2007, Ghassoun & Löwner 2017)
- We divide parameters representing
 - Production
 - Deposition
 - Dilution
 - the local wind field and therefore
 - the urban morphology

Estimated effects of LoD and buffer sizes on parameter correlation



- Influence of **buffer sizes** on the correlation between parameter and measured values
 - **averaging effect** of the calculated parameter in large buffers
 - **overestimation of effects** (production) in close vicinity in small buffers
- Influence of **LoD** on the correlation between parameter and measured values
 - **change of dimension** when defining the parameter (from surface to volume to more volume)
 - **different volume calculations** because of f.i. building installations
 - **different shape** (including roof shape) of projected surface f.i. shielding the wind

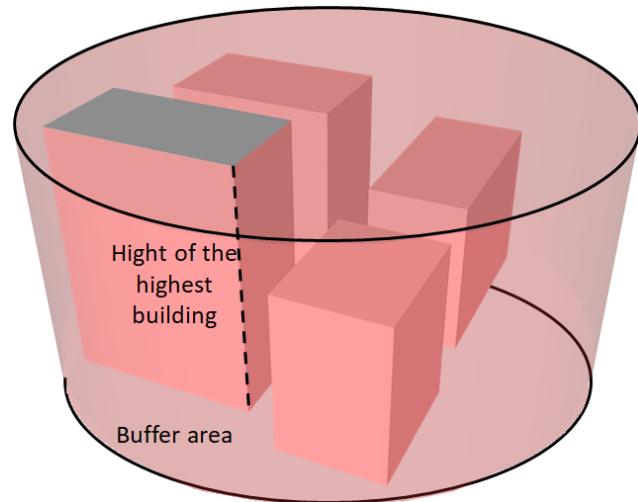
Parameters and buffer widths



	LoD0	LoD1 extruded using <measuredHeight>	LoD1 extruded using LoD1 envelope height	LoD1 extruded using LoD2 boundarySurfaces height	LoD2
averaged heights of buildings		X	X	X	
Height-width ratio		X	X	X	
Porosity	X	X	X	X	
Frontal Area Index	X	X	X	X	X
Building surfaces		X	X	X	X

Buffers from 50m and 100m, 200m,...1000m were analyzed, except for the frontal area index analysis (50m and 100m – 500m)

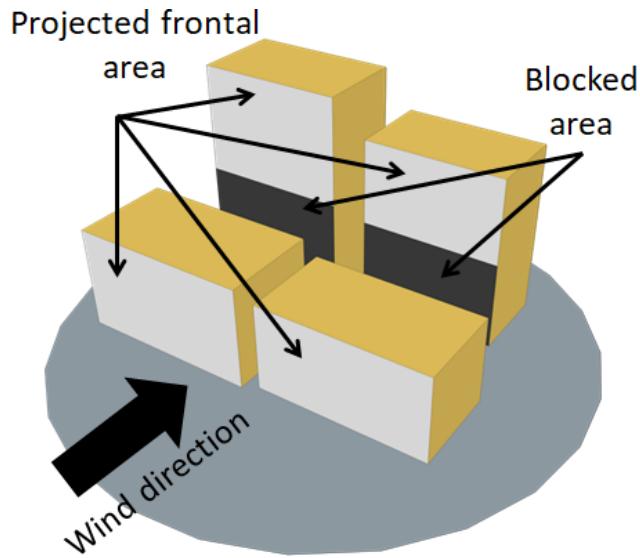
Porosity in 3D



- General information about the **ventilation capacity** of an area
 - descriptive parameter for the dilution capacity
 - **high values should result in lower UFP values**

