Evaluation on the effect of regional joint control measures in changing photochemical transformation: A comprehensive study of the optimization scenario analysis

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Abstract: Heavy haze usually occurs in winter in eastern China. To control the severe air pollution during the season, comprehensive regional joint-control strategies were implemented throughout a campaign. To evaluate the effectiveness of these strategies and to provide some insights into strengthening the regional joint-control mechanism, the influence of control measures on levels of air pollution were estimated with an integrated measurement-emission-modeling method. To determine the influence of meteorological conditions, and the control measures on the air quality, in a comprehensive study, the 2nd World Internet Conference was held during December 16~18, 2015 in Jiaxing City, Zhejiang Province in the Yangtze River Delta (YRD) region. We first analyzed the air quality changes during four meteorological regimes; and then compared the air pollutant concentrations before, during and after the regulation under static meteorological conditions. Next, we conducted modeling scenarios to quantify the effects caused due to the air pollution control measures. We found that total emissions of SO_2, NO_x, $PM_{2.5}$ and VOCs in Jiaxing were reduced by 56%, 58%, 64% and 80%, respectively; while total emission reductions of SO_2, NO_x, $PM_{2.5}$ and VOCs over the YRD region are estimated to be 10%, 9%, 10% and 11%, respectively. Modelling results suggest that during the campaign from December 8 to December 18, $PM_{2.5}$ daily average concentrations decreased by 10 $\mu g/m^3$ with an average decrease of 14.6%. Our implemented optimization analysis compared with previous studies also reveal that local emission reductions play a key role in air quality improvement, although it shall be supplemented by regional linkage. In terms of regional joint control, to implement pollution channel control 48 hours before the event is of most benefit in getting similar results. Therefore, it is recommended that a synergistic emission reduction plan between adjacent areas with local pollution emission reductions as the core part should be established and strengthened, and emission reduction plans for different types of pollution through a stronger regional linkage should be reserved.

Keywords: $PM_{2.5}$; regional joint control; YRD

1 Introduction

High concentrations of $PM_{2.5}$ has attracted much attention due to its impact on visibility (Pui et al., 2014), human health (West et al., 2016) and global environment. To control air pollution situation in China, the Ministry of Ecology and Environment of the People's Republic of China has released a lot of policies, which can generally be divided into long-term action plans (such as the Clean Air Action Plan (2013-2017), the Five-year Action Plans) and short-term control measures (such as Clean Air Protection during Mega Events, Air Pollution Warning and Protection Measures). China has successfully implemented some mega event air pollution control plans and ensured good air quality, including the 2008 Beijing Olympics (Kelly and Zhu, 2016); the 2010 World Expo in Shanghai (CAI-Asia, 2010); the 2010 Guangzhou Asian Games (Liu et al., 2013); the 2014 Asia-Pacific Economic Cooperation Forum (APEC) (Liang et al., 2017); 2014 Summer Youth Olympics in Nanjing (CAI-Asia, 2014) and the 2015 China Victory Day Parade (Victory Parade 2015) (Liang et al., 2017), etc. After implementation of these control measures, it is important to understand how effective these strategies are.

The 2nd World Internet Conference was held in Tongxiang,Jiaxing, Zhejiang during 16-18 December, 2015. To reduce air pollution during the conference, Zhejiang Province and the Regional Air-pollution Joint Control Office of the Yangtze River Delta (YRD) region developed an Action Plan for Air Pollution Control during the Conference (henceforth referred to as the Action Plan), which clarified target goals, time periods for implementing controls, regions in which the controls would be applied, and the control measures to be implemented, as described below. **Targets:** achieve an Air Quality Index (AQI) below 100 in "key areas", an AQI below 150 in "control areas", and to achieve significant improvement of the air quality in the surrounding (or buffer) regions outside the control areas. **Time Periods:** the time periods of interest for implementing various controls include the early stage (3 months before the conference), the advanced stage (2 weeks to 4 days before the conference) and the central stage (3 days before and 2 days after the conference). **Regions:** areas within a 50km radius, within a 100km radius and outside of a 100km radius from the centre of Tongxiang were classified as key areas, control areas and buffer areas, respectively. These areas cover 9 cities including Jiaxing, Huzhou, Hangzhou, Ningbo and Shaoxing in Zhejiang province, Suzhou and Wuxi in Jiangsu province and Xuancheng in Anhui province, as shown in Fig.1.

