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Air pollution knowledge assessments (APnA) for 20 Indian cities

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ABSTRACT

Delhi, with a population of 22 million (1.6% of national total) is one of the most polluted capital cities in the world. Nearly 50% of the published literature in India focus on air pollution in Delhi. However, air pollution impacts are not limited only to the capital city. Yet, there is little information and attempt to quantify these impacts for Tier 1 and 2 cities, even though they account for > 30% of India's population. To remedy this vacuum of information, the Air Pollution knowledge Assessments (APnA) city program deliberately focuses on 20 Indian cities, other than Delhi. We established baseline multi-pollutant high-resolution emissions inventory, after collating information from multiple resources detailed in this paper, which was used to estimate spatial concentrations of key pollutants across city's urban airshed using WRF-CAMx chemical transport modeling system. The inventory includes anthropogenic sources, such as transport (road, rail, ship, and aviation), large scale power generation (from coal, diesel, and gas power plants), small scale power generation (from diesel generator sets for household use, commercial use, and agricultural water pumping), small and medium scale industries, dust (road resuspension and construction), domestic (cooking, heating, and lighting), open waste burning, and open fires and non-anthropogenic sources, such as sea salt, dust storms, biogenic, and lightning. The emissions inventory is currently in use for 3-day advance air quality forecasting for public release on an on-going basis. Using meteorological parameters and big data like gridded speed maps from google, the emissions inventory is dynamically updated. The results from this research will be valuable to local and national policy makers - especially the information on source contributions to air pollution.

1. Introduction

That air pollution is a serious environmental health issue in India is not under debate. The global burden of disease assessments estimated 0.74 million and 1.1 million premature deaths in 1990 and 2016 due to outdoor $PM_{2.5}$ and ozone pollution and 0.99 million and 0.78 million premature deaths in 1990 and 2016 due to household (indoor) $PM_{2.5}$ pollution (GBD, 2018). In the monitoring database released by the World Health Organization (WHO) in April 2018 covering 100 countries for the period of 2011 and 2016, India has 14 of the top 15 cities with the worst $PM_{2.5}$ pollution. Among megacities of the world, Delhi tops the list for PM_{10} pollution (WHO, 2018). What is unclear however is the extent of the problem in India. The maps in Fig. 1 compares surface $PM_{2.5}$ pollution levels for 1998, 2010, and 2016 extracted from Van Donkelaar et al. (2016). These data were produced using a combination of satellite observations, global chemical transport model using global emission inventories, and available ground-based measurements. Of the 640 districts in India (as per 2011 Census), 27% exceed the current annual standard of $40 \mu g/m^3$ in 1998, 45% and 63%

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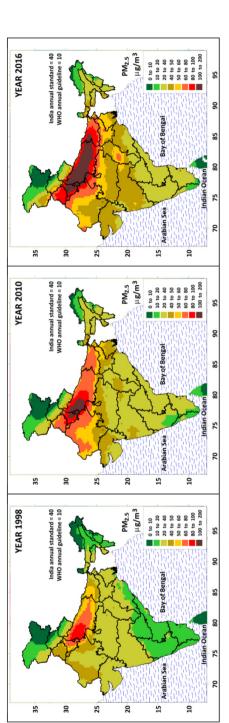
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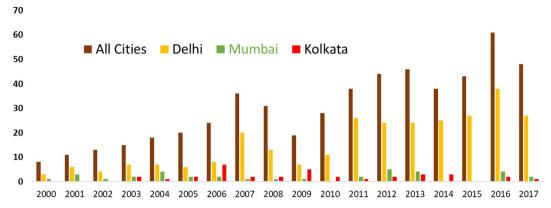


Fig. 2. Number of journal articles published between 2000 and 2017 (from SCOPUS search) with some reference to air pollution research in an Indian city.

exceed the same in 2010 and 2016. 99.5% of them exceed the WHO guideline of $10 \,\mu g/m^3$ in 2016.

With urbanization, domestic migration is not only increasing population in the major metropolitan cities, but also in medium and small cities. These towns are adjusting their infrastructure needs, such as transport and waste management, to accommodate the growing population and economy. To take the air pollution policy debate further at the national scale, we need to move beyond anecdotal evidence, quantify and spatially map out pollution, and assess the impact of sources at the local scale (Guttikunda et al., 2014).

Big cities have some published studies that quantified source contributions, which led to some policy discussion (Pant and Harrison, 2012). The Central Pollution Control Board (CPCB, New Delhi, India) conducted a source apportionment study for six cities – Bengaluru, Chennai, Delhi, Kanpur, Mumbai, and Pune in 2006 (CPCB, 2011). It has been at least 12 years since CPCB has conducted any other source apportionment studies. There is very little information for other cities on ambient monitoring and source contributions. Of the number of scientific studies conducted to identify the sources of $PM_{2.5}$ and PM_{10} pollution via chemical analysis and receptor modeling since 2000, 70% repeatedly covered Delhi, Mumbai, Kolkata, Chennai, Hyderabad, and Agra (Guttikunda et al., 2014). We distilled relevant journal articles by conducting a full search on the SCOPUS database (title, abstract, and keywords) for all articles published between 2000 and 2017 for 50 Indian cities. This was done to get an idea of the state of research on topics related to air pollution - emission inventories, industrial studies, dispersion modeling, source apportionment, emission factors, and ambient monitoring. Accordingly, we collected 543 papers that mentioned at least one Indian city. Unsurprisingly, Delhi is the most studied city in India with 283 hits (> 50% of the total). Mumbai, the commercial capital of India and Kolkata, tied for a distant second with 35 hits each (Fig. 2). The trend in the number of publications is also an indication of the importance given to the air pollution problem in Delhi by the researchers, media, and national/international organizations.

Designing an effective air quality management plan for a city requires robust data on the sources of air pollution. The Air Pollution knowledge Assessments (APnA) city program is an attempt to fill the vacuum of information and by creating a baseline of air pollution related information for Indian cities, a necessary starting point for the policy makers, academic researchers, and citizens to chart out strategies for better air quality. We believe that establishing the baseline or quantifying the extent of the problem in cities other than the metropolitan cities is the starting point for change.

In this paper, we present our work on building these baselines for 20 Indian cities as (a) review of city characteristics such as ambient air quality, urban growth, census fields, geography, and meteorology (b) multi-pollutant emission inventory at 1-km resolution for the city airsheds and (c) WRF-CAMx chemical transport modeling to study trends and estimate source contributions to $PM_{2.5}$ and PM_{10} pollution.

2. City airsheds

For 20 Indian cities, we collated city geography and related characteristics as listed in Table 1. The list includes 2 megacities (Bengaluru and Chennai) with population above 10 million; 4 cities with population between 5 and 10 million (Chandigarh, Kanpur, Patna and Pune); 12 cities with population between 2 and 5 million; 2 cities (Dehradun and Ranchi) with population between 1 and 2 million. This list covers 13 states and 1 union territory (Chandigarh). These 20 cities host 90 million inhabitants (approximately 7.5% of national total) over an airshed of 44,000 sq.km (approximately 1.3% of national total). Table 2 catalogues a brief description of the selected cities.

The Supplementary resources include the Google KML files of individual city airsheds. For each city, on average 25% of airsheds area is designated as urban and extends beyond the political boundaries to include sources that may contribute to air pollution in the main city. 7 of 20 cities are on the Indo-Gangetic plain, the most polluted region in India (Fig. 1); 2 coastal cities with big commercial ports (Chennai and Kochi); 1 city in the hills (Dehradun); 14 North Indian cities and 6 South Indian cities.

For each airshed, we extracted built up area from Pesaresi et al. (2016) for 1975, 1990, 2000, and 2014 (Table 1). Fig. 3(a) offers a comparison of the built-up areas between 1990 and 2014 for 4 cities – Agra, Bengaluru, Chennai, and Patna. Remaining figures are

Table 1 List of cities included in the India APnA city program.

| | City name | State name | Domain pop | Urban pop | Airshed grids | Airshed area | Urban area | Built-up area (sq.km) | |
|----|-------------|----------------|------------|-----------|----------------|--------------|------------|-----------------------|------|
| | | | million | % | # x * # y | sq.km | % | 1975 | 2014 |
| 1 | Agra | Uttar Pradesh | 4.1 | 91% | 40 × 40 | 2230 | 26% | 18 | 107 |
| 2 | Amritsar | Punjab | 2.3 | 85% | 40×40 | 2327 | 27% | 65 | 119 |
| 3 | Bengaluru | Karnataka | 11.9 | 93% | 60×60 | 4574 | 27% | 147 | 631 |
| 4 | Bhopal | Madhya Pradesh | 2.7 | 97% | 40×40 | 2157 | 26% | 47 | 141 |
| 5 | Bhubaneswar | Odisha | 3.7 | 88% | 30×50 | 1982 | 27% | 17 | 118 |
| 6 | Chandigarh | Chandigarh | 5.0 | 85% | 60×60 | 5176 | 27% | 300 | 420 |
| 7 | Chennai | Tamil Nadu | 10.7 | 93% | 50×50 | 3177 | 27% | 108 | 522 |
| 8 | Coimbatore | Tamil Nadu | 3.0 | 88% | 50×50 | 3156 | 19% | 42 | 224 |
| 9 | Dehradun | Uttarakhand | 1.2 | 88% | 40 	imes 20 | 1149 | 19% | 18 | 59 |
| 10 | Indore | Madhya Pradesh | 3.1 | 93% | 40×40 | 2150 | 28% | 28 | 247 |
| 11 | Jaipur | Rajasthan | 4.8 | 94% | 40×40 | 2223 | 28% | 48 | 294 |
| 12 | Kanpur | Uttar Pradesh | 5.2 | 90% | 40×30 | 1661 | 26% | 25 | 190 |
| 13 | Kochi | Kerala | 3.9 | 94% | 40×40 | 2012 | 26% | 135 | 374 |
| 14 | Ludhiana | Punjab | 2.8 | 87% | 40×40 | 2310 | 28% | 122 | 233 |
| 15 | Nagpur | Maharashtra | 3.7 | 96% | 40×40 | 2125 | 27% | 123 | 240 |
| 16 | Patna | Bihar | 5.3 | 85% | 60 	imes 30 | 2473 | 19% | 22 | 85 |
| 17 | Pune | Maharashtra | 6.9 | 97% | 40×40 | 2091 | 28% | 46 | 280 |
| 18 | Raipur | Chhattisgarh | 3.2 | 87% | 60×30 | 2390 | 19% | 110 | 314 |
| 19 | Ranchi | Jharkhand | 1.9 | 87% | 40×40 | 2159 | 22% | 19 | 71 |
| 20 | Varanasi | Uttar Pradesh | 4.2 | 71% | 40×40 | 2193 | 16% | 37 | 100 |

Example - Airshed grids 60×30 denotes 60 longitudinal and 30 latitudinal grids at 0.01° resolution; Airshed area is the area covered by these 60×30 grids; Note - Built-up area for 1975 and 2014 are extracted from the global human settlement (GHS) layer of landsat imagery (Pesaresi et al., 2016); Gridded population is from GRUMP (2015) and Landscan (2013); Urban pop % = fraction of population in the grids designated as urban in the select airshed; Urban area % = fraction of grids designated as urban in the select airshed.

included in the Supplementary. Built up area on average increased by 200% between 1975 and 2000 and by 40% between 2000 and 2014. Chandigarh, one of the original planned cities, showed the least increase in built-up area, while cities with the largest percentage change are Bhubaneswar, Indore, Jaipur, Kanpur, and Pune.