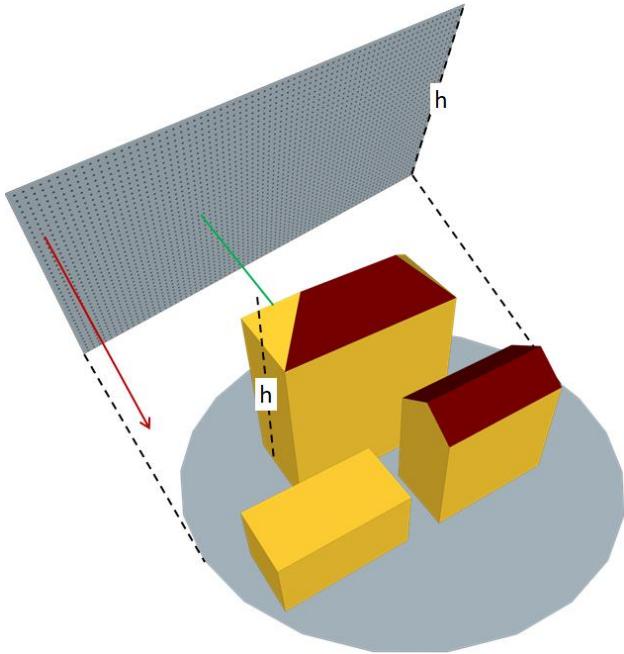
1. Identify **every building** within a certain buffer
2. Get the height of the **highest building** in the buffer (according to different height information)
3. Calculate the **volume of the 3D-buffer**
4. **Sum all the volumes of all the buildings** within the buffer according to different height information
5. Calculate porosity

Frontal Area Index from LoD1



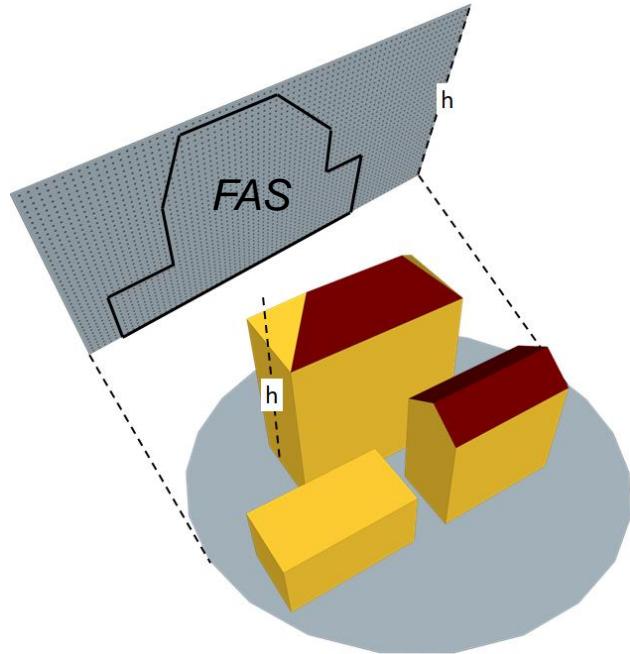
- Descriptive concept to describe the **effect of buildings shielding the wind**. The higher it is, the less the wind is expected to pass through an area.
 - descriptive parameters for the dilution capacity
 - low values should result in higher UFP values
- Area of the **buffer** divided by the **total blocking area perpendicular to the wind direction** of all buildings within a buffer (Frontal Area Surface)
- Represents the deteriorating ventilation effect
- Again, different height information was applied
- **For LoD1:** Calculated with the Urban Roughness Toolbox (ArcGIS 10.5) from René Burghard

