



**Our air is making us sick.**  
Advanced analytics can help.

Ramboll Shair™



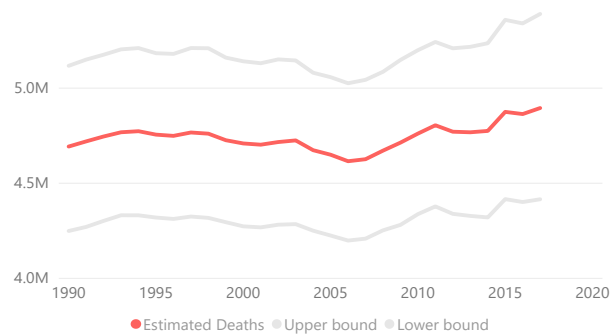
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# Choking on air, drowning in data

## 5 million lives lost

The simple act of breathing kills **5 million** people every year (Figure 1, IHME 2019). At the regional level, governments have made great strides toward reducing air pollution concentrations and their per-capita health effects. For example, the county of Los Angeles, California (USA) has seen significant decreases in smog in the last half decade. Air quality, however, is affected on a level far more local than the regional scale at which we often observe and regulate it. Recently published research has shown air pollution values **differing substantially even between neighboring city blocks** (EDF 2017), yet the global average distance between a population and its nearest fine particulate pollution monitor is relatively massive at **220 km** (Martin, et al. 2019). Recent efforts to fill in the particulate matter monitoring gaps with techniques like satellite sensing, chemical transport, and statistical modeling have been shown to overestimate or underestimate actual values by upwards of  $50 \mu\text{g}/\text{m}^3$  – **5 times the WHO recommended exposure level** of  $10 \mu\text{g}/\text{m}^3$  (Martin, et al. 2019). The Shair model is predicated on the belief that sensor data and advanced analytics can help.



**Figure 1:**

Global total all-cause deaths attributable to air pollution 1990-2017. Per-capita burden is decreasing, but, as population grows, so, too, does the total disease burden of air pollution (IHME 2019B).

## Sensor data overload

Local agencies, cities, and community have recently begun their own ef-

forts to **fill the data gaps between regulatory monitors** by setting up their own monitors and monitoring networks. These networks are largely comprised of **mid-to-low cost sensors** that can be easily mounted, configured, and deployed. A quick look at data from just a single manufacturer of low-cost sensor-based monitors (**Purple Air**) reveals roughly **5,500** of their units are in use outdoors within the continental United States. This manufacturer alone has nearly three times the presence of the United States Environmental Protection Agency — which operates roughly 2,000 various regulatory monitors across the United States— and sensor adoption is increasing at an exponential rate. Within the next few years, we expect to see tens of thousands around the world. If integrated properly into traditional air quality assessment methods, the large (and growing!) network of mid-to-low-cost sensors can dramatically reduce the average distance between a population and the nearest monitor. In effect, such integration can increase our understanding of hyper-local changes in exposure-relevant air quality at a global scale.

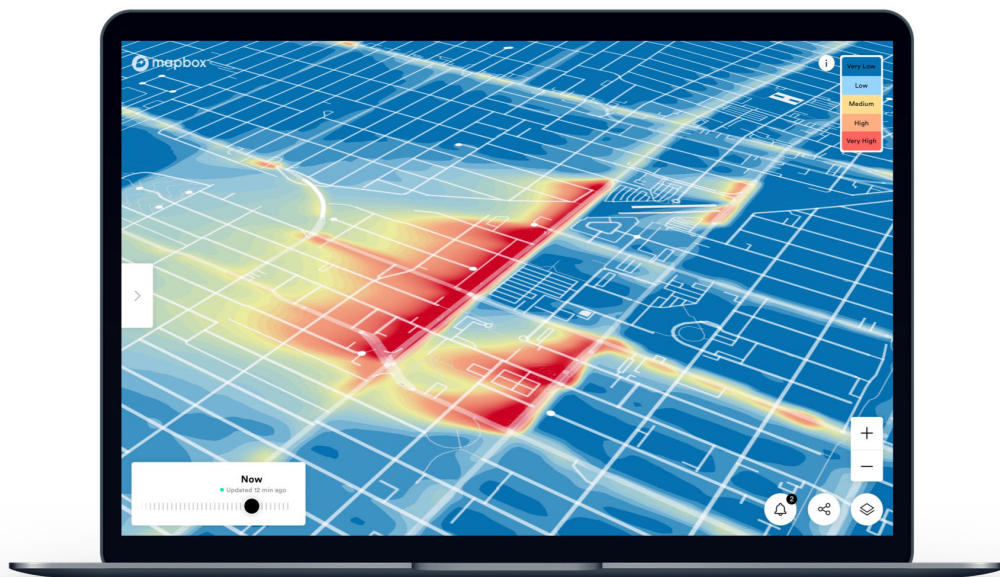
On their own, sensor data alone are poorly-suited for assigning hotspot **causation** to emissions sources and, in effect, understanding how hotspots can be prevented. This is especially true in air sheds with complex source profiles. To answer these questions, emissions inventories and physical and chemical transport modeling methods are needed. Similarly, we are unlikely to see a density of sensors that is so high that it replaces the need for filling data gaps with modeling or interpolation -- these data gaps will become smaller but not disappear.