3. Air quality monitoring

The CPCB and the State Pollution Control Boards (SPCBs) maintain and operate the national ambient monitoring programme (NAMP). This is the only official ambient monitoring database available in India since 1990. The monitoring network is slowly expanding. In 2015, there were 206 manual monitoring stations in 46 cities. As of May 2018, there are 700 manual stations and 117 continuous stations (33 stations are operating in Delhi). In September 2017, there were 74 continuous stations in operation – a 50% increase over 6 months. The continuous ambient monitoring stations report pollution levels for all the criteria pollutants every 15 min. This data is available in real time on CPCB's website.

The manual stations measure $PM_{2.5}$, PM_{10} , SO_2 , and NO_2 and the procedure includes manually changing and collecting samples. While $PM_{2.5}$ as a criteria pollutant was introduced in 2009, it was only added as a pollutant to be measured at the manual stations in 2016. The Supplementary includes all available manual monitoring network data for the years between 2011 and 2015. For the manual stations, details on time of sample collection and time of filter change are unknown. Standard protocol suggests that at least 100 samples per station are required. In 2016, number of collection days for all stations averaged 94.2 for PM_{10} , 93.9 for SO_2 , 94.6 for NO_2 , and 71.3 for $PM_{2.5}$.

The WHO (2018) report utilized the data from both manual and continuous networks. Table 3 lists a summary of the data used in the WHO report. Of the 14 Indian cities in the top 15 most polluted cities, except for Delhi, Lucknow, and Agra, the remaining 11 cities – Kanpur, Faridabad, Varanasi, Gaya, Patna, Muzaffarpur, Srinagar, Gurgaon, Jaipur, Patiala, and Jodhpur, have one monitoring station each. A sample of one is not enough to know the true spatial and temporal trend of air pollution in any city.

Including Delhi, all the cities are operating less than the required number of monitoring stations to truly represent the mix of sources and activities in the city. Unlike cities in Europe and the United States, where vehicle exhaust is the main source of pollution, the range of sources in Indian cities is wide - from large and small industries to vehicle exhaust to road dust to biomass burning every day for cooking and heating. Using the CPCB's thumb rule (CPCB, 2003), which considers the total population of an airshed and the mix of activities, we estimate a need for 4000 continuous monitoring stations in India – 2800 in the urban areas and 1200 in the rural areas, to represent a population of 1.3 billion. These details at the state and the district level are presented in the Supplementary. For example, Uttar Pradesh with a population of over 200 million requires 558 stations and Delhi with a population of over 20 million requires 77.

Table 4 is a summary of the available ambient monitoring data and the number of manual and continuous monitoring stations operational in 20 Indian cities from this study. Air monitoring capacity is low in all the cities. As of May 2018, 6 of 20 cities do not have any continuous monitoring stations; 11 cities have only one continuous monitoring station. Bengaluru added 5 new continuous

| General description of the cities included in | the India APnA city program. |
|---|------------------------------|
|---|------------------------------|

| | City | Description |
|----|-------------|--|
| 1 | Agra | On the banks of river Yamuna, the city is known for the Taj Mahal – a UNESCO World Heritage site, which attracts at least 8 million tourists every year. Agra's economy is made up of a thriving small-scale industry sector connected to handicrafts, leather goods, and iron foundries, tourism, and agriculture. The delicate inlay and carving work in white marble of the Taj Mahal started getting affected by the rising air pollution levels. In 2000, the Supreme Court mandated a 50-km safe-zone known as the "Taj Trapezium Zone" (TTZ) – which will be free of polluting industry and fueled vehicles. This has had scant impact on the pollution levels in the city, as Agra ranked 4th most polluted in 2016. |
| 2 | Amritsar | City is home to the Golden Temple (the spiritual center for the Sikh community) and lies on the Grand Trunk road (GT Road) from Delhi to Lahore in Pakistan. The temple visitors frequently take inter-city buses contributing to the day to day commercial and transportation activities in the city. Within the city, majority of the short trips are covered by rickshaws, auto rickshaws, and taxis. Amritsar's economy is made up of tourism, carpets and fabrics, farm produce, handicrafts, service trades, and light engineering. Coupled with small and heavy vehicles plying along the GT Road, vehicle exhaust and road resuspended dust has become a major source of air pollution in the city, especially due to 50,000 unregistered autorickshaws and double that of the registered ones. A majority of these use petrol mixed with kerosene. |
| 3 | Bengaluru | Bengaluru (formerly Bangalore), the capital of Karnataka was a pioneer in that it was one of the first information technology (IT) hubs of India. What was once a laid-back retiree town, is now a bustling metropolis with 10 + million inhabitants. The population grew exponentially from 1980s with several companies, especially IT industry, establishing base in the city, creating higher demand for transport and construction. The unplanned growth in the city boosted the infrastructure activities that the municipality attempted to ease by constructing a flyover system and by imposing one-way traffic systems, which were unable to adequately address the on-road issues and associated increase in air pollution. As the demand for housing has grown, the city finds itself spreading beyond its erstwhile boundaries to accommodate it. |
| 4 | Bhopal | Bhopal, the capital of Madhya Pradesh, has been selected under National Smart Cities program's first round for integrated urban development. The old city of Bhopal is home for small and medium industries covering electrical goods, medicinal products, cotton, chemicals, jewelry, flour milling, cloth weaving, painting, matches manufacturing, wax manufacturing, and sporting equipment. It also houses the Bharath Heavy Electricals Limited (BHEL), which is one of the largest engineering companies in India that manufactures coal-fired power plant boilers (among many other heavy machinery). Mandideep, another industrial suburb to the south of the city, is the |
| 5 | Bhubaneswar | largest industrial area in Madhya Pradesh. Bhopal is also synonymous with the Union Carbide Bhopal Disaster in the early 1980s. Cuttack was replaced by Bhubaneswar as the capital of Odisha (formerly Orissa) in 1949 and is one of the first planned cities in India, along with Jamshedpur and Chandigarh. Together they are referred to as the twin cities of Odisha. The city lies on the banks of Mahanadi river and between the naturally rich plains of Daya river and Kuakhai Rivers. As a tier-2 city, it is emerging as a center for education and IT. With increasing population, construction activity, and transport demand and limited waste management options, air pollution is reaching critical levels, even in the summer months. |
| 6 | Chandigarh | Chandigarh is a union territory that serves as the capital for the states of Haryana (on the east) and Punjab (on the north, west and south). Chandigarh airshed includes Patiala and Ambala, all within a 30 km radius of each other. Chandigarh, was one of the first planned cities in India and is known for its urban design, by Swiss architect Lr Corbusier, and sometimes referred to as "Pensioner's Paradise". There are 15 medium to large industries and over 2500 smaller units manufacturing paper, basic metals and alloys, machinery, food products, sanitary ware, auto parts, machine tools, pharmaceuticals, and electrical appliances. |
| 7 | Chennai | As the capital city of Tamil Nadu, Chennai is one of four metropolitan cities of India with 10 + million inhabitants. With its proximity to the Bay of Bengal and thus access to markets in East Asia, it is also an important port city. Apart from trade and shipping, the automobile industry (30% of India's auto industry), chemical and petrochemical industry, software services, medical care, and manufacturing form the foundation of the economic base for Chennai. The Ennore Port, the first major corporate port with an annual capacity of 30 million tons of cargo in 2012–2013, handles coal (most of the supply is for the two thermal power plants with dedicated feeder lines running from the ports), iron ore, oil, and commercial commodities. |
| 8 | Coimbatore | After Chennai, Coimbatore is the second largest city in Tamil Nadu, located on Noyyal river bank and surrounded by the Western Ghats, houses > 25,000 small, medium and large industries covering textile and cotton industry, manufacturing, poultry farming, education, IT, and health care. |
| 9 | Dehradun | Dehradun, is the capital of Uttarakhand, carved out of the state of Uttar Pradesh in 2000. The city capital lies in the Doon Valley, in the foothills of the Himalayas, between the river Ganges on the east and the river Yamuna on the west. The city was developed as a getaway from the hot summers of the plains and hosts several institutions such as the Indian Military Academy (IMA), ITBP Academy, Indira Gandhi National Forest Academy (IGNFA), Zoological Survey of India (ZSI), Forest Research Institute (FRI) among several others. Construction and transport are the fastest growing sectors in the region. The geographical and meteorological conditions do not allow for easy dispersal of emissions. |
| 10 | Indore | Indore is 200 km from Bhopal, is the largest city and the commercial center in the state of Madhya Pradesh and part of the 100 smart cities program in India. A network of national and state highways through the city (NH3 to Agra and Bombay; NH59 to Ahmedabad and Betul; SH27 to Burhanpur; and SH34 to Jhansi) made it an important transport and logistics hub in the country. Major industrial areas are in Pithampur and Sanwer special economic zones focusing on soybean processing, automobile, IT, and pharmaceuticals. |
| 11 | Jaipur | As the state capital of Rajasthan, Jaipur is located 260 km from New Delhi and forms a part of a popular tourist circuit with Agra (240 km). Jaipur also serves as a gateway to other tourist destinations in Rajasthan such as Jodhpur, Jaisalmer, Udaipur, and Mount Abu. City is an educational and administrative center and its economy is fueled by tourism, gemstone cutting for jewelry, luxury textiles, and IT. During winter months, flights towards New Delhi airport are often diverted to Jaipur airport under foggy conditions. Another telling statistic that illustrates the level of traffic congestion in Jaipur is that 60% of the city roads are used for parking – the highest in any city in India (CSE, 2017). |
| 12 | Kanpur | Kanpur is one of the largest industrial towns in North India, on the Indo-Gangetic plain, and the largest city in the state of Uttar Pradesh. Kanpur supports the largest textile and leather processing sectors in the region. Kanpur briefly attempted to put in place a public-private partnership system for solid waste management, which after failing led to one of the major causes of air, water, and waste management issues in the city. In 2011, CPCB's assessment of air pollution in six cities (Delhi, Mumbai, Kanpur, Pune, Chennai and Bangalore) reported that air pollution sources in Kanpur are industries (biggest cause) followed by open waste burning, vehicles, road dust, and domestic cooking (CPCB, 2011). |
| 13 | Kochi | |