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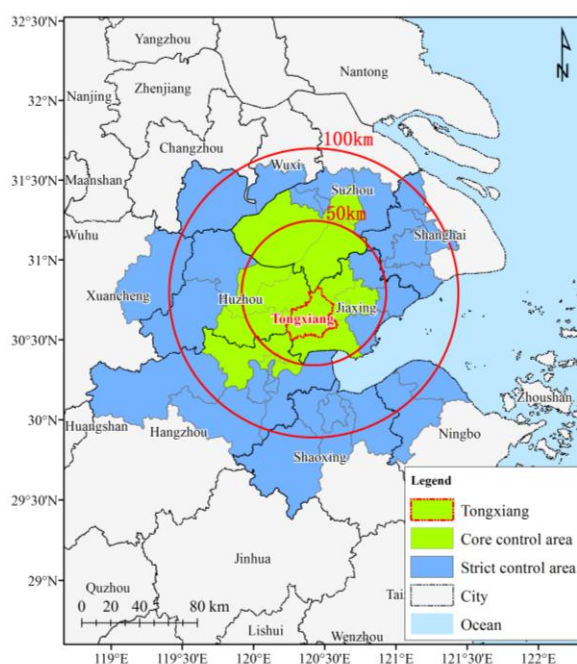


Fig.1 Controlled regions in the Action Plan for Air Quality Control during the World Internet Conference

Many studies have provided descriptive analysis of changing concentrations of air pollutants during mega events; some have reported the emission reductions and related air quality changes (Wang, et al., 2009; Wang, et al., 2010; Liu, et al., 2013; Tang, et al., 2015; Li, et al., 2016; Wang, et al.,2016; Sun, et al., 2016; Wang, et al., 2015; Chen, et al., 2017; Han, et al., 2016; Qi, et al., 2016). However, different air pollution control targets, different control measures, and different locations, may cause significantly different effects among those strategies. In this paper, the reduction in $PM_{2.5}$ achieved through the Action Plan is investigated further to help quantify the level of $PM_{2.5}$ reduction that can be attributed to different aspects of the Action Plan. An integrated emission-measurement-modelling method described in the next section including analysis of multi-pollutant observations, backward trajectory and potential source contribution analyses, estimates of pollutant emission reductions, and photochemical model simulations were adopted to conduct a comprehensive assessment of the impact of control measures on air quality improvement based on three aspects: meteorological conditions, pollutant emission reductions of local sources, and regional contributions.

2 Methodology

In order to strengthen the regional air pollution joint-control mechanism in the YRD region, various measures and their implementation were systematically reviewed, and the qualitative and quantitative relationships among the implementation of measures, changes in emissions of air pollution sources and air quality improvement were studied. Specifically, the impact of measures such as management and control of coal-burning power plants, production restriction and suspension of industrial enterprises, motor vehicle limitation and work site suspension,

dust control were investigated. In addition, the role of meteorology (in particular, transport) was assessed in terms of its influence on the relevance and effectiveness of various measures, and ways of optimising air quality control measures and emergency emission reductions under heavy pollution during major events were evaluated.

To assess the effectiveness of the various controls outlined in the Action Plan, emission reductions associated with those controls were calculated, and photochemical modelling was conducted to determine the change in $PM_{2.5}$ attributed to specific controls. On this basis, an assessment of how to optimise control measures was carried out with respect to both the area in which the emission reduction took place, as well as the start time for implementing the controls (i.e., how far in advance do the controls need to be implemented). Analysis of the numerical modelling results is focused on the effectiveness of the control measures with respect to regional transport of pollutants in the YRD region.