Frontal Area Index from LoD2



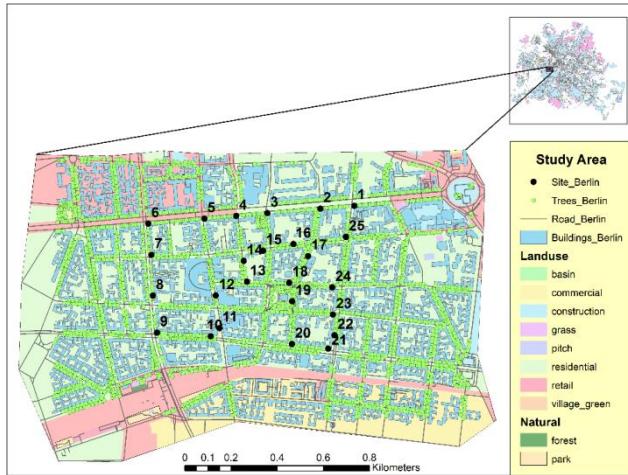
- Total blocking area perpendicular to the wind direction of all buildings within a buffer (Frontal area Surface) is expected to be represented in a more realistic way
 - Correlation with UFP values should be better
1. Select all (LoD2) buildings in a corresponding buffer
 2. Get **all surfaces** of all buildings within the buffer and cut them **into triangles** applying a fan triangulation
 3. Create a **point array** with 1 m resolution, arranged perpendicular to the wind direction filling an area of the width of the buffer time the height of the highest building
 4. Complete **every point** in the point array with a **direction vector** towards the wind direction
 5. Compute revealing **half ray line against every triangle** applying the Möller-Trumbore intersection algorithm (Möller and Trumbore, 1997).

Frontal Area Index from LoD2



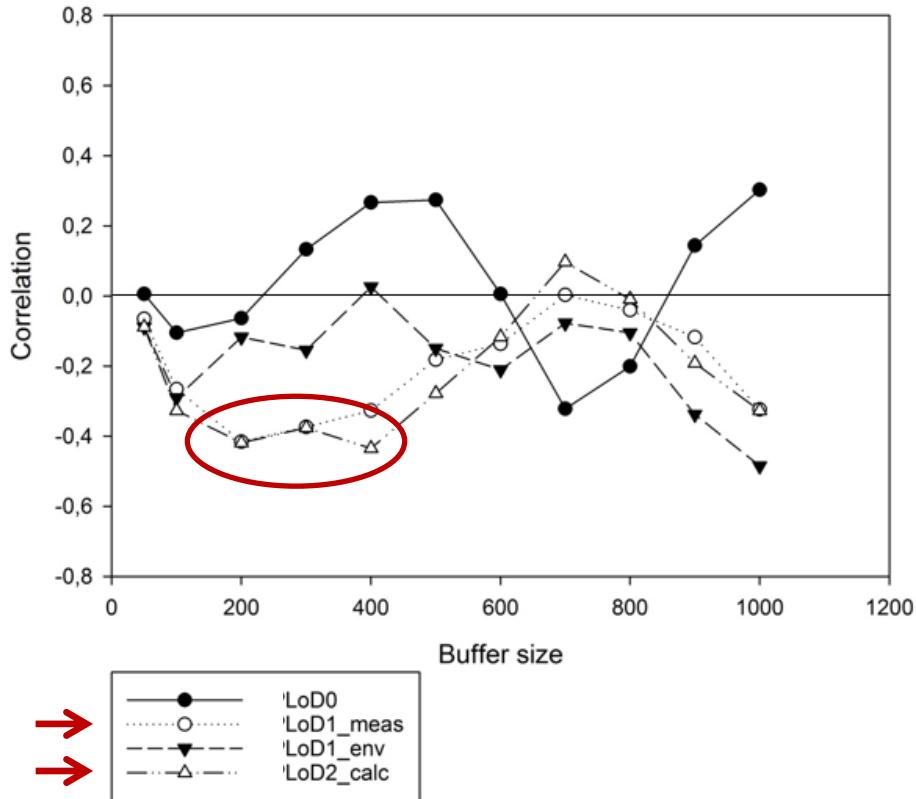
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Research area and data