Table 2 (continued)

| | City | Description |
|----|----------|--|
| 14 | Ludhiana | Kochi in the state of Kerala is an important port town and houses the Southern Naval Command of the armed forces, the largest international container trans-shipment terminal in India (the Cochin shipyard), and the offshore mooring of the Bharath Petroleum Corporation Limited (BPCL). It is also an important chemical industrial and manufacturing hub and an emerging IT center in Kerala. Ludhiana is an agricultural and industrial town in Punjab housing several small scale industrial units producing auto parts, appliances, machine parts, and apparel. The city has been on the most polluted list, drawn by WHO for multiple years. The distributed industries, generator use, agriculture burning, adulterated fuel for three-wheelers and adverse meteorological conditions (in the winter months), and |
| 15 | Nagpur | open field burning (post-harvest season before the winter months) are responsible for its bad air quality. Nagpur holds the distinction of being the geographical center of the Indian peninsula. Nagpur lies on the Deccan plateau and is the 3rd largest city and a major commercial and political center of the state of Maharashtra. Agriculture (in particular oranges and ayurvedic medicine processing) forms a large share of its economy, along with IT, manufacturing, transport systems, and mining. Due to its proximity to the coal belt in Central India, power generation from coal is large (47% of state's power is generated around Nagpur). The government is investing in Nagpur's infrastructure development to make it a freight movement hub. |
| 16 | Patna | As the state capital of Bihar, Patna with 72 wards is the largest in the state, located on the banks of rich Indo-Gangetic plain, and is among the states with the highest population density. The city economics is supported by agricultural activities (> 30% is rice), small scale manufacturing industries (including brick kilns, pulses, shoes, scooters, masur, chasra, electrical goods, and cotton yarn), and power generation (3000 MW power plant 100 km west of the city). |
| 17 | Pune | Once referred to as a city for students and retirees, Pune has now established manufacturing, glass, sugar, and metal forging industries, and more recently IT and auto industry. The Hinjewadi IT Park (located west of the Pimpri Chinchwad satellite city) is a project started by Maharashtra Industrial Development Corporation and Magarpatta city under a special economic zone scheme to the east has over 20,000 inhabitants, mostly working on IT. Construction of townships and developments mean that raw material from stone quarries, river sand and bricks are in constant demand. Fugitive dust emissions from the quarrying process, use of diesel for power generation in the commercial and industries is on the rise. |
| 18 | Raipur | In Chhattisgarh, Raipur-Durg-Bhilai (RDB) tri-city area hosts the new administrative capital of the state, interconnected by an expressway forming the industrial corridor, and located on the banks of the Mahanadi River. RDB is the largest steel manufacturing hub in India and is home to several foundries, metal-alloy plants, steel casting and rolling, sponge iron plants, cement production, and a large chemical (formalin) manufacturing plant. Bhilai Steel plant is the largest integrated steel plant in India. Most of the electricity generation at the power and the steel plants is coal-based. While Durg-Bhilai are largely industrial, Raipur has emerged as the commercial hub. Naya Raipur (New Raipur) is a satellite city located 17 km away from Raipur covering 8000 ha is expected to house 450,000 inhabitants. |
| 19 | Ranchi | Ranchi is the state capital of Jharkhand (the state was formed in 2000) and is part of the smart cities program in India. Much of the city's economy is driven by processing industry based on resources from the forests of Jharkhand. These include timber, matches, rayon, medicinal plants and seeds. It is a relatively small city, but from the time the state was formed in 2000, the number of vehicles in the city has increased 20 times. With an increase in population – this erstwhile hill station is now dealing with haphazard construction and insufficient systems for solid waste management leading to open burning. |
| 20 | Varanasi | Formerly Banaras, Varanasi lies on the banks of the Ganges river in Uttar Pradesh and is center for pilgrimage and tourism in India (~3 million domestic and 200,000 visitors annually). This puts a significant burden on city's solid waste management infrastructure, with much of the waste either burnt or dumped in landfills or in the river). City is famous for several handcrafted goods such as muslin and silk fabrics, perfumes, ivory works, and sculpture, diesel locomotive works, and Bharat Heavy Electricals Ltd. |

monitoring stations in 2018, bringing their total to 10, which is 25% of the recommended capacity to represent approximately 11.9 million inhabitants.

All cities exceed the annual PM_{10} standard of $60 \,\mu\text{g/m}^3$. Major sources of PM_{10} include combustion of coal, kerosene, petrol, diesel, biomass, cow dung, and waste as well as dust. Coimbatore reported just above the standard. 6 cities - Agra, Bengaluru, Jaipur, Kanpur, Ludhiana, and Raipur, recorded 4–6 times the annual standard. These are among the fastest growing Indian cities, traditionally known to be dusty due to a lot of construction activities and dust on the roads which is resuspended when vehicles pass.

All cities comply with the annual SO₂ standard of $50 \,\mu\text{g/m}^3$. Major sources of SO₂ include combustion of coal and diesel. Bengaluru, Chennai, and Pune (most populated of the 20) recorded the highest SO₂ concentrations, indicating higher consumption of diesel in personal vehicles, freight vehicles, and diesel generator sets.

9 of 20 cities exceed the annual NO_2 standard of $40 \,\mu\text{g/m}^3$. Major sources of NO_2 include combustion of petrol, diesel, and gas (i.e., transport related emissions) and large industries. Bengaluru, Chennai, Jaipur, Nagpur, Kanpur, and Pune record the highest concentrations.

Satellite observations coupled with global model simulations provide a useful baseline to monitor progress and validation (van Donkelaar et al., 2016). Table 3 lists data extracted for years 1998 and 2016. Refer to the Supplementary for data for all years in between and all 640 districts. These methodologies are an integral part of the global burden of disease estimates for outdoor and indoor air pollution (Brauer et al., 2016; GBD, 2018; Shaddick et al., 2018) and the premature death due to outdoor $PM_{2.5}$ estimates published by WHO (WHO, 2016). The model estimates underrepresent the pollution levels due to a mix of unknowns, such as model grid resolution, uncertainties in the emission inventories, lack of enough ground-based measurements to corelate with the satellite observations. Of the 20, only 5 cities show concentrations under the national ambient standard of $40 \,\mu\text{g/m}^3$ – Chennai and Kochi are coastal cities and Bhubaneswar close to the coast, with the advantage of land-sea breeze that disperses most of the emissions from the industries, vehicles, and ports; Bengaluru and Coimbatore, also in South India have the advantage of strong winds to disperse most of the emissions. North India is landlocked, has a greater population density and a higher concentration of settlements. As a result, along with meteorological factors, there is an increase of 50–100% in their ambient $PM_{2.5}$ pollution levels between 1998 and 2016.

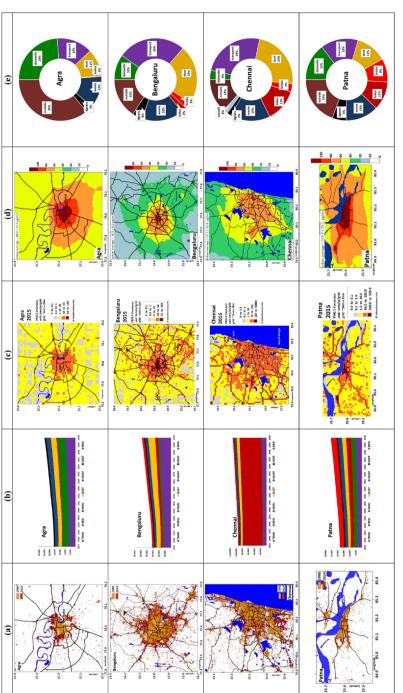


Fig. 3. Results for 4 out of 20 cities (a) Urban built-up area between 1990 and 2014 (b) Annual PM_{2.5} emissions (in tons/year) between 2015 and 2030 under the business as usual scenario (RESI = cooking, heating, and lighting; TRAN = road, rail, shipping, for 20 Indian cities in 2015 brick kilns and including thermal power plants; WAST = open waste burning; DGST = diesel generator sets; BRIC = brick kilns in the selected airshed) (c) Gridded annual PM_{2.5} emissions (in tons/year/grid) in 2015 (d) Modeled annual average PM_{2.5} concentrations (in µg/m³) in 2015 and (e) Modeled annual source contributions to PM_{2.5} in 2015 (outside is the contribution of boundary conditions, representing the influence of regional sources). All the figures for 20 Indian cities are included in the Supplementary.

| Breakdown of the PM2.5 | concentrations data | presented in WHO | (2018) |
|------------------------|---------------------|------------------|--------|
| | | | |

| Number of cities with measured $PM_{2.5}$ | 33 |
|---|---|
| Number of cities with measured + converted $PM_{2.5}$ | 135 |
| Average PM _{2.5} for cities with measured concentrations | $84.2 \pm 46.2 \mu g/m^3$ |
| Average PM _{2.5} for cities with converted concentrations | $53.5 \pm 30.4 \mu g/m^3$ |
| National annual ambient standard for PM2.5 | $40 \mu g/m^3$ |
| World Health Organization annual guideline for PM _{2.5} | $10 \mu\text{g/m}^3$ |
| Number of cities with 1 station only for PM2.5 measurements | 23 |
| Number of cities with 1,2,3 stations for converted PM _{2.5} | 30,28,33 |
| Number of cities with $4 +$ stations for converted PM _{2.5} | 7 |
| Year of data reported for PM _{2.5} measurements (number of cities) | 2016(33) |
| Year of data reported for converted PM _{2.5} (number of cities) | 2011 (1), 2012 (97), 2013 (1), 2014 (1), 2015 (2) |