2.1 Measurements

The On-line observational station was set up at the Shanxi supersite of Zhejiang Province (30.82 $^{\circ}$N, 120.87 $^{\circ}$E), which was located at the core area for pollution-control measures. On-line hourly $PM_{2.5}$ mass concentration, carbonaceous aerosols, elements, and ionic species were measured by the Synchronized Hybrid Ambient Real-time Particulate Monitor (SHARP, model 5030, Thermo Fisher Scientific Corporation, USA), the OC/EC carbon aerosol analyzer (Model-4, Sunset Laboratory Corporation, USA), the Xact multi-metals monitor (XactTM 625, PALL Corporation, USA), and the Ambient Ion Monitor-Ion Chromatograph (AIM IC, model URG 9000, URG Corporation, USA), respectively. Meteorological parameters, including wind speed, wind direction, temperature, pressure, and relative humidity, were measured as well.

$PM_{2.5}$ concentration data quality conform to the standards of data quality control published by Ministry of Ecology and Environment of the People's Republic of China.

A semi-continuous Sunset OC/EC analyser was used to measure OC and EC mass loadings at the observation site by adopting NIOSH-5040 protocol based on thermal-optical transmittance (TOT). The ambient air was first sampled into a $PM_{2.5}$ cyclone inlet with a flow rate of 8 L·min^{-1}. The OC and EC were collected on a quartz fiber filter with an effective collection area of 1.13 cm^2. The analyzer was programmed to collect aerosol for 45 min at the start of each hour, followed by the analysis of carbonaceous species during the remainder of the hour. The analysis procedure is described in detail by Huang et al. (2018)

The ionic concentrations of nitrate, sulphate, chloride, sodium, ammonium, potassium, calcium and magnesium (Na^+, K^+, Ca^{2+}, NH_4^+, Mg^{2+}, NO_3^-, SO_4^{2-}, Cl^-) in the fine fraction ($PM_{2.5}$) were measured with a 1-hour time resolution using the AIM IC. The sample analysis unit is composed by an anion and a cation ion

chromatographs (Dionex ICS-1100), which was using guard columns with potassium hydroxide eluent (KOH) for the anion system and methane sulfonic acid (MSA) eluent for the cation system. The limit of the detection reported by the manufacturer is 0.1 ug/m^3 for all species. The operation principle of AIM-IC is described in detail by Markovic et al. (2012)

Hourly ambient mass concentrations of sixteen elements (K, Ca, V, Mn, Fe, As, Se, Cd, Au, Pb, Cr, Ni, Cu, Zn, Ag, Ba) in PM$_{2.5}$ were determined by the Xact multi-metals monitor. In brief, the Xact instrument samples the air through a section of filter tape at a flow rate of 16.7 lpm using a PM$_{2.5}$ sharp cut cyclone. The exposed filter tape spot then advances into an analysis area where the collected PM$_{2.5}$ is analyzed by energy-dispersive X-ray fluorescence (XRF) to determine metal mass concentrations. The sequence of sampling and analysis were performed continuously and simultaneously on an hourly basis.

2.2 Potential Source Contribution Analysis

TrajStat is a HYSPLIT model developed by Chinese Academy of Meteorological Sciences and NOAA Air Resources Laboratory based on geographic information system (GIS). It uses statistical methods to analyze air mass back trajectories to cluster trajectories and compute potential source contribution function (PSCF) with observation data and meteorological data included (Wang et al., 2009).

PSCF analysis is a conditional probability function using air mass trajectories to locate pollution sources. It can be calculated for each 1° longitude by 1° latitude cell by dividing the number of trajectory endpoints that correspond to samples with factor scores or pollutant concentrations greater than specified values by the number of total endpoints in the cell (Zeng et al., 1989). Therefore, pollution source areas are indicated by high PSCF values. Since the deviation of PSCF results could increase with the raise of distance between cell and receptor, therefore a weight factor (W$_{ij}$) was adopted in this study to lower the uncertainty of PSCF results. PSCF and W$_{ij}$ calculations are described in Eq. (1) and Eq. (2), where m$_{ij}$ is the number of trajectory endpoints greater than specified values in cell (i, j), n$_{ij}$ is the number of total endpoints in this cell (Zeng et al., 1989; Polissar et al., 1999).

$$P = \frac{m_{ij}}{n_{ij}} \cdot W(n_{ij}) \tag{1}$$

$$W(n_{ij}) = \begin{cases} 1.00, & 80 < n_{ij} \\ 0.70, & 20 < n_{ij} \leq 80 \\ 0.42, & 10 < n_{ij} \leq 20 \\ 0.05, & n_{ij} \leq 10 \end{cases} \tag{2}$$

In this study, the TrajStat modelling system was used to analyze potential source contribution areas of PM$_{2.5}$ in Jiaxing during different pollution episodes with the combination of Global Data Assimilation System (GDAS) meteorological data provided by the NCEP (National Center for Environmental Prediction). Polluted air mass

143   trajectories corresponded to those trajectories with PM$_{2.5}$ hourly concentration higher than 75 μg/m^3.