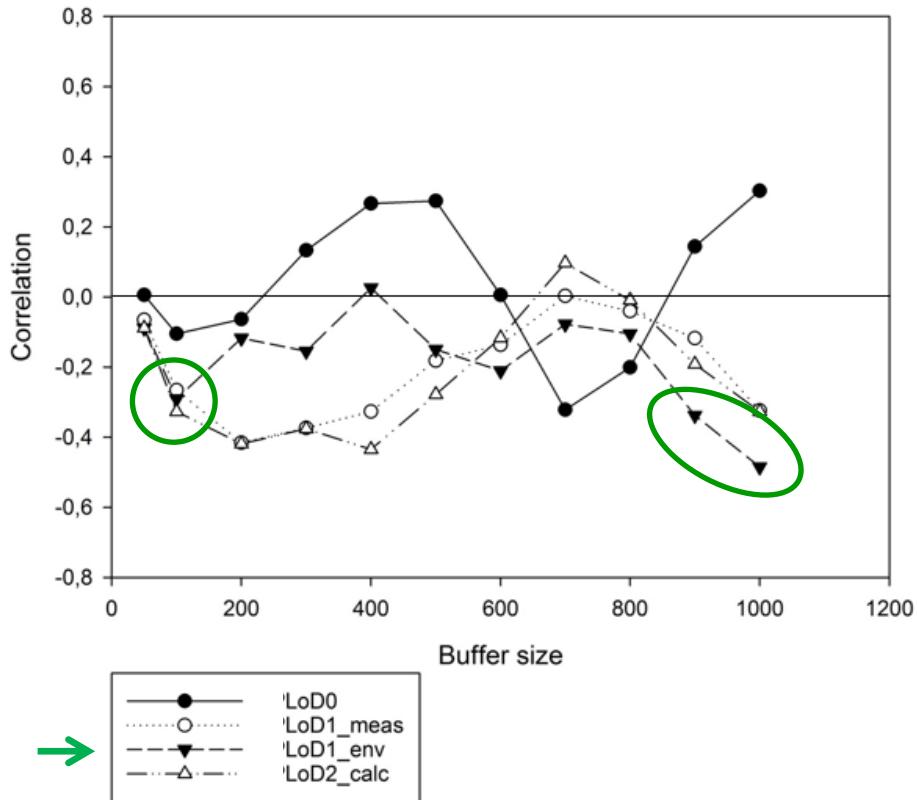
- Research Area: **1x2 km in Berlin, Germany**
- **25 measuring points** for Total number concentrations (TNC) with a hand-held particle counter device (TSI 3007) (size from 10 to about 1000 nm, range 0 to 100,000 particle per cm³).
- **CityGML data (*LoD1 and LoD2*)** from the open data portal of the city of Berlin (<https://fbinter.stadt-berlin.de/fb/index.jsp>)
- ***LoD0* representation was extracted from *LoD1* data**
- Data extraction performed **directly from CityGML files on pure Python** scripting with no interconnected database or spatial packages
- **Parameter calculation** was performed either in ArcGIS for LoD0 and LoD1 representations or
- with self-programmed **Python Scripts** for LoD1 and LoD2 representations

Results on the Porosity



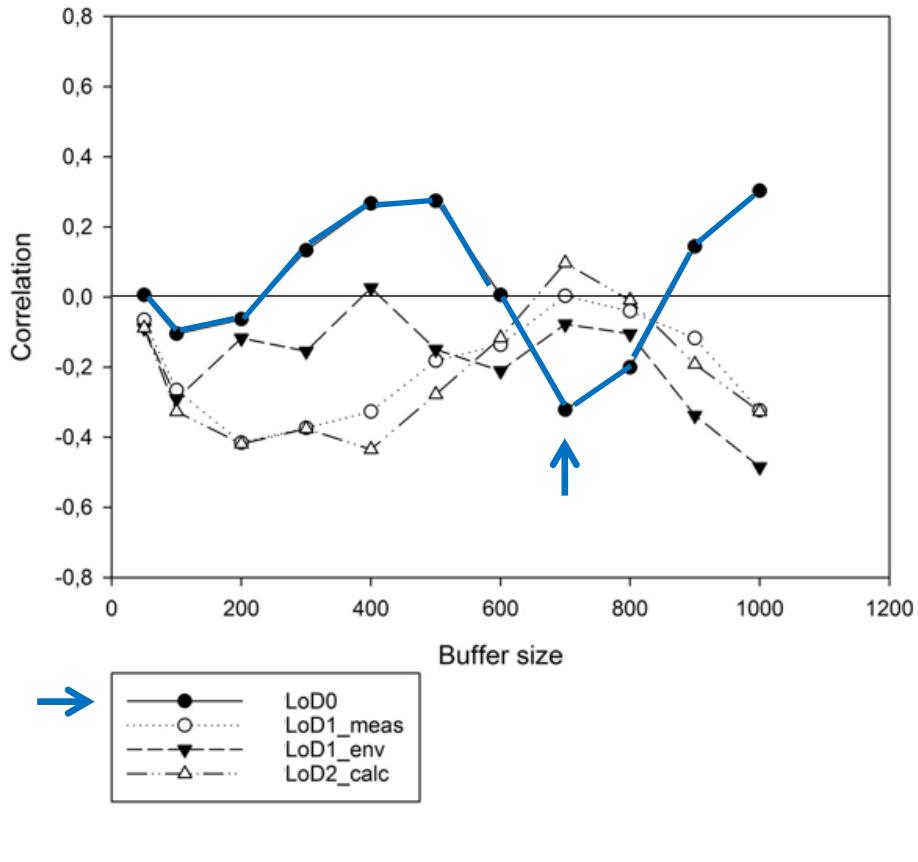
- ‘Porosity’: a **value of how penetrable** an area is for the wind
- Expected: negative correlation with high fine dust values
- Best performers: **LoD2_culc** and **LoD1_meas** showing parallel behaviour with an high negative correlation of approx. *0.435 in 200m-400m buffers*