Table 4

Number of operational stations (as of May 2018), number of minimum stations required, annual average $PM_{2.5}$, PM_{10} , SO_2 , and NO_2 concentrations (in $\mu g/m^3$), and modeled surface $PM_{2.5}$ concentrations from Fig. 1, in the 20 cities under the APnA city program.

| | | Monitoring stations | | | 2011–2015 manual | | | 2015-2017 continuous | PM _{2.5} from Fig. 1 | |
|----|-------------|---------------------|-------|-------|-------------------|-----------------|-----------------|----------------------|-------------------------------|-------|
| | City name | Man. | Cont. | Reco. | PM ₁₀ | NO_2 | SO_2 | PM _{2.5} | 1998 | 2016 |
| 1 | Agra | 6 | 1 | 23 | 362.2 ± 193.1 | 43.9 ± 22.1 | 9.3 ± 5.4 | 98.6 ± 86.1 | 77.1 | 116.9 |
| 2 | Amritsar | 3 | 1 | 18 | 195.6 ± 44.9 | 38.9 ± 7.6 | 13.8 ± 2.6 | 65.9 ± 64.8 | 41.5 | 79.1 |
| 3 | Bengaluru | 16 | 10 | 41 | 302.5 ± 208.0 | 69.2 ± 44.0 | 30.9 ± 23.5 | 32.3 ± 24.2 | 19.7 | 28.9 |
| 4 | Bhopal | 5 | 1 | 18 | 196.3 ± 117.3 | 23.2 ± 16.9 | 2.9 ± 2.5 | 47.8 ± 35.1 | 31.7 | 56.5 |
| 5 | Bhubaneswar | 5 | 0 | 22 | 128.0 ± 102.7 | 26.9 ± 17.2 | 3.0 ± 1.5 | | 17.1 | 37.9 |
| 6 | Chandigarh | 5 | 1 | 27 | 226.9 ± 103.6 | 44.3 ± 32.4 | 4.0 ± 1.9 | 59.4 ± 62.7 | 44.8 | 78.7 |
| 7 | Chennai | 11 | 3 | 38 | 199.8 ± 101.5 | 65.5 ± 37.1 | 39.7 ± 31.8 | 47.7 ± 73.7 | 17.6 | 32.6 |
| 8 | Coimbatore | 3 | 0 | 19 | 62.5 ± 39.7 | 26.9 ± 15.3 | 4.0 ± 1.7 | | 14.1 | 26.0 |
| 9 | Dehradun | 3 | 0 | 13 | 170.3 ± 53.8 | 26.3 ± 8.0 | 23.9 ± 7.3 | | 53.5 | 87.4 |
| 10 | Indore | 3 | 1 | 20 | 139.2 ± 69.8 | 18.9 ± 3.9 | 11.3 ± 3.7 | 40.4 ± 23.1 | 27.7 | 46.1 |
| 11 | Jaipur | 6 | 3 | 26 | 313.9 ± 141.3 | 79.8 ± 25.3 | 14.5 ± 7.5 | 70.4 ± 76.8 | 43.1 | 68.3 |
| 12 | Kanpur | 8 | 1 | 27 | 421.1 ± 188.5 | 70.7 ± 32.0 | 14.0 ± 11.7 | 111.4 ± 79.7 | 69.7 | 106.6 |
| 13 | Kochi | 7 | 0 | 23 | 221.6 ± 167.0 | 38.4 ± 30.5 | 10.7 ± 10.6 | | 13.9 | 26.1 |
| 14 | Ludhiana | 4 | 1 | 20 | 335.0 ± 149.6 | 48.0 ± 14.4 | 18.7 ± 6.0 | 74.4 ± 46.1 | 44.9 | 83.9 |
| 15 | Nagpur | 7 | 1 | 22 | 192.1 ± 112.5 | 64.7 ± 36.0 | 20.7 ± 18.8 | 53.9 ± 26.7 | 28.4 | 47.3 |
| 16 | Patna | 2 | 1 | 26 | 162.5 ± 91.7 | 33.0 ± 24.3 | 4.6 ± 5.0 | 121.3 ± 89.5 | 52.1 | 90.8 |
| 17 | Pune | 4 | 1 | 30 | 162.2 ± 104.3 | 82.5 ± 45.0 | 40.3 ± 20.9 | 43.1 ± 25.7 | 30.9 | 47.5 |
| 18 | Raipur | 2 | 0 | 19 | 275.2 ± 81.0 | 35.3 ± 13.3 | 12.9 ± 5.0 | | 29.6 | 54.6 |
| 19 | Ranchi | 1 | 0 | 16 | 190.1 ± 52.3 | 35.1 ± 5.4 | 18.3 ± 2.0 | | 32.5 | 55.6 |
| 20 | Varanasi | 2 | 1 | 23 | 139.3 ± 15.9 | 26.2 ± 6.7 | 18.4 ± 2.5 | 106.2 ± 83.4 | 51.7 | 94.0 |

Note: Man. = manually operated stations where filter papers are collected 2 times per week on average; Cont. = continuous ambient monitoring stations reporting data every 15 min; Reco. = recommended number of continuous monitoring stations for the city.

4. Urban emission inventories

4.1. Methods and data

We built an emissions inventory for each of the 20 cities for year 2015 for the following pollutants – particulate matter (PM) in four classes (a) PM_{10} (b) $PM_{2.5}$ (c) black carbon (BC) and (d) organic carbon (OC), SO_2 , nitrogen oxides (NO_x), carbon monoxide (CO), non-methane volatile organic compounds (NMVOCs), and carbon dioxide (CO_2). We also projected the emission loads to year 2030 under a business as usual scenario after accounting for policy interventions already under implementation (e.g. accelerated introduction of liquified petroleum gas (LPG) for cooking). The emissions inventory developed under the APnA programme is the first of its kind for most of them. The main challenge is that useful and formatted data necessary to build an emissions inventory is hard to come by. We thus accessed data from a wide array of sources for each sector.

We built the emissions inventory at a resolution of 0.01° (approximately 1-km). This includes anthropogenic sources, such as transport (road, rail, ship, and aviation), large scale power generation (from coal, diesel, and gas power plants), small scale power generation (from diesel generator sets for household use, commercial use, and agricultural water pumping), small and medium scale industries, dust (road resuspension and construction), domestic (cooking, heating, and lighting), open waste burning, and open fires and non-anthropogenic sources, such as (where relevant) sea salt, dust storms, biogenic, and lighting. Using geospatial platforms, the inventory is formatted such that it can be used for urban and regional chemical transport modeling.

We use the SIM-air (Simple Interactive Models for better air quality) family of tools to estimate total emissions. This is a function of fuel burnt which is then converted to emission loads using relevant emission factors. We used the same methodology for similar projects for 10 Indian cities – Pune, Chennai, Ahmedabad, Indore, Surat, Rajkot, Hyderabad, Chennai, Vishakhapatnam, and Delhi

(Guttikunda and Jawahar, 2012; Guttikunda and Calori, 2013; Guttikunda and Kopakka, 2014; Guttikunda et al., 2014); national transport sector (Guttikunda and Mohan, 2014); national power plant sector (Guttikunda and Jawahar, 2014); and Delhi transport sector (Goel and Guttikunda, 2015). We used multiple sources to collate a library of emission factors for transport, industrial, and domestic sectors (CPCB, 2011; Pandey et al., 2014; Sadavarte and Venkataraman, 2014; IIASA, 2015; Goel and Guttikunda, 2015; Sakar et al., 2016). For this purpose, we documented and digitized information available from the annual reports, maps, and databases such as census and ambient measurements. A copy of the India specific emission factors for all the sectors from IIASA (2015), by fuel and by activity, is available online (registration required).

Data to build the emissions inventory includes;

- Static Data CPCB, state PCBs, Census Bureau, Niti Aayog, National Sample Survey Office (NSSO), Ministry of Road Transport and Highways (MoRTH), Society of Indian Automobiles Manufacturers (SIAM), Directorate general of Civil Aviation (DGCA), Ministry of Petroleum and Natural Gas (MoPNG), Ministry of Statistics and Program Implementation, Annual Survey of Industries (ASI), Central Electrical Authority (CEA), National Power Program (NPP), Cement Manufacturers Association (CMA), and Municipalities.
- Dynamic inputs NASA satellite feeds on open fires, dust events, and lightning, 3D meteorological feeds processed through Weather Research and Forecasting (WRF) model, traffic speed maps (a paid service from google), weekday and weekend profiles for transport sector (pre-decided based on speed profiles), power demand and consumption rates from the load dispatch centers, and annual/seasonal reports from various sectors listed above.
- Monitoring data: monitoring data from official and unofficial networks to evaluate model performance.
- Google: We used a paid API service from Google to map various establishments in the city hotels, hospitals, restaurants, bus stops, train stops, traffic lights, fuel stations, cinema halls, residential complexes, institutions, banks, bars, cafes, worship places, funeral homes, markets, and parks. We then used the data as layers to spatially allocate estimated total emissions to 1-km x 1-km grids.

A detailed list of these resources is available @ India-APnA (2017). Table 5 summarizes resources and methodologies applied by sector. We did not conduct any primary surveys in any of the cities. Table 6 lists the key statistics used to create the emissions inventory.

4.2. Urban emissions inventory

Table 7 is a summary of the emissions inventory for the 20 urban airsheds in India. Projections to 2030 under the business as usual scenario are influenced by the city's social, economic, landuse, urban, and industrial layout and hence the projected (increasing and decreasing) rates that we assume in this study are an estimate only. We based the vehicle growth rate on the sales projection numbers from SIAM; industrial growth on the gross domestic product of the state; domestic sector, construction activities, brick demand, diesel usage in the generator sets, and open waste burning on population growth rates from Census-India (2012). We used these estimates to evaluate the trend in the total emissions and their likely impact on ambient PM_{2.5} concentrations through 2030. Fig. 3(b) shows the total PM_{2.5} emissions between 2015 and 2030 under the business as usual scenario for Agra, Bengaluru, Chennai, and Patna. Figures for all the 20 cities are presented in the Supplementary and more details are available for each pollutant at India-APnA (2017). The totals summarized in Table 7 and Fig. 3(b), do not include natural source emissions like seasalt and dust storms.