144   **2.3 Model setup**

145   **2.3.1 Model selection and parameter settings**

146   In this study, the WRF-CMAQ/CAMx air quality numerical modelling system was used to evaluate the

147   improvement in air quality resulting from the control measures outlined in the Action Plan. It takes into account of

148   modeling variations from different air quality models. For the mesoscale meteorological field, we adopted the WRF

149   model Version 3.4 (https://www.mmm.ucar.edu/wrf-model-general), the CAMx model Version 6.1

150   (http://www.camx.com/) and the CMAQ model Version 5.0 (Nolte et al., 2015; http://www.cmascenter.org/cmaq/).

151   The chemical mechanism utilized in CMAQ was the CB05 gas phase chemical mechanism (Yarwood, et al., 2005)

152   and AERO5 aerosol mechanism, which includes the inorganic aerosol thermodynamic model ISORROPIA (Nenes,

153   et al., 1998) and updated SOA yield parameterizations. The gaseous and aerosol modules used in CAMx are the

154   CB05 chemical mechanism and CF module, respectively. The aqueous-phase chemistry for both models is based

155   on the updated mechanism of the Regional Acid Deposition Model (RADM) (Chang et al., 1987). Particulate Source

156   Apportionment Technology (PSAT) coupled in the CAMx is applied to quantify the regional contributions to PM$_{2.5}$

157   as well. The WRF meteorological modeling domain consists of three nested Lambert projection grids of 36km-

158   12km-4km, with 3 grids larger than the CMAQ/CAMx modeling domain at each boundary. WRF was run

159   simultaneously for the three nested domains with two-way feedback between the parent and the nest grids. Both the

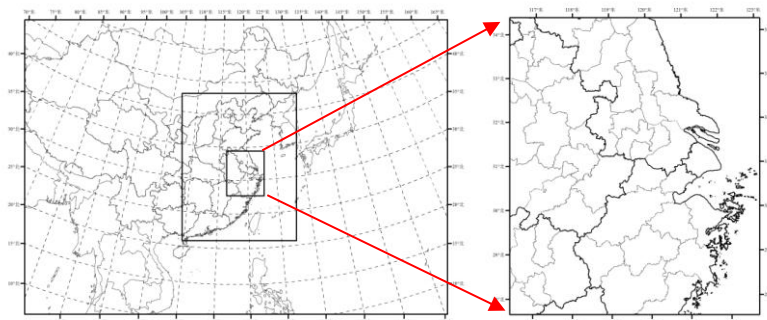
160   three domains utilized 27 vertical sigma layers with the top layer at 100hpa, and the major physics options for each

161   domain listed in Table 1. For the CMAQ/CAMx modelling domain shown in Figure 2, we adopted a 36-12-4km

162   nested domain structure with 14 vertical layers, which were derived from the WRF 27 layers. The two outer domains

163   cover much of eastern Asia and eastern China, respectively, while the innermost domain covers the YRD region.

164   The simulation period was from 1-18 December, 2015, during which 1-7 December was utilized for model spin-up

165   and 8-18 December was the key period for analysis of the modelling results with control measures.