### **Making sense of air quality data**

At Shair, we combine **sensor data** with tried and tested methods like **dispersion** and **weather modeling** to uncover geographic and temporal hotspots as well as the identities of **responsible emissions sources**. We

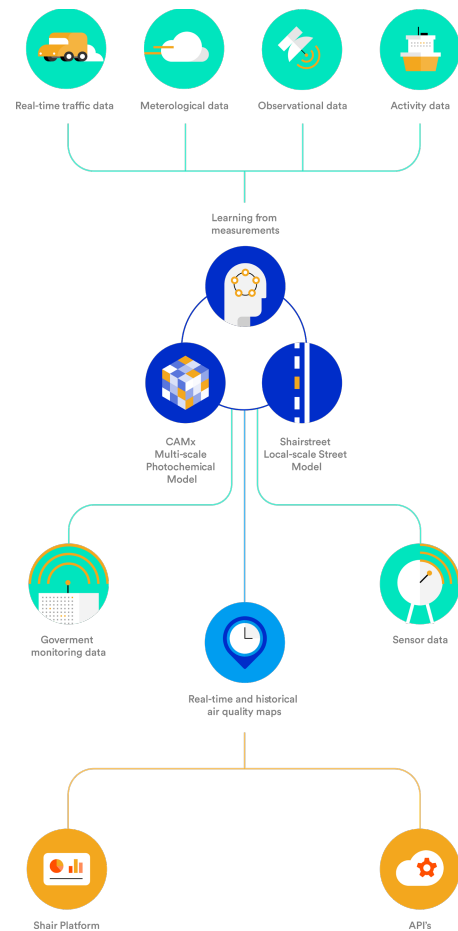
use street-level and regional physical and chemical transport models (CAMx) informed by traditional emissions inventories and novel realtime traffic estimation methods to estimate concentrations at a hyper-local 10 meter resolution. Furthermore, we carefully adjust these estimates to more accurately reflect measurements taken both by sensors-based and regulatory grade monitors. **Anchoring this model in measurement data** improves realtime temporal and spatial accuracy. It allows us to improve our street-level model and potentially identify emissions sources missing from local inventories.

The final result is a **near-realtime** look at the most significant sources of pollution affecting a population and how levels of that pollution vary around your city, **facilitating meaningful actions** for reducing exposures.