The main emission sources for all cities include vehicle exhaust, on-road resuspended dust, construction dust, industrial exhaust, and domestic cooking and heating. While the methodology and the main data resources applied for each of the cities are the same, the final inputs for a city's emissions inventory are customized for each city. For example, of the 20 cities; Chennai and Nagpur airsheds host a power generation unit; Raipur's airshed hosts India's largest integrated steel manufacturing unit with co-generation; small and medium scale industries are major emission sources in Tier 2 cities like Amritsar, Bhopal, Chandigarh, Coimbatore, Kanpur, Ludhiana, and Pune; Chennai and Kochi host India's large commercial ports. Similarly, the meteorology is different for cities in the North and cities in the South - thus affecting space and water heating patterns. The spread of brick kilns in and around each city is different - accordingly we have mapped the location of these for individual cities and spatially allocated emissions. Data from google maps is valuable in calculating average vehicle speeds for each city (e.g. vehicle speeds in a megacity like Chennai are different from smaller cities).

Industries, dust (due to movement of the vehicles on the roads), and vehicle exhaust are the major sources of PM_{10} and $PM_{2.5}$, other sources include residential fuel combustion, garbage burning, construction activities, and generator usage at commercial and residential locations. Industries contribute the largest share of emissions for SO₂, NO_x and VOCs. For CO, the share of emissions is spread equally between vehicle exhaust, domestic cooking and heating, and industries. The level of uncertainty for the emissions inventory is about 20–30% (Guttikunda and Jawahar, 2012).

We used multiple layers of spatial proxies to grid the estimated emissions and create a gridded inventory (Fig. 4). The details of the methodology are described in Guttikunda and Jawahar (2012). We have created the emissions inventory on a GIS platform of 0.01° spatial resolution. This resolution lends itself well to atmospheric chemical transport modeling. Certain sectors such as domestic heating change daily as per meteorological conditions or forest fires that are dynamically updated every day. Fig. 3(c) shows maps of 4 cities with gridded total PM_{2.5} emissions for 2015. PM_{2.5} gridded emission maps for all the 20 cities are presented in the Supplementary. Emission hotspots in each city - highways, high population density areas, brick kiln clusters, airport, and large point sources - clearly show up.

Summary of methodologies and resources utilized for building the emissions inventory for Indian cities under the APnA city program (India-APnA, 2017).

| Sector | Methodology and data resources |
|----------------------------------|--|
| Meteorology | Weather Research and Forecasting (WRF) model with global inputs from NCEP (2016) was used to build all the necessar 3-dimensional meteorological fields, such as wind speeds, wind directions, temperature, relative humidity, pressure, precipitation, mixing layer heights, and surface threshold velocities (and others), at 0.01° spatial resolution and 1-h temporal resolution for the year 2015. A summary of the meteorological fields for the 20 cities as graphs and data files (in available in the Complementary) |
| Population | excel csv format) is available in the Supplementary. Census-India (2012) database at the district level and gridded population from GRUMP (2015) (available at 30-s spatiar resolution) was used to create 0.01° resolution population databases for the cities. At the national scale, the population database was also projected to 2030, using state level birth rates and death rates (by age group). The built-up area information in Fig. 3 was used to designate the grids and the gridded population as urban and rural. |
| On-road transport | ASIF principles (Schipper et al., 2000) - total travel activity (A), modal shares (S) in vehicle-km traveled per day, moda energy intensity (I) representing energy use per kilometer and an emission factor (F) defined as the mass emitted per vehicle-km traveled - were used to calculate the on-road vehicle exhaust emissions. The total vehicle exhaust emission are available by vehicle type, vehicle age, and fuel type and calculated following the methodology discussed in Guttikunda and Mohan (2014) for Indian states and districts. A database of average emissions factors for current and projected fleets is available as Supplementary material in Goel and Guttikunda (2014). Annual reports of MoRTH publis the number of registered motor vehicles under various categories by state and by city, for all major vehicle types – heav duty vehicles (HDVs), light duty vehicles (LDVs), buses (in city and long distance), 4-wheelers (cars, jeeps, sports utilit vehicles, taxis), 3-wheelers (passenger and freight), and 2-wheelers (scooters, mopeds, and motorcycles). Monthly report by SIAM publish national vehicle sales for major categories. |
| On-road dust | Methodologies detailed in USEPA (2006) and Kupiainen (2007) were utilized to build a base on-road dust resuspensio inventory. In addition to this, the precipitation fields after the WRF meteorological model processing were utilized to adjust for soil moisture in the grids. If the grids had precipitation of $> 1 \text{ mm/h}$, then the on-road dust resuspension i zeroed and maintained at minimum levels for the next 2 h. The grid-level speed maps from google maps (pre-defined was also used to adjust on-road resuspension, when the average (cumulative) speed in the grid is under 5 km/h – an indication of congestion and speed under the threshold to support resuspension. These corrections were introduced to reduce any overestimation. |
| Aviation | Flight landing and takeoff (LTO) statistics are obtained from DGCA and http://flightstats.com with information availabl by airport. Assuming average occupancy per flight of approximately 180, we estimated the idling emissions of cars dropping off and picking up passengers and emissions from general airport operations (luggage handling, fueling, shuttling of passengers, cargo movement, and other support activities) supported by cars, SUVs, buses, fuel trucks, taxin vehicles, and tractors, often using diesel as fuel. The average LTO emission rates are obtained from IIASA (2015) and or road fleet average emission rates from Goel and Guttikunda (2014). |
| Shipping | To emissions from domestic and cargo handling at the ports and along the shipping routes intersecting the selected airsheds (for the coastal cities) are obtained from the Task Force on Hemispheric Transport of Air Pollution (TF HTAP http://htap.org). No primary inventory was developed for this sector. |
| Domestic cooking | The HH10 database of Census-India (2012) for all 640 districts includes nine fuel categories (crop residue, wood, coa cow dung, kerosene, coal and charcoal, LPG, biogas, electric, and a small share of others), urban-rural fuel usage profiles and inside-outside cooking preferences. The electric share of households is taken as zero emissions. The survey record only the primary fuel for cooking. In the calculations, the share of the fuel preferences at the district level is used as proxy. Average energy consumed by household for cooking is calculated based on NSSO survey database, which lists amount of food varieties cooked at the state level (Pandey et al., 2014) which is assumed constant for all the districts within a state. Gridded population (discussed above), landuse information to urban-rural classification within the grid and district, and emission factors from IIASA (2015) were overlaid for gridded emissions from household solid, liquid and gaseous fuel consumption for cooking, Cooking emissions are constant for all days in a year. |
| Domestic lighting | The HH7 database of Census-India (2012) for all 640 districts includes four fuel categories (kerosene, solar, electricity and a small share of others) and urban-rural fuel usage profiles. The solar and electric share of households is taken as zer emissions. The survey records only the primary fuel for lighting. Similar to the cooking methodology, gridded populatio (discussed above), landuse information to urban-rural classification within the grid and district, and emission factors from IIASA (2015) were overlaid for gridded emissions from household liquid fuel consumption for lighting. Lighting emissions are constant for all days in a year. |
| Domestic space and water heating | In addition to cooking food and lighting, water and space heating emissions for the domestic sector were estimated after linking the surface temperature profiles at the grid level and the population by age groups. This was used to improve the emissions in two ways (a) no more approximation of the heating needs by north or south or summer or winter. A runnin average temperature was tracked to trigger the need for space heating needs (b) it is our assumption that hot water requirement for age groups under 15 and over 55 years is all year round, and rest of the age with varying usage. The water heating and space heating emissions vary daily. |
| Open waste burning | Global trash burning database is available at 0.1° x 0.1° using a methodology involving waste generation rates, waste mis collection efficiencies, open waste burning rates, and emission factors (Wiedinmyer et al., 2014; IIASA, 2015). Total waste generation in India is approximately 150,000 tons per day at the rate of 0.5 kg per day per person, of which w estimate approximately 25% is burnt in the urban areas and more in the rural areas with no waste management system i place (Annepu, 2012). Because of smoke, air pollution, and odor complaints, the local authorities have banned open waste burning, but it continues unabated at makeshift landfills. Linked to the built-up area map for the airshed (Fig. 3 and notes from the municipal reports on local waste management activities, emissions were estimated assuming that th urban areas have higher waste collection efficiency. The modeled grid level precipitation data, a threshold of 1 mm/h i used to zero the open waste burning emissions. |

Table 5 (continued)

| Sector | Methodology and data resources |
|-----------------------|---|
| Power plants | The database of power plants includes geographical location, number of boiler units, coal characteristics, coal consumption rates, and installed control equipment. These details were documented from their respective environmental impact assessment reports and data published by the state electricity boards (public entities) and private operators (Guttikunda and Jawahar, 2014; NPP, 2018). Under NAMP, all the large point sources such as the coal-fired thermal power plants, are required to conduct continuous emissions monitoring from all the operational stacks, for all the criteria pollutants. However, these emission rates are not public and only reports based on intermittent audits are presented as averages (MoEFCC, 2010; CEA, 2013; MoEFCC, 2015; CEA, 2017; MoEFCC, 2018). A summary of the emission factors and an analysis of current and planned power plants is presented in Lu et al. (2013), Guttikunda and Jawahar (2014), Sadavarte and Venkataraman (2014), and IIASA (2015). |
| Diesel generator sets | The spatial allocation of the in-situ emissions from burning of diesel in small and large generator sets was refined using the hotspot information available as geospatial layers, including the locations of hotels, hospitals, malls, markets, funeral homes, religious worship centers, industrial areas, apartment complexes, commercial centers, and telecom towers. These layers were extracted from open sources such as Open Street Maps and paid sources such as google API. The shortages in the power supply at the state and the grid level were sourced from CEA reports, NPP statistics, and load dispatch center reports. Emission factors by generator set size are summarized in Sahu et al. (2015). |
| Heavy industries | Other than the thermal power plants, details of heavy industries covering fertilizers, cement, refineries, iron and steel, and mineral ore processing were collated using information from respective sector annual reports, http://www. indiastats.com, and emission factors from Sadavarte and Venkataraman (2014) and IIASA (2015). The plant details include location information and production rates. |
| Light industries | The heavy industries is further substantiated with the industrial fuel consumption data (as solid, liquid, and gaseous fuels) from the Ministry of Statistics, conducted as part of the annual survey of industries (ASI, 2015; MSME, 2017). Broad categories covered under this survey are food processing, textile works, leather works, wood processing, paper and ink, coke products, pharmaceuticals, rubber and plastics works, glass works, manufacturing & repairs, power generation, and waste & water treatment. The fuel consumption data by various industrial categories is based on the fuel purchase receipts at the district level and the emission factors database is from Sadavarte and Venkataraman (2014) and IIASA (2015). |
| Construction | Besides the traditional manufacturing industries, construction is among the fastest growing sectors in India. The demand for traditional red and fired clay bricks is high, along with cement, and the location of these clusters in and around the urban airsheds were tagged using the open google earth imagery. Most of them are located along the river, with easy access to top soil and water for raw material. Traditionally, the rectangle shaped clay bricks are sun dried and readied for firing in "clamps" - a pile of bricks with intermittent layers of sealing mud and fuel. Technology for firing has changed for mass production with the fixed chimney brick kilns having a capacity to produce 10,000 to 20,000 bricks per day (World- Bank, 2010; Mathiel et al., 2012; Guttikunda et al., 2013; Rajarathnam et al., 2014). This fuel would vary from agricultural waste to biofuels like cow dung and wood to fossil fuels like coal and heavy fuel oil. Emission factors for various technologies were summarized from Weyant et al. (2014), IIASA (2015), Akinshipe and Kornelius (2018). The inventory also includes fugitive dust estimates for construction sites based on empirical functions proposed in USEPA (2006) and operational guidelines from CPCB (2017). |
| Open fires | The location of open fires (from VIIRS and MODIS satellite feeds @ https://worldview.earthdata.nasa.gov), along with pixelated information of landuse fractions (agricultural, forest, urban, water, and arid) were used to estimate the total emissions. The Fire INventory from NCAR (FINN) model estimates total emissions using these satellite feeds, provide emissions information suitable for chemical transport modeling at a horizontal resolution of 1 sq.km. The global archives dating back to 2002, along with a database of landuse based emission factors, and data processing procedures to prepare model ready emission inventory are available @ http://bai.acom.ucar.edu/Data/fire (Akagi et al., 2011; Wiedinmyer et al., 2011). |
| Dust storms | The GOCART dust module from WRF simulates the dust outbreaks and also propagates these emissions in the CAMx dispersion model as a separate source. |
| Seasalt | A CAMx pre-processor calculates seasalt emissions, where applicable, using the 3-dimensional meteorological input for WRF. |