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167   Fig 2. Modeling domain

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Table 1 Parameterization scheme of the physical processes in the WRF model

Physical Processes	Parameterization Scheme	Reference
Microphysical Process	Purdue Lin Scheme	(Lin, 1983)
Cumulus Convective Scheme	Grell-3 Scheme	(Grell and Dévényi, 2002)
Road Process Scheme	Noah Scheme	(Ek, 2003)
Boundary Layer Scheme	Yonsei University (YSU) Scheme	(Hong, 2006)
Long-wave Radiation	RRTM Long-wave Radiation Scheme	(Mlawer et al., 1997)
Short-wave Radiation Scheme	Goddard Short-wave Radiation Scheme	(Chou and Suarez, 1999)

170   Initial and boundary conditions (IC/BCs) for the WRF modeling were based on 1-degree by 1-degree grids

171   FNL Operational Global Analysis data that are archived at the Global Data Assimilation System (GDAS). Boundary

172   conditions to WRF were updated at 6-hour intervals for D01.

173   Anthropogenic source emission inventory in YRD is based on a most recent inventory developed by our group

174   (Huang et al., 2011; Li et al., 2011; Liu et al., 2018). The emission inventory for areas outside YRD in China is

175   derived from the MEIC model (Multi-resolution Emission Inventory of China, latest data for

176   2012(http://www.meicmodel.org) and anthropogenic emissions over other Asian region are from the MIX emission

177   inventory for 2010 (Li et al., 2017). Biogenic emissions are calculated by the MEGAN v2.1 (Guenther et al., 2012).

178   We further developed a controlled emission inventory to account for the control measures based on the emission

179   reduction requirements described in the control measures paln and the control measures for the emergency air

180   pollution warning. These estimates are basically according to the control measures and reduction requirements for

181   specific source sectors and cities described the control plan. The Sparse Matrix Operator Kernel Emissions (SMOKE,

182   https://www.cmascenter.org/smoke) model is applied to process these emissions for modeling inputs that is more

183   detailed emission processes and not usually used in China.

2.3.2 Model performance

184

185   Prior to evaluating the effectiveness of the control measures and reactions, the performance of the modelling

186   system was evaluated to ensure it was able to reasonably reproduce the observed meteorological conditions and

187   PM$_{2.5}$ levels. Statistical indexes used for model evaluation include Normalised Mean Bias (NMB), Normalised

188   Mean Error (NME) and Index of Agreement (IOA). The equations to calculate these statistical indexes are as follows:

$$NMB = \frac{\sum(P_j - O_j)}{\sum O_j} \times 100\% \tag{3}$$

$$NME = \frac{\sum |P_j - O_j|}{\sum O_j} \times 100\% \tag{4}$$

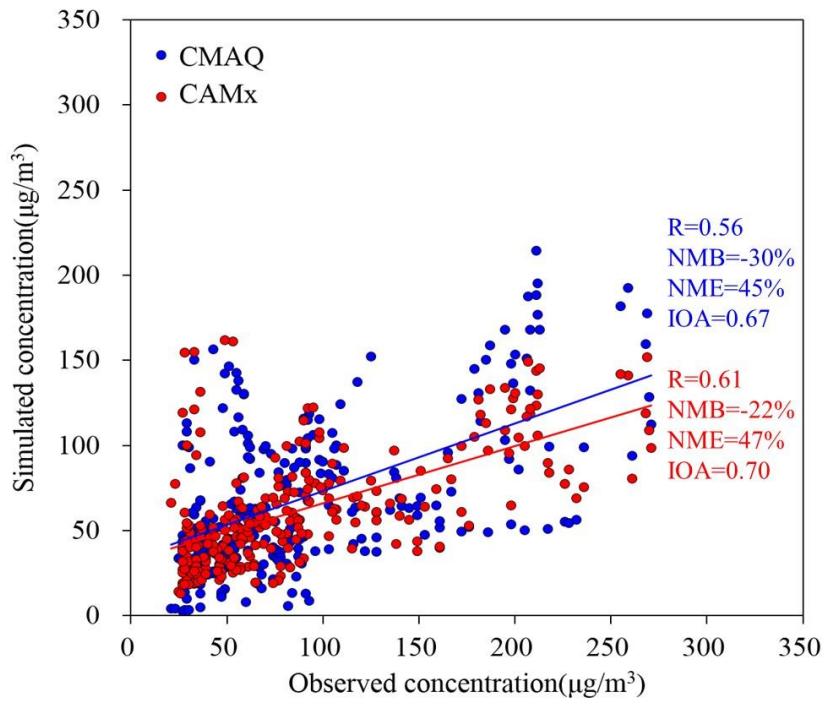
$$IOA = 1 - \frac{\sum(P_j - O_j)^2}{\sum(|P_j - \bar{O}| + |O_j - \bar{O}|)^2} \tag{5}$$

189   where P$_j$ and O$_j$ are predicted and observed hourly concentrations, respectively. $\bar{O}$ is the average value of

190   observations. IOA ranges from 0 to 1, with 1 indicating perfect agreement between model and observation.