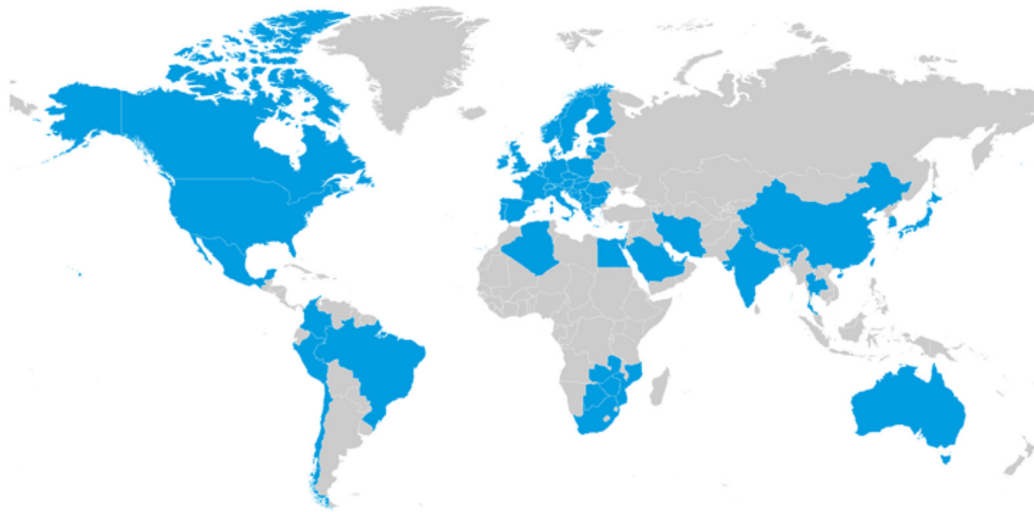


# How Shair works

The Shair team's air quality model (eponymously named "Shair") adds value to sensor network data by uncovering emission sources and their contributions. The model is **sensor-agnostic** and can ingest information from any air quality monitor or local emissions inventory. Shair operates in the **cloud** to translate large and unwieldy quantities of air quality data into the information needed to influence action and mitigation efforts by cities and air pollution control agencies. Shair outputs summary **datasets**, **maps**, and **visualizations** through a custom interface that can be integrated into web applications or data management systems. Shair is built on a backbone of the **CAMx** regional air quality model with a custom street-level urban pollution model called "Shairstreet." Near-realtime data are integrated into Shair (e.g., sensor data, reference data, meteorology, traffic data, and other activity data) to "**anchor**" the model to the **near-realtime measurements**. All of this is performed in the cloud on Google Cloud Platform (GCP) for increased speed and reliability.



**Figure 2:**  
The Shair architecture.



**Figure 3:**  
CAMx applications (blue) worldwide. (Source: [www.camx.com](http://www.camx.com))

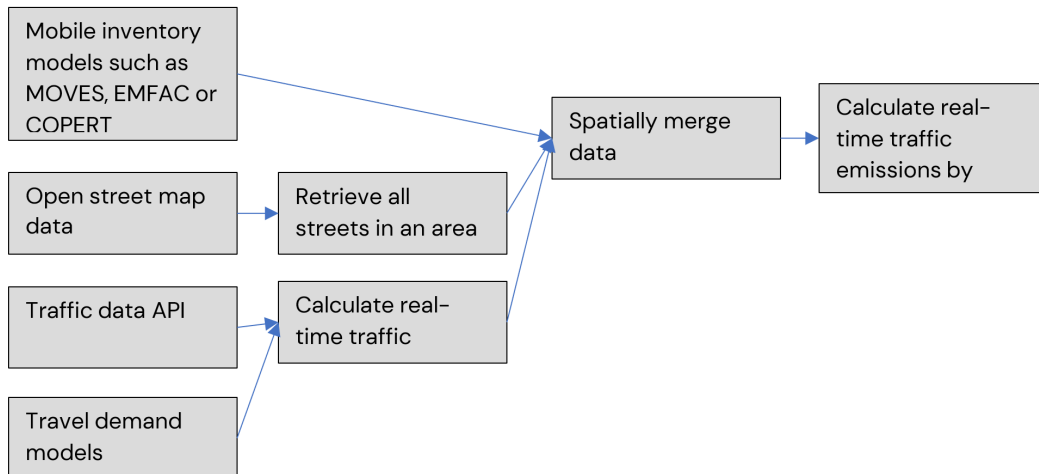
## CAMx

The Shair workflow begins with the Comprehensive Air Quality Model with extensions (CAMx), which is used to produce **hourly estimates** of local ground level **NO<sub>2</sub>** and **PM<sub>2.5</sub>** concentrations at a 200 m x 200 m scale. CAMx is a sophisticated **open-source** photochemical grid model developed by scientists at Ramboll. CAMx has been used extensively for decades to support **scientific research** and **regulatory** air quality assessments for local, state, regional, and federal agencies (Figure 3). It has been well-tested and trusted since 1996 and is updated frequently. In addition to PM<sub>2.5</sub> and NO<sub>2</sub>, CAMx can be used to model transport and deposition of emissions of air toxics, ozone (O<sub>3</sub>), and other reactive tracers. CAMx allows Shair to account for photochemical reactions that produce O<sub>3</sub>, track the contributions from different **source categories** and use some of the same inputs that are used by local government agencies



for air quality modeling. Shair also takes advantage of CAMx's parallel processing capabilities to ramp up modeling speeds and ensure timely outputs.

Critical to the proper implementation of any air quality model are reliable emissions inventories and accurate meteorological information. Government agencies such as the Bay Area Air Quality Management District (BAAQMD) in California collect data and calculate emission inventories for different sources such as industry, traffic, ports, and airports. These inventories are then used by BAAQMD and other government agencies for studying and quantifying impacts of these emissions to regional air quality. Shair utilizes these where they are available. However, these emission inventories are not available in real-time and are either a snapshot of the past or forecasts for the future. Therefore, where additional data is available, these inventories are fully replaced or modified to **match current, real-time conditions**. Shair uses meteorological data from the National Center for Atmospheric Research's Weather Research & Forecasting model (WRF, pronounced "wharf").