5. Modeled particulate sector contributions

5.1. Chemical transport model

We used the Comprehensive Air Quality Model with Extensions (CAMx), to estimate particulate concentrations for each urban airshed. CAMx is an open-source Eulerian photochemical dispersion model that supports (a) 3-dimensional advection linked to 3-dimensional meteorological data at the grid level, including plume rise calculations for point sources (b) scavenging schematics in the form of dry and wet deposition (c) multiple chemical mechanisms to characterize photochemistry (d) gas to aerosol conversions (from SO₂ to sulfates, NO_x to nitrates, and VOCs to secondary organic aerosols) (e) links to online emission calculations for certain sources like seasalt and (f) modular processing of area and point sources to estimate contributions by pre-defined region and source. We use the WRF meteorological model to derive meteorological data (3D wind, temperature, pressure, relative humidity, and precipitation fields) (data files are presented in the Supplementary). The vertical resolution of the model was stretched over 28 layers under 12 km, to improve our simulations of ground level concentrations. The lowest layer is designated at 30 m and there are 12 layers within 1 km.

To account for activities outside the designated urban airshed, we use boundary conditions from a model run covering the Indian Subcontinent (details on national emissions inventory and modeling framework are available online @ http://www.indiaairquality. info). We used the MOZART global model to get boundary conditions for the Indian Subcontinent. CAMx modeling system (http://

| Key l | household, | geography, | and | transport | statistics | for | 20 | Indian | cities. |
|-------|------------|------------|-----|-----------|------------|-----|----|--------|---------|
| | | | | | | | | | |

| | City name | А | В | С | D | E | F | G | Н | Ι | J | Κ | L |
|----|-------------|------|------|------|------|------|------|----------|-----|------|-----|-----|------|
| 1 | Agra | 36.2 | 6.7 | 37.4 | 79.7 | 91.1 | 33.2 | high | _ | 800 | - | _ | 0.90 |
| 2 | Amritsar | 46.9 | 11.7 | 58.3 | 96.0 | 85.8 | 26.8 | high | 120 | 300 | - | - | 0.93 |
| 3 | Bengaluru | 44.3 | 17.5 | 75.5 | 98.0 | 93.3 | 40.2 | low | 700 | 6000 | - | - | 5.56 |
| 4 | Bhopal | 43.7 | 12.4 | 57.8 | 92.9 | 74.2 | 36.2 | medium | - | 400 | - | - | 1.10 |
| 5 | Bhubaneswar | 33.8 | 6.8 | 32.8 | 71.5 | 63.8 | 37.0 | medium | 160 | 300 | yes | - | 1.30 |
| 6 | Chandigarh | 46.7 | 25.7 | 71.6 | 98.4 | 94.2 | 40.7 | high | 160 | 600 | yes | - | 0.80 |
| 7 | Chennai | 46.6 | 13.2 | 82.4 | 99.1 | 94.1 | 41.0 | very low | 430 | 2400 | yes | yes | 4.93 |
| 8 | Coimbatore | 47.1 | 9.2 | 71.5 | 94.8 | 81.8 | 36.2 | low | 120 | 6000 | yes | - | 1.90 |
| 9 | Dehradun | 47.1 | 14.9 | 72.4 | 96.3 | 90.4 | 26.5 | high | 10 | 750 | - | - | 0.50 |
| 10 | Indore | 46.6 | 10.5 | 67.8 | 96.5 | 78.2 | 30.2 | medium | 110 | 900 | yes | - | 1.71 |
| 11 | Jaipur | 45.7 | 12.5 | 49.8 | 86.4 | 90.2 | 26.6 | high | 200 | 1600 | yes | - | 2.25 |
| 12 | Kanpur | 30.3 | 6.2 | 51.5 | 63.0 | 69.7 | 35.2 | high | 125 | 1200 | yes | - | 1.46 |
| 13 | Kochi | 40.7 | 17.1 | 63.2 | 97.4 | 93.7 | 6.1 | very low | 260 | 400 | yes | yes | 0.60 |
| 14 | Ludhiana | 50.4 | 16.8 | 68.3 | 98.3 | 93.4 | 29.4 | high | 190 | 3000 | yes | - | 1.77 |
| 15 | Nagpur | 42.8 | 7.2 | 60.2 | 92.1 | 69.4 | 30.0 | medium | 130 | 900 | yes | - | 1.30 |
| 16 | Patna | 18.4 | 5.2 | 33.8 | 57.1 | 70.0 | 30.1 | high | 310 | 450 | - | - | 1.00 |
| 17 | Pune | 48.8 | 13.2 | 67.9 | 92.7 | 87.8 | 46.5 | medium | 150 | 3000 | yes | - | 2.34 |
| 18 | Raipur | 20.6 | 3.6 | 19.4 | 84.6 | 47.5 | 38.7 | medium | 120 | 800 | yes | - | 1.80 |
| 19 | Ranchi | 25.3 | 6.1 | 29.6 | 63.0 | 47.7 | 19.2 | high | 110 | 100 | - | - | 0.60 |
| 20 | Varanasi | 29.1 | 4.5 | 37.4 | 62.0 | 77.0 | 28.5 | high | 450 | 200 | - | - | 0.77 |

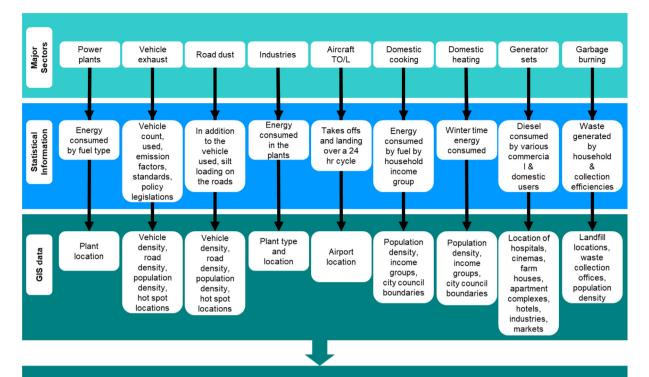
Note: A to F statistics are extracted from Census-India (2012). A = % households with at least one motorcycle; B = % households with at least on 4wheeler (cars and vans); C = % households with primary cooking supported by gas and electricity; D = % households with primary lighting supported by electricity; E = % households with permanent houses; F = % households with < 2 rooms; G = need for winter heating; H = number of mapped brick kilns in the city's selected airshed from open google earth imagery; I = number of small scale industries listed under ASI (2015); J = existence of heavy industries (including power plants) in the city's selected airshed; K = coastal or not; L = total number of registered vehicles in 2015 in millions (MoRTH, 2016).

Table 7 Estimated annual emissions totals for the 20 Indian cities in 2015.

Annual emissions (CO2 in million tons/year and rest in tons/year) со NMVOC City name PM_{2.5} PM10 BC OC NOv SO₂ CO₂ 1 Agra 8350 15,600 2100 2950 10,250 79,250 16,400 950 1.55 2 Amritsar 8600 13,450 2000 2600 12.050 16.500 2400 72.100 2.4 3 Bengaluru 31,300 67,100 9350 8450 56,900 335,550 83,500 5300 10.42 4 Bhopal 5900 13.650 1300 1550 8950 54.300 15.100 900 1.65 5 Bhubaneswar 11,250 22,400 2700 3600 22,250 129,050 27,350 1350 2.28 6 Chandigarh 18.300 33,600 3450 4850 56.300 160.350 35.850 3200 3.91 7 Chennai 94,050 132.350 14,500 11,250 207,400 345,550 92,900 26,600 9.76 Coimbatore 14,100 23,900 3500 3050 25,150 113,100 28,200 2650 8 3.13 9 Dehradun 4650 7500 1300 1700 7200 39,450 8600 750 1.04 3250 10 Indore 11,850 27,600 3150 15,200 83,700 22.250 1400 2.72 17.200 34,800 4700 5250 20,400 31.550 2650 3 7 9 11 Jaipur 137.600 43,900 5550 24,750 12 Kanpur 34,550 5150 166,000 34,100 2450 2.39 13 Kochi 9150 16,400 2250 2100 63,900 69.550 14,850 20,900 2.4 14 Ludhiana 14,500 24,200 4550 3850 31,200 128,550 25,900 3750 3.53 86,300 6900 7900 15 Nagpur 67.100 128.100 176.850 52,900 9600 6.89 Patna 18,050 29,500 4800 6150 18,350 171,450 27,650 3850 2.86 16 5600 4150 3950 17 Pune 17,700 36.900 37.000 166.050 41,900 5.81 18 Raipur 41,500 59,650 9150 9050 60,700 163,300 118,150 7600 3.13 19 13,150 24,300 3350 4550 13,150 128,500 25,300 1950 2.31 Ranchi 20 Varanasi 12.100 17,450 3050 4850 14,050 134,400 21.850 2300 1.79

www.camx.com) has a pre-processor module to link MOZART model results. The WRF-CAMx modeling system, boundary conditions, and the national emissions inventory for ozone precursor pollutants was evaluated against ground based and satellite observations in Sarkar et al., 2016.