Observational data from the Shanxi supersite in Jiaxing City were compared with model results for model evaluation verification. Table 2 shows the summary statistics for the main meteorological parameters simulated with the WRF model and hourly $PM_{2.5}$ concentrations simulated by CMAQ. Among the meteorological parameters, wind speed is slightly over predicted with the NMB value of 28%, while temperature, relative humidity and pressure all have IOA values greater than 0.9. Figure 3 compares the simulated and observed $PM_{2.5}$ concentrations at the Shanxi supersite. In general, model predicted data are lower than the observed data with the NMB value of -22% to -30%, the NME value of 45% to 47% and the IOA value of 0.67 to 0.70 (Table 2). These underestimations may be due to three reasons: Firstly, winter underestimation of $PM_{2.5}$ (especially SOA) is a common issue with CMAQ or CAMx simulations over China (Hu et al., 2017; Li et al., 2016), which can be explained by a lack of model calculated oxidants or missing reactions (Kasibhatla et al., 1997) of SOA formation pathways (Appel et al., 2008; Foley et al., 2010). Secondly, uncertainty still exists in the regional emission inventory, including the basic emissions inventory and the control scenarios. Thirdly, the wind speed is slightly overestimated over the region, with NMB and NME of 28% and 33%, causing fast dispersion of air pollutants. Overall, these statistics for both the meteorological parameters and simulated $PM_{2.5}$ are generally consistent with the results in other published modelling studies(Zheng et al., 2015;Wang et al., 2014;Zhang et al., 2011;Fu et al., 2016;Li et al., 2015b;Li et al., 2015a), which suggests that the simulation performance is acceptable.

Table 2 Statistics of simulation verification for meteorological parameters and hourly $PM_{2.5}$ concentration

Statistical indexes	Wind speed	Temperature	Relative humidity	Air pressure	CAMx-$PM_{2.5}$	CMAQ-$PM_{2.5}$
NMB	28%	3%	-9%	0%	-30%	-22%
NME	33%	14%	12%	0%	45%	47%
IOA	0.81	0.97	0.93	1.00	0.67	0.70

208
209 Fig. 3 Scatter plot of the simulated and observed PM$_{2.5}$ at the Shanxi supersite

210 **2.3.3 Method for quantifying the effectiveness of a control**

211   Quantifying the PM$_{2.5}$ reduction in response to emission reductions was done using the so called Brute Force

212 Method (BFM) (Burr and Zhang, 2011), where a baseline scenario was simulated using unadjusted emissions (i.e.,

213 those emissions that would have occurred in absence of the Action Plan) and a campaign scenario was modelled

214 based on the emission controls outlined in the Action Plan. In both cases, the same meteorology and chemical

215 boundary conditions were utilized to drive the photochemical model simulations. Through a comparative analysis

216 of the scenarios, a relative improvement factor (RF) for a given atmospheric pollutant, resulting from emission

217 controls, can be calculated and combined with ground based observations to assess the improvement in air quality

218 associated with those emission controls.

219 $$RF = (C_b\text{-}C_s)/C_b \qquad\qquad (6)$$

220 $$C_d=C_o·RF \qquad\qquad (7)$$

221   where C_b is the simulated pollutant concentration in the baseline scenario (μg/m^3), C_s is the pollutant

222 concentration in the campaign scenario (μg/m^3), C_o denotes the actual observed concentration at the site (μg/m^3)

223 and C_d is the concentration improvement caused by the control measures (μg/m^3). Utilizing models in a relative

224 sense to assess the efficiency of emission controls on air quality is common practice in regulatory modelling, with

225 the assumption that there may be biases in the absolute concentrations simulated by a modelling system, but that

226 the relative response of that system will reflect the response observed in the atmosphere (US EPA, 2014).