**Figure 4:**

Overview of the workflow and data sources used by Shair for estimating traffic flows.



### **Realtime traffic emissions estimates**

Shair is capable of estimating realtime road traffic emissions based on traffic speed and congestion data provided via API from traffic reporting services. Shair currently leverages **realtime data** produced by HERE, a traffic API owned by trusted geospatial data providers Nokia and NAVTEQ. Shair can also be applied in areas with limited data availability from **traffic APIs** and traffic counts from local agencies. In these instances, the methodology for deriving traffic flows varies based on the data that is available. A general overview of the steps for estimating traffic flows and emissions is presented in Figure 4.

Ultimately, Shair calculates traffic emissions from **traffic flows** and regional **pollutant-specific traffic emission factors**. In the US, MOVES and EMFAC can be utilized to derive fleet-average emission factors for each hour of day, day of week, month, year, and speed. Depending on the model, additional emission factors could be prepared for different roadway and vehicle classifications. Outside of the US the COPERT model can be used for this purpose. These models are usually developed and maintained by government agencies and are based on local vehicle registration data.

Vehicle classification is an important factor in calculating emissions and it also affects the traffic induced turbulence discussed in the Shairstreet section below. However, such data cannot be easily obtained or estimated. The current version of Shair uses **travel demand model** results to estimate the average vehicle fleet composition by different road types and time of day.

### **Shairstreet**

Shairstreet is the roadway component of the Shair air quality modeling system that provides a high-resolution (**10m x 10m**) pollutant concen-

tration map from the real-time traffic emissions described above. Shair-street works best when it is coupled with CAMx, though it can also be used as a standalone model to estimate concentrations of pollutants from traffic in urban areas.

The Shairstreet model is comprised of **three main components**:

- An urban street canyon model
- A simplified nitrogen oxide ( $\text{NO}_x$ ) chemistry model
- A dispersion model for pollutant concentrations away from the roadways

Traffic pollution can have a great impact in urban areas where traffic density and concentrations of vehicle exhaust gases are high and where buildings adjacent to the street produce “street canyons” that modify windflow conditions and restrict dilution. Capturing the complexity of windflow patterns in street canyons is essential to understanding how pollutant concentrations change on streets and sidewalks – where exposure potential is increased – and how they move throughout a region. Shairstreet can use several different methods to account for street canyon effects.

The default Shairstreet street canyon model developed at Shair is based on the **Operational Street Pollution Model (OSPM)**, a semi-empirical model developed by researchers at the National Environmental Research Institute of Denmark (Berkowicz et al 1997). OSPM, which has been used and updated over the last 20 years, operates under the assumption that the pollutant concentration of a street canyon has 3 major contributors: direct contributions emitted from traffic affected by ground-level wind, contributions from pollution in air that is recirculated by a well-mixed vortex generated inside of the canyon, and background concentrations

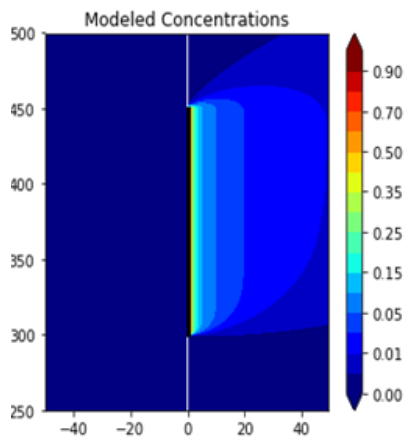
entrained above the street canopy (i.e., above the street canyon and the buildings that create it).

Shairstreet's default street canyon model assumes that vertical dispersion is governed primarily by mechanical turbulence with thermal stratification ignored, and that street-level mechanical turbulence is created by near-surface wind and vehicles in the street. **Realtime vehicle counts** and **average vehicle speeds** for all road links – which are based on speed and congestion data collected from a service like the HERE API – are used as inputs for calculating street level vertical turbulence. The vehicle induced turbulence is especially important on calm days, when the ambient turbulence is small, in determining pollutant concentrations. The street level wind speed is derived from canopy-level windspeeds provided by **WRF** and is calculated assuming a logarithmic reduction of the wind speed over the distance between the canopy top and the near-surface level and accounting for the angle of wind flow at the canopy top. Traffic induced turbulence is characterized in the same approach as OSPM, which accounts for flow distortion and related turbulence produced by vehicles moving on the street. For roads without adjacent buildings (i.e., no street canyon), street level concentrations are assumed to be affected by only the surface level wind and traffic induced turbulence.