The total PM concentrations includes primary aerosols (metals, BC, and OC) and secondary contributions (from chemical conversion of gaseous SO_2 to sulfate aerosols and gaseous NO_x to nitrate aerosols). In CAMx, the primary PM is modeled in two bins to account for differential deposition and advection characteristics – coarse fraction (CPRM) comprises of PM_{10} to $PM_{2.5}$ mass and fine fraction (FPRM) comprises of everything less than $PM_{2.5}$ mass. All the secondary aerosols are considered as part of $PM_{2.5}$ mass. We set up the dispersion model calculations to also estimate sector contributions to ambient $PM_{2.5}$ concentrations in the urban airshed of each city.



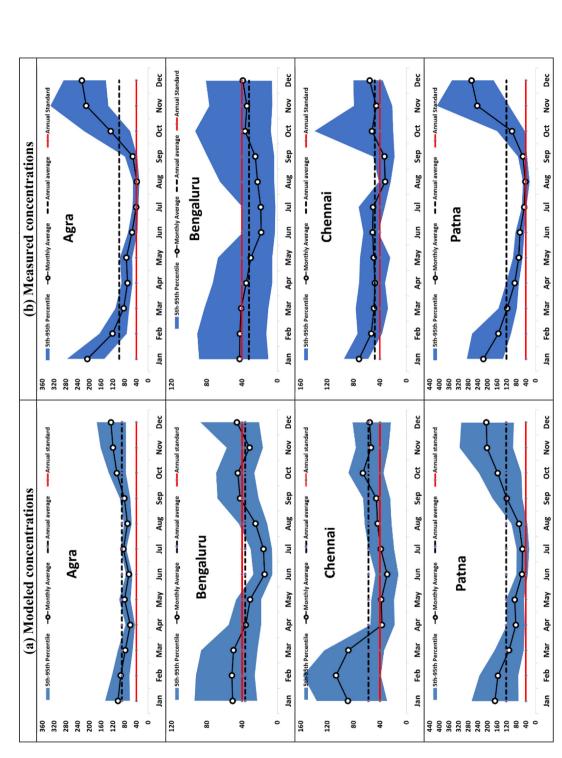
Emissions @ 0.01° grid resolution for PM, SO₂, NO_x, CO, & NMVOC

Fig. 4. Multiple geospatial layers and data sources accessed to spatially distribute the annual emissions to 0.01° grids over the urban airshed selected for 20 Indian cities.

Table 8

Comparison of (A) WRF-CAMx modeled $PM_{2.5}$ annual average (for the urban parts of the airshed) (B) Measured annual average concentrations and standard deviation of all the available continuous monitoring data (in brackets – number of continuous monitoring stations operational in the city). (C) satellite-derived $PM_{2.5}$ annual average concentration from van Donkelaar et al. (2016) (all values in $\mu g/m^3$).

| | City name | Modeled | A - Measured (continuous stations only) | B – 2016 satellite-derived from Fig. 1 |
|----|-------------|------------------|---|--|
| 1 | Agra | 89.0 ± 27.1 | 98.6 ± 86.1 (1) | 116.9 |
| 2 | Amritsar | 83.4 ± 25.4 | 65.9 ± 64.8 (1) | 79.1 |
| 3 | Bengaluru | 36.5 ± 19.7 | 32.3 ± 24.2 (3) | 28.9 |
| 4 | Bhopal | 49.9 ± 16.0 | 47.8 ± 35.1 (1) | 56.5 |
| 5 | Bhubaneswar | 47.7 ± 26.5 | | 37.9 |
| 6 | Chandigarh | 58.1 ± 18.4 | 59.4 ± 62.7 (1) | 78.7 |
| 7 | Chennai | 57.5 ± 31.3 | 47.7 ± 73.7 (3) | 32.6 |
| 8 | Coimbatore | 19.4 ± 9.5 | | 26.0 |
| 9 | Dehradun | 51.2 ± 17.6 | | 87.4 |
| 10 | Indore | 66.3 ± 31.1 | 40.4 ± 23.1 (1) | 46.1 |
| 11 | Jaipur | 99.6 ± 29.5 | 70.4 ± 76.8 (3) | 68.3 |
| 12 | Kanpur | 114.1 ± 44.4 | $111.4 \pm 79.7 (1)$ | 106.6 |
| 13 | Kochi | 29.1 ± 15.7 | | 26.1 |
| 14 | Ludhiana | 90.2 ± 33.7 | 74.4 ± 46.1 (1) | 83.9 |
| 15 | Nagpur | 84.9 ± 40.8 | 53.9 ± 26.7 (1) | 47.3 |
| 16 | Patna | 122.2 ± 68.2 | $121.3 \pm 89.5 (1)$ | 90.8 |
| 17 | Pune | 56.3 ± 28.4 | 43.1 ± 25.7 (1) | 47.5 |
| 18 | Raipur | 82.3 ± 43.1 | | 54.6 |
| 19 | Ranchi | 73.0 ± 36.7 | | 55.6 |
| 20 | Varanasi | 78.4 ± 26.9 | $106.2 \pm 83.4 (1)$ | 94.0 |



5.2. Modeled particulate concentrations

Table 8 summarizes the modeled annual average $PM_{2.5}$ concentrations. We also present PM concentration maps in Fig. 3(d) for Agra, Bengaluru, Chennai, and Patna. Concentration maps for all the 20 cities are presented in the Supplementary. Except for Coimbatore and Bengaluru (both from Southern India), all the urban airsheds exceed the national annual standard of $40 \mu g/m^3$ for $PM_{2.5}$.

In case of $PM_{2.5}$, only 14 cities have at least one station operational. This data is summarized as measured in Table 8, along with the number of monitoring stations operational in each city. A direct comparison of the modeled and measured concentration is not available, due to limited number of (often one) operational monitors. Even in case of Bengaluru, while there are 10 operational continuous monitoring stations, 5 stations came online only in 2018 with no data in the public domain and of the remaining 5, only 3 report $PM_{2.5}$ concentrations. The average and standard deviation value for the modeled concentrations in the table represents data for all the urban grids and 12 months. Fig. 5(a) shows the dispersion model results by month to highlight the seasonality of air quality over the course of a year for 4 cities and rest are presented in the Supplementary. The daily average variation by month for all the data from the continuous monitoring stations (1 for Agra and Patna and 3 for Bengaluru and Chennai) is presented in Fig. 5(b). The monthly average maps are not presented in this paper and the data is available upon request. The modeled graphs present the variation (5th and 95th percentile) in the $PM_{2.5}$ monthly average concentrations among the urban grids, along with the annual average concentration.

We capture the quantitative ranges and the seasonality in the model results and note that the sample size is not enough to make any scientific conclusions. There is also lack of information on the representativeness of the location of the continuous stations. Often the first continuous monitoring station is located near the pollution control board for safety and maintenance reasons. As more stations come online in each of these cities, we will revisit the modeling results to further explore the trends and to assess the overall assumptions in our calculations. At this stage, we are expecting the emissions and dispersion modeling results to start a much-needed conversation in the policy circles to examine the hot spots in the city and to push for an increase in the monitoring stations to spatially and temporally represent the pollution levels in the city.

For all the cities, variation in the 24-hr concentration values is large, with the general trend pointing to higher levels of pollution during the winter months and lower levels during the summer and monsoon months, which coincides with the trend in the meteorological fields – lower mixing heights, lower wind speeds, and lower precipitation rates during the winter months and vice versa during the summer months. This seasonality is strongly associated with the mixing heights (summary tables are presented in the Supplementary). The variation is very high for the cities on the Indo-Gangetic plain when compared to the 4 Southern cities (Bengaluru, Chennai, Coimbatore, and Kochi), which experience a uniform mixing height. For example, on average, Bengaluru's winter time heights are at least 2 times higher than those reported for Northern cities, which means on average there is at least double the volume of air to disperse emissions and consequently result in better air quality. While meteorology is crucial in determining the levels of air pollution in any city, it cannot not be part of the control strategy to reduce pollution.

5.3. Source contributions

For all cities, industries, vehicle exhaust, domestic cooking and heating, open waste burning, diesel generator sets, road dust, and construction dust are the primary sources of ambient $PM_{2.5}$ pollution, apart from the regional contributions. This list is not an unknown. With the changing emission patterns, it has become a prerequisite to ascertain the source contributions pertinent to a city's geographical and economic conditions, to prioritize a series of control options for both regulatory and health purposes (Sagar et al., 2016). We summarize the modeled percent contributions of these sources to 20 Indian cities in Table 9 and Fig. 3(e).

We estimate these contributions using a bottom-up technique following the sequence of building an emissions inventory, using WRF to process meteorological data, and using CAMx for dispersion modeling. Ideally, this process should be accompanied by a topdown technique where filter samples are collected at representative locations in the city for representative seasons, followed by chemical analysis for metals, ions, and carbon species and statistical receptor modeling with the help of pre-defined library of source profiles. CPCB (2011) using the samples and surveys collected in 2006 was the last study published for Delhi, Bengaluru, Chennai, Pune, Kanpur, and Mumbai. In the past 10 years, these 6 cities and many from across India have at least doubled their vehicle fleet and increased their demand for construction material, electricity, and other industrial products, and consequently air pollution (Fig. 1). While top-down methods ascertain the chemical signature of a fuel in a given sample of air, they fail to pin point the source itself, such as diesel consumed in a truck versus a generator set, or if the source of pollution is local or regional. This gap is best addressed in the bottom-up technique, accompanied with the boundary conditions to account for outside contributions and the emission inventory providing the details at the source level (by fuel type, by vehicle type, by region or grid, and by season).