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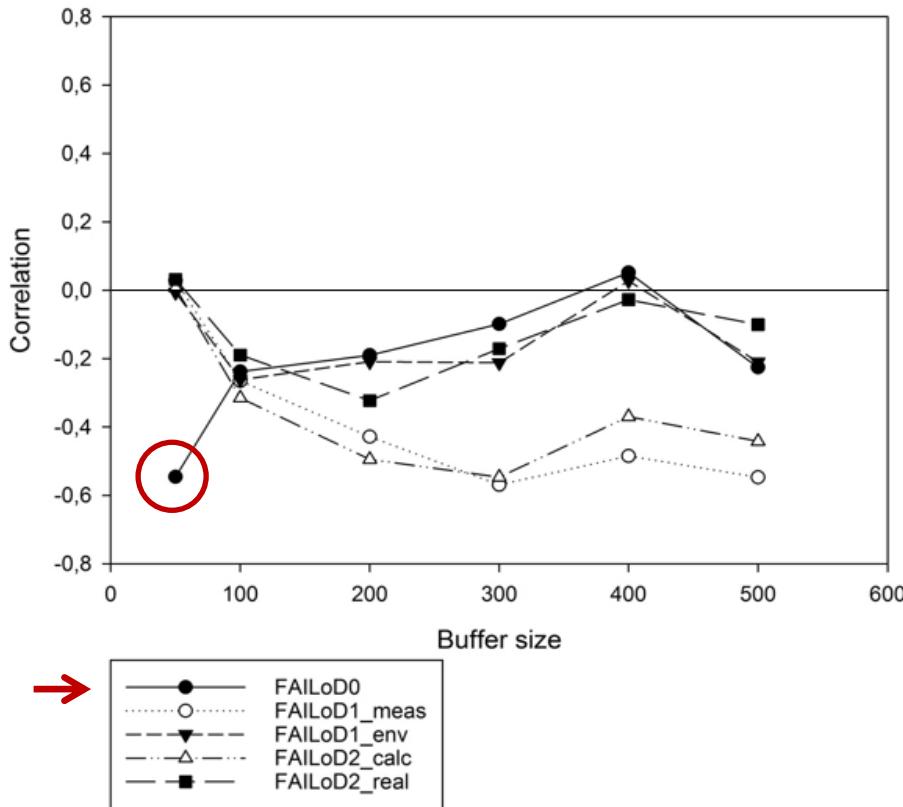
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- **LoD1_env** shows expected high negative correlations only in a buffer of 100m, 900m, and 1000m