Atmospheric dispersion of point emission sources can be well-modeled as Gaussian plumes. Estimating concentrations from line sources – as the Shair model estimates traffic emissions for road segments – only has an exact Gaussian solution when wind flow is perpendicular to that line source, but perpendicular airflow is frequently not a reality. Some models, like AERMOD, approximate line sources as a high number of point or area sources for dispersion calculations, but this is computationally expensive when there are a large number of roadways and related line

segments. Instead, Shairstreet employs a computationally efficient method for approximating **Gaussian line source dispersion** based on work published by researchers at the University of California, Riverside and the National Center for Atmospheric Research (VH2006) (Venkatram and Horst 2006). This method has been shown to produce small, acceptable errors when compared to an exact solution. The resulting dispersed concentration values, similar to those demonstrated in Figure 5, are then used to inform the concentrations of pollutants in nearby grid cells of the Shair model (note: concentrations in Figure 5 are normalized).

Shairstreet also **accounts for the formation of  $\text{NO}_2$** , which is a complex



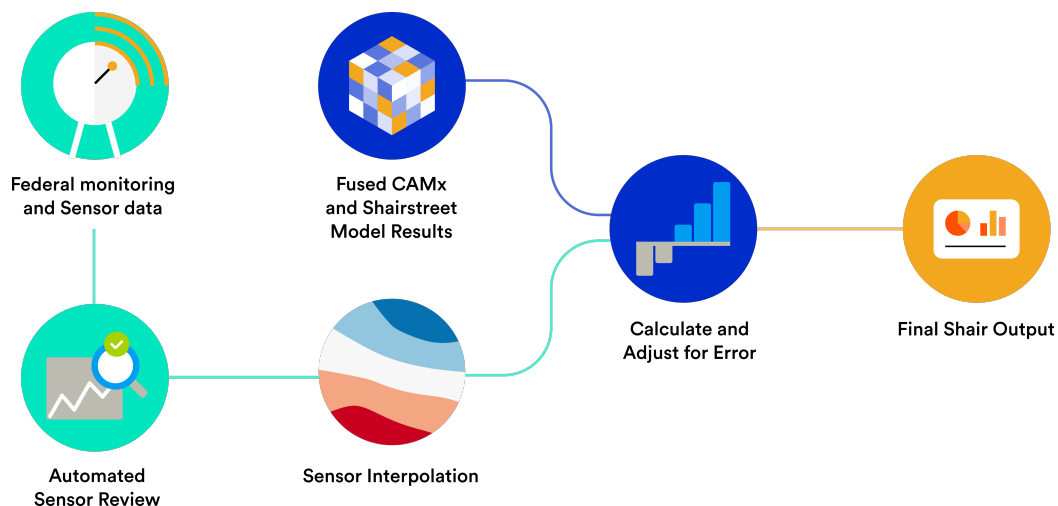
**Figure 5:**  
Example implementation of VH2006 dispersion for a line source with wind at 45 degrees from the horizontal using normalized concentrations.

photochemical process that depends on the amount of  $\text{NO}_x$  and  $\text{O}_3$  available. To model  $\text{NO}_2$ , Shairstreet uses a chemistry model published in 2011 (During et al 2011), which accounts for the photolytic decomposition of  $\text{NO}_2$ , temperature, mixing time, and compound residence time along with background concentrations of  $\text{NO}_x$  and  $\text{O}_3$ , which can be taken from nearby reference monitors.

Since Shairstreet calculates traffic concentrations resulting from each individual road link, Shairstreet can also be used as a tool for urban planning informed by air quality.

## Integrating measurement data in near-realtime

A major innovation of the Shair model is its integration of near-realtime measurement data from both regulatory and sensor-based monitors to reduce model bias. All modeled pollutant concentrations are **compared to measurements** taken across the modeled domain. Errors discovered in the model are then adjusted for. The process Shair uses for this (“Shairsense”) is described graphically in Figure 6. The **most recent measurements** available are automatically called via API from all sensors and regulatory monitors within the modeling domain to which we have been given access. Values between these measurements are interpolated using objective analysis methods like the Barnes method or kriging to produce an estimate of air quality at every point throughout the modeled domain. The difference between the values rooted in measurements and the modeled values are compared at each point to **estimate model error in near-realtime**. Model output is then **adjusted** for that error prior to production of the final Shair model output. As a result, Shair’s model output is **anchored in near-realtime measurement data**.