GBD-MAPS (2018) is the recent study to model source contributions at the national level, highlighting the same sources. Domestic cooking and heating was reported as the largest contributor (24%), which is double the percentages observed in the cities. This primarily points to the fact that larger proportions of population are using biomass and coal as their primary fuel for cooking and heating in the rural areas. In the global model simulations at coarser resolutions, the national shares tend to also get diluted like in case of transport with 2% (with most of these emissions concentrated over a very small percentage of grids in the modeling domain) as compared to an average 18.0% in the study. While the national scale simulations raise the awareness of the problem at the central level, the urban scale simulations highlight the shares of the local sources and the importance of regional coordination for effective management of the problem.

The regional contributions are an important source of pollution (from a low of 13-15% in case of Chennai and Bengaluru to

Annual average source contributions to WRF-CAMx modeled $PM_{2.5}$ concentrations for 20 Indian cities in 2015 (A = domestic cooking, heating, and lighting; B = transport including road, rail, shipping, and aviation; C = dust including on-road resuspension and construction; D = brick kilns; E = all industries excluding brick kilns and including thermal power plants; F = open waste burning; G = diesel generator sets; H = natural source for coastal cities only; Outside = contribution of boundary conditions to the urban airshed, representing the influence of regional sources).

| | | Modeled contributions | | | | | | | | |
|----|-----------------|-----------------------|------------------|------------------|-----------------|------------|------------------|-----------------|-------|-------------------|
| | City name | А | В | С | D | E | F | G | Н | Outside |
| 1 | Agra | 23.8% | 13.9% | 10.7% | 0.0% | 0.2% | 12.4% | 2.7% | | 35.9% |
| 2 | Amritsar | 10.6% | 10.5% | 7.1% | 2.1% | 7.3% | 6.1% | 3.1% | | 52.7% |
| 3 | Bengaluru | 9.8% | 26.5% | 23.0% | 2.5% | 2.1% | 16.1% | 4.0% | | 15.6% |
| 4 | Bhopal | 10.2% | 14.1% | 17.1% | 0.0% | 2.8% | 8.7% | 4.9% | | 41.8% |
| 5 | Bhubaneswar | 15.9% | 17.0% | 20.8% | 4.0% | 0.6% | 5.7% | 3.6% | | 32.0% |
| 6 | Chandigarh | 11.4% | 10.6% | 12.6% | 1.3% | 1.3% | 8.9% | 2.6% | | 50.8% |
| 7 | Chennai | 3.6% | 24.5% | 23.5% | 3.1% | 12.8% | 15.5% | 1.6% | 1.8% | 13.3% |
| 8 | Coimbatore | 6.4% | 18.3% | 13.7% | 1.0% | 11.1% | 14.1% | 2.4% | | 32.5% |
| 9 | Dehradun | 14.3% | 14.2% | 4.4% | 0.4% | 1.3% | 19.6% | 3.8% | | 41.7% |
| 10 | Indore | 8.1% | 26.9% | 22.7% | 2.0% | 2.4% | 7.8% | 2.0% | | 27.8% |
| 11 | Jaipur | 13.4% | 24.1% | 17.5% | 1.7% | 2.4% | 8.4% | 2.2% | | 29.9% |
| 12 | Kanpur | 33.8% | 13.7% | 8.9% | 1.2% | 6.5% | 8.8% | 4.1% | | 22.5% |
| 13 | Kochi | 9.5% | 20.2% | 16.3% | 3.8% | 4.1% | 3.8% | 4.5% | 16.5% | 20.9% |
| 14 | Ludhiana | 7.8% | 16.3% | 12.3% | 2.8% | 7.9% | 9.2% | 2.6% | | 40.7% |
| 15 | Nagpur | 6.8% | 17.2% | 10.9% | 3.2% | 26.7% | 11.6% | 1.8% | | 21.4% |
| 16 | Patna | 14.6% | 14.8% | 12.1% | 9.3% | 11.2% | 12.9% | 5.4% | | 19.2% |
| 17 | Pune | 5.8% | 24.0% | 23.4% | 2.6% | 9.8% | 6.4% | 2.8% | | 24.7% |
| 18 | Raipur | 11.8% | 17.2% | 11.5% | 1.4% | 22.8% | 6.2% | 2.8% | | 25.8% |
| 19 | Ranchi | 18.0% | 21.1% | 14.1% | 3.2% | 1.1% | 12.2% | 1.3% | | 28.5% |
| 20 | Varanasi | 20.9% | 13.5% | 8.2% | 6.1% | 0.2% | 16.2% | 3.3% | | 31.2% |
| | 20 city average | $12.9~\pm~7.1\%$ | $18.0~\pm~5.1\%$ | $14.6~\pm~5.8\%$ | $2.6~\pm~2.2\%$ | 6.8 ± 7.4% | $10.6~\pm~4.2\%$ | $3.1 \pm 1.1\%$ | | $30.5 \pm 10.9\%$ |

50-52% in Amritsar and Chandigarh). This shows that air pollution in India is not limited to cities alone and *peri* urban and rural areas also generate and contribute to pollution. This is especially true for cities on the Indo-Gangetic plain, given the density of population in the region and concentration of settlements, which makes it harder to distinguish between rural and urban areas based on modeled PM_{2.5} concentrations alone (as seen in Fig. 1). Large external contributions in Amritsar, Chandigarh, and Ludhiana from the state of Punjab, highlight their proximity to sources like power plants not located in the airsheds, seasonal dust storms from the neighboring state of Rajasthan, sources of emissions from across the Indian border, and open field burning emissions after the harvest season, which can only be addressed in a regional framework of air quality management. The cities with outside contributions between 30 and 40% are mostly Tier 2 category. This percentage will likely reduce to about 15% as the city grows and emissions from within the city increase over time.

We also projected results to 2030 under a business as usual scenario by modeling emissions, dispersion, and source apportionment. For brevity we have only included emission projection graphs in Fig. 3(b). The gridded emission maps, and modeled concentration maps, and source contributions for 2030 are available at India-APnA (2017).

We list a few emission management options for the main pollution sources, the benefits of which can only be evaluated if and when they are implemented. At the national scale, these options were evaluated as what-if scenarios in GBD-MAPS (2018) and Venkataraman et al. (2018) for the period of 2015–2050, and the benefits are clear – no emissions at the sources will result in better air quality in the future.

- Two of the 20 cities lie on the coast Kochi and Chennai. These two cities host large commercial ports, with significant contributions to PM_{2.5} and SO₂ pollution from freight movement (ships and on roads heavy duty trucks). This sector can benefit from a freight management program. For example, movement of freight on rail, restrictions on vehicle types entering the port area, restrictions on the fuel quality used by ships anchored at the ports.
- All the cities need to aggressively promote public and para transportation systems, to reduce the use of millions of passenger vehicles and their respective emissions. By 2030, the vehicle exhaust emissions are expected to remain constant or tend lower, if and only if, Bharat 6 (equivalent of Euro-6) fuel standards are introduced nationally in 2020, as recommended by the Auto Fuel Policy. A sustainable transport policy must also promote non-motorized transport (walking and cycling) infrastructure, to not only reduce the contribution of vehicle exhaust emissions but also to reduce on-road dust re-suspension.
- By 2030, the share of emissions from residential cooking and lighting is expected to decrease with an increasing share of LPG and electrification available for primary cooking under the Prathan Mantri Ujjwala Yojana scheme (MoPNG, 2018).
- All the 20 cities lack a comprehensive waste management system. The practice of open waste burning is hard to regulate and monitor. This sector contributes significantly to air pollution unless the municipalities address this internally, starting from reducing waste generation, collecting waste efficiently, and managing the collected waste.
- Upgrading the brick kiln technology from the current fixed chimney and clamp-style baking to (for example) zig-zag for the approximately 4000 brick kilns mapped in these 20 urban airsheds will improve their overall energy efficiency and reduce

emission loads (World-Bank, 2010; Maithel et al., 2012).

- Coal-fired power plants and large industries with captive power plants need to enforce stricter emission standards for all the criteria pollutants to reduce their share of influence on urban air quality. The densely populated Indo-Gangetic plain has cities and rural areas near these large point sources that strongly influences each other's ambient air quality (Guttikunda and Jawahar, 2014).
- A regional air quality management, spanning multiple states, districts, and stakeholders is necessary to reduce the pollution loads for all the cities. In April 2018, draft of the National Clean Air Programme (NCAP) is a step in the right direction designed to build institutional and technical capacity of CPCB and SPCBs "to meet the prescribed annual average ambient air quality standards at all locations in the country in a stipulated timeframe" (CPCB, 2018).

6. Policy implications of a knowledge base

In policy circles, the debate on air quality in India and how to best tackle it, is limited to metropolitan cities. Yet, India lives in its villages and smaller towns. We need to look at the state of air pollution in Tier 1 and 2 cities, quantify its extent and understand its sources. With a growth in population and ever-increasing concentration of settlements (especially in the Indo-Gangetic plain), the need for baselines and a start to tackle air pollution is urgent. In the short term, there is an immediate need to increase the ground level monitoring network, which can further support establishing air pollution trend lines for the cities and help monitor action plans.

Urban India is growing rapidly and infrastructure and management systems are yet to catchup with growing urban population, more waste generation (per capita and totals), greater share of motorized transport for individual and commercial purposes, an increase in industrial and manufacturing activity, and a growing demand for clean fuels for cooking and heating. With the lack of systems, cities resort to ad-hoc methods to deal with increased pressure on existing infrastructure. Hence cities need to start planning by anticipating the challenges they will face as they grow and be proactive about solutions to reduce air pollution.

For most of the cities in this study, this is the first time an emissions inventory has been built for the city, followed by a dispersion modeling exercise to assess the particulate pollution trend and identify source contributions. Local governments can use these numbers as a start to develop an air pollution control policy. Apart from numbers, identifying the spatial spread in pollution hotspots lends itself to immediate action. With information, cities can plan for the long-term action by simulating different scenarios that are customized to each city. Simply publishing these numbers will not reduce pollution. But in the absence of any other concrete information, results from this study can support the policy makers establish baselines and plot a roadmap to improve air quality in the long run.

While the analysis presented in this paper focused on the long-term air pollution knowledge assessment, we use the same information base for on-going short-term air quality forecasting (for 3-days) releasing data on an open platform (details @ http://www.indiaairquality.info).

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Appendix A. Supplementary data

Supplementary data associated with this article can be found in the online version, at doi: https://doi.org/10.1016/j.uclim.2018. 11.005. These data include the Google maps of the most important areas described in this article.

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