Results on the Porosity



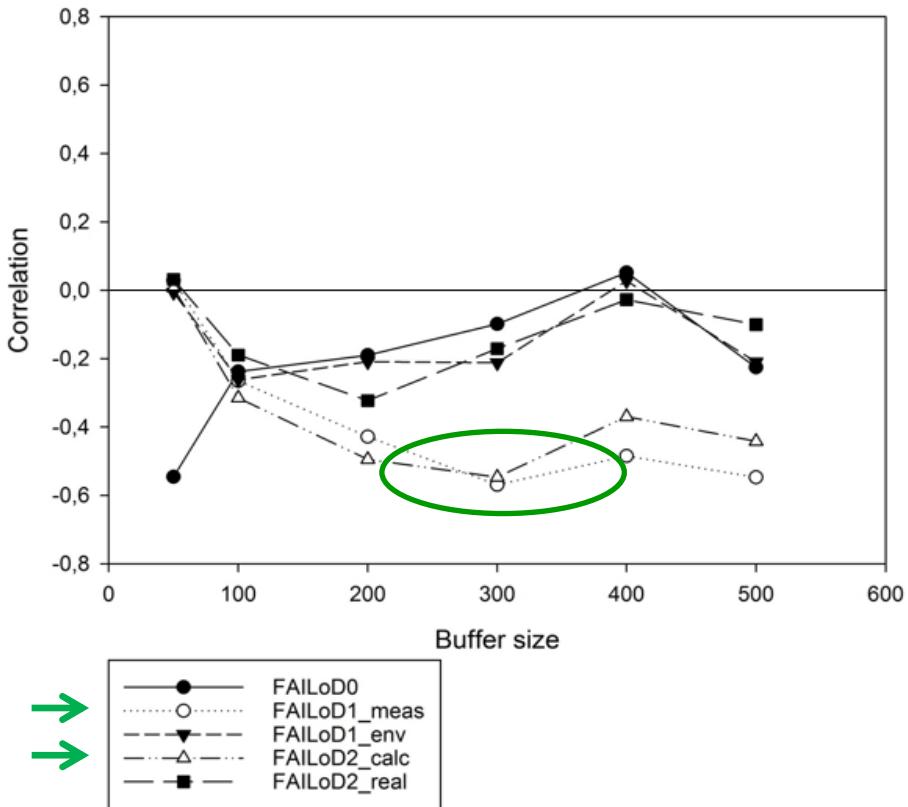
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- **LoD1_env** shows expected high negative correlations only in a buffer of 100m, 900m, and 1000m
- **LoD0** seems not to perform, even correlation in buffer of 700m is not really high

Results on the Frontal Area Index



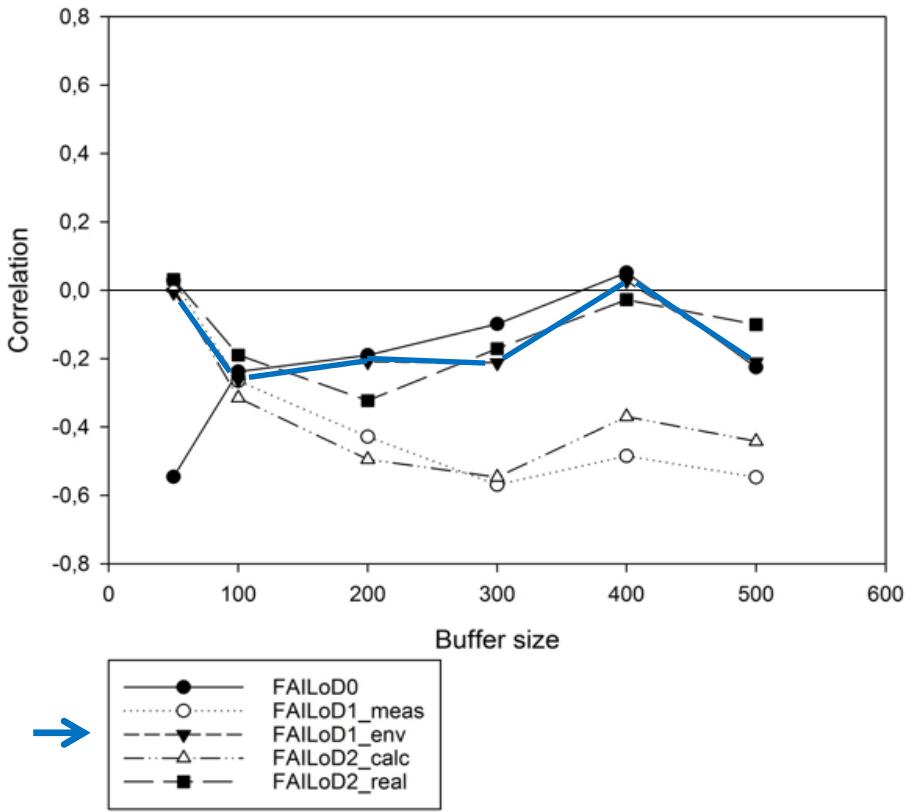
- ‘Frontal Area Index’: a **value of the deteriorating ventilation** effect through wind shielding surfaces
- Expected: **negative correlation** with high fine dust values (because of using its reciprocal)
- Best performer: **LoD0** showing highest negative correlation in a 50m buffer but bad correlations in other buffers