227 **3 Results and discussion**

3.1 Photochemical transformation changes of air pollutants during the campaign

Ground observational data show that from December 1 to December 23, Jiaxing City experienced four distinct physical and chemical processes that contributed to the observed pollution levels at different periods. For each of these processes, this study utilized the integrated emission-measurement-modeling method to analyze the evolution of air quality from several aspects, including the backward air flow trajectory, potential contribution source areas, meteorological conditions and the variation of PM$_{2.5}$ concentration.

3.1.1 Pollution process before the campaign with local emission accumulation as the main contributor

The first time period of interest was from December 6 to December 8. Analysis about the potential source contribution areas resulting from PSCF modelling suggests that the polluted air mass primarily originated from the northwest and northerly airstreams, passing Shandong, the eastern coastal areas of Jiangsu and Shanghai and into northern Zhejiang, as is shown in Fig. 4. Analysis of the large-scale weather patterns showed that the polluted air mass occurred in Beijing, Tianjin, Shandong peninsula and northern Jiangsu as a result of cold air with polluted air mass transported into the region on the morning of December 5. In the southern part of Shandong province, the PM$_{2.5}$ concentration peak appeared on the morning of December 6, while the PM$_{2.5}$ concentration peak appeared around midnight on December 7 at the coastal area of Jiangsu. On December 6, the development of warm and humid air flow, resulted in increasing ground humidity, which contributed to the growth of secondary fine particles and the gradual accumulation of polluted air mass in northern Zhejiang and the surrounding areas of Shanghai. On December 7, affected by the surface high-pressure system, the spread of plume was slow, and the spatial extent of the plumes in northern Zhejiang expanded. Therefore, during this time period, the pollution was primarily affected by regional transport and worsened by stagnant local conditions in Jiaxing.

(a)