**Figure 6:**  
Shair’s automated framework for integrating measurements and model output.

One issue faced by community members and policy makers relying on data in near-realtime is that scientists have not had a chance to manually verify values or to check for false measurements and equipment failures. The result can be misinformed decision making, especially when people are worried about acute exposure events. Therefore, Shair has developed an automated review process using sensor operating specifications and historical trends. Our review process **flags suspicious measurements** and possible outliers to prevent immediate integration with Shair. Once flagged, these data can be examined for trends to **identify unknown emission sources** or failing hardware.

### **Fully Automated and Cloud Native**

Once the Shair model is customized, configured, and tuned for a geographic region, it is deployed in a **completely automated** way to the cloud. The Shair workflow leverages Google's cloud computing architecture – GCP – to estimate near-realtime hyper-local air quality concentrations in a manner that is **scalable** to large spatial domains in a compute-**resource efficient** way. On GCP, Shair utilizes cloud-native tools like Kubernetes and Argo Workflows to execute all phases of real-time traffic forecasting, Shairstreet, data quality control, and Shairsense. Data can be disseminated to customers in a secure way via our **web/phone apps** and **APIs** which include data and map-tile capabilities.

## **Driven to empower communities**

Shair has been methodically designed to provide **reliable, science-backed information** on local air quality trends to empower a wide variety of users to devise and implement effective and defensible air quality interventions. Shair output allows city planners, environmental managers, and regulators to identify spatial and temporal **pollution hotspots** with

detailed air quality maps rooted in near-realtime measurements. Planners and regulators can also deploy Shair to gain a clearer understanding of how **source contributions** in their region change across neighborhoods, meteorological conditions, and time facilitating more efficient and effective mitigation efforts. Our online mapping solution and data insights can be used to boost a community's **environmental engagement**, enabling citizens to plan **healthier commutes and recreation**. Every aspect of Shair is designed with **actionability** in mind to support users' efforts to improve their health and environment.



## About us

Ramboll is an international consulting firm focused on building a more sustainable world. Ramboll's network of experts hold deep domain knowledge in air pollution measurement and mitigation. The Ramboll Shair team focuses on integrating state-of-the-science dispersion modeling approaches with emerging low-cost sensing solutions to produce hourly estimates of air quality at a 10m x 10m resolution and to trace measured pollutants back to their sources for effective mitigation.

## Our Team

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## Contact us!

Request a demo or find out more by visiting [www.ramboll-shair.com](http://www.ramboll-shair.com) or sending us an email at [shair@ramboll.com](mailto:shair@ramboll.com). Shair can be configured to meet your specific needs. Don't have any of your own sensors yet? The Shair team would be delighted to introduce you to one of our sensor hardware partners.

## References

Berkowicz et al. 1997. "Modelling traffic pollution in streets. Roskilde: National Environmental Research Institute." [http://orbit.dtu.dk/files/128001317/Modelling\\_traffic\\_pollution\\_in\\_streets.pdf](http://orbit.dtu.dk/files/128001317/Modelling_traffic_pollution_in_streets.pdf).

During et al. 2011. "A simplified NO/NO<sub>2</sub> conversion model under consideration of direct NO<sub>2</sub>-emissions." *Meteorologische Zeitschrift* (20): 67-73. doi:10.1127/0941-2948/2011/0491.

EDF. (2017, June 4). Study: Urban air pollution varies from block to block. Retrieved September 9, 2019, from <https://www.edf.org/media/study-urban-air-pollution-varies-block-block>

IHME. 2019A. United States - All causes attributable to Particulate matter pollution, both sexes, all ages. University of Washington. Accessed 27 2019, June. <https://vizhub.healthdata.org/gbd-compare/>.

IHME. 2019B. Global, Both sexes, All Ages, All causes, risk: Air pollution. Accessed September 6, 2019. <http://ghdx.healthdata.org/gbd-results-tool>.

Martin, R, M Brauer, A van Donkelaar, G Shaddick, U Narain, and S Dey. 2019. "No one knows which city has the highest concentration of fine particulate matter." *Atmospheric Environment*. doi:10.1016/j.aeaoa.2019.100040.

Venkatram and Horst. 2006. "Approximating dispersion from a finite line source." *Atmospheric Environment* (40): 2401-2408. doi:10.1016/j.atmosenv.2005.12.014.