Results on the Frontal Area Index



- ‘Frontal Area Index’: a **value of the deteriorating ventilation** effect through wind shielding surfaces
- Expected: **negative correlation** with high fine dust values (because of using its reciprocal)
- Best performer: **LoD0** showing highest negative correlation in a 50m buffer but bad correlations in other buffers
- Good performers: **LoD2_calc** and **LoD1_meas** again showing parallel behaviour with an high negative correlation of approx. 0.56 in a 300m buffer

Results on the Frontal Area Index



- ‘Frontal Area Index’: a **value of the deteriorating ventilation** effect through wind shielding surfaces
- Expected: **negative correlation** with high fine dust values (because of using its reciprocal)
- Best performer: **LoD0** showing highest negative correlation in a 50m buffer but bad correlations in other buffers
- Good performers: **LoD2_calc** and **LoD1_meas** again showing parallel behaviour with an high negative correlation of approx. 0.56 in a 300m buffer
- Non performer: **LoD1_env** shows weak correlation over all buffer sizes

Conclusion

- Influence of buffer sizes and different LoD representations on the correlation between parameter and measured values
- Target: We need the smallest buffer and the simplest data representation possible to fasten our modelbuilding in the context of Web-services
- Mainly focusing on parameters representing dilution processes
 - Porosity
 - Frontal area index
- LoD1 models extruded from LoD1 <measuredHeight> and LoD2 <bldg:boundedBy> calculation showed almost parallel behavior
- Best performers are found along all different LoD representation, except LoD1 extruded from LoD1 Envelope calculation
- General statements for best performing buffer sizes cannot be made
- Work should be extended in terms of different urban system (compare more cities)

Thank you for your attention...

Bedankt voor je aandacht!

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Get the paper:

<https://www.isprs-ann-photogramm-remote-sens-spatial-inf-sci.net/IV-4-W6/41/2018/isprs-annals-IV-4-W6-41-2018.pdf>

Related Topics:

Ghassoun, Y., Löwner, M.-O. (2017): Land use regression models for total particle number concentrations using 2D, 3D and semantic parameters, *Atmospheric Environment*. [PDF](#)

Ghassoun, Y. & Löwner, M.-O. (2017): Comparison of 2D & 3D parameter-based models in urban fine dust distribution modelling. In: A. Abdul-Rahman (ed.), *Advances in 3D Geoinformation, Lecture Notes in Geoinformation and Cartography*, DOI 10.1007/978-3-319-25691-7_13, 2017, 231-246. [PDF](#)

Löwner, M.-O., Gröger, G., Benner, J., Biljecki, F., and Nagel, C. (2016): Proposal for a new LoD and multi-representational concept for CityGML, *ISPRS Ann. Photogramm. Remote Sens. Spatial Inf. Sci.*, IV-2/W1, 3-12, doi:10.5194/isprs-annals-IV-2-W1-3-2016. [PDF](#)