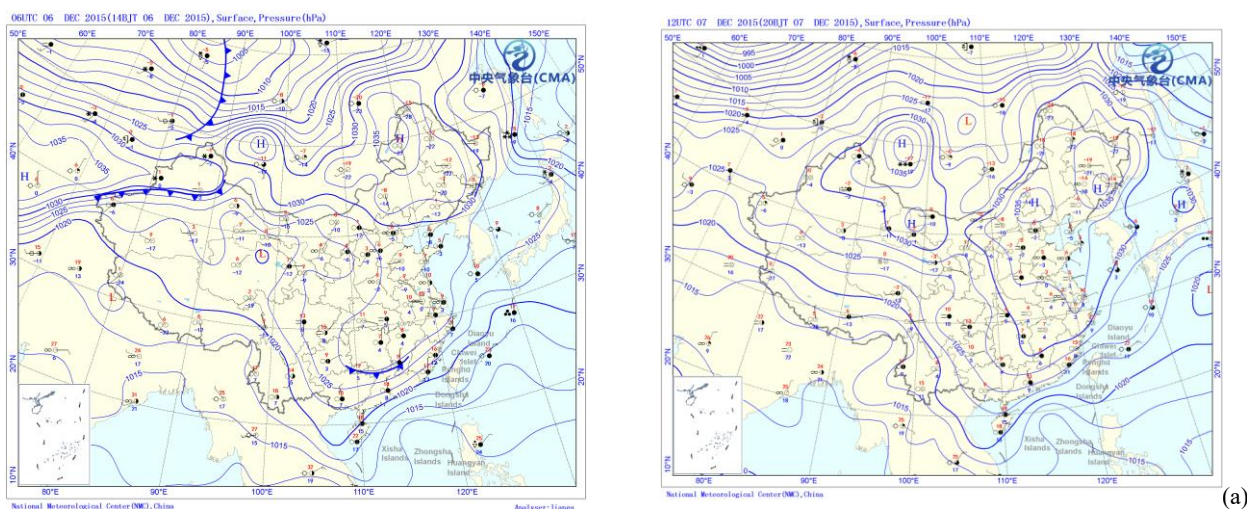
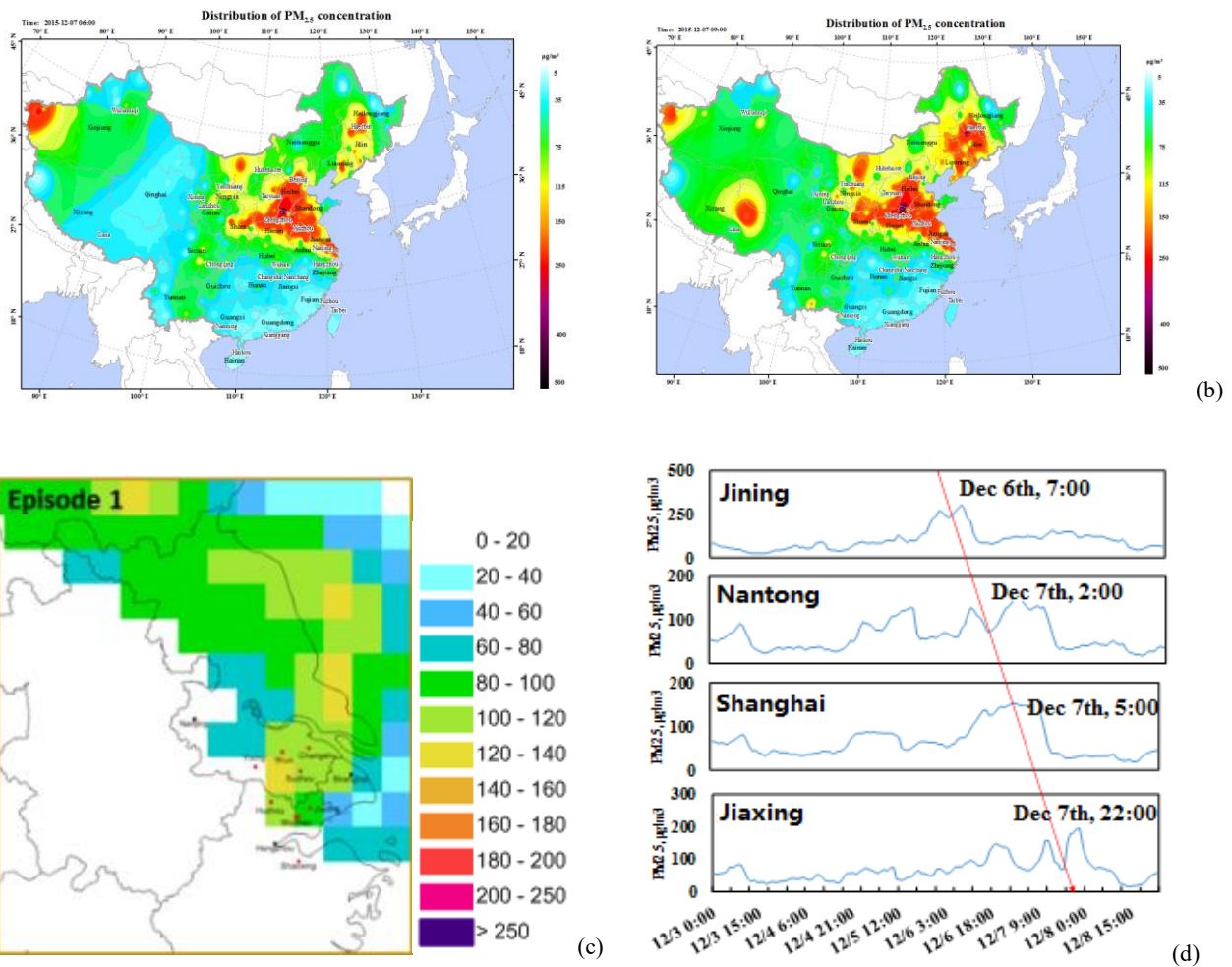



Fig. 4 Analysis of (a) the large-scale weather patterns, (b) distribution of PM2.5 concentrations, (c) potential source regions, (d) Observed PM2.5 time series for selected sites during December 6 to December 8, 2015

3.1.2 Pollution process during the campaign with the southward motion of the weak cold air

The second time period of interest was from December 10 to December 11. Analysis about potential source contribution areas suggests that the polluted air mass mainly came from northern regions, passing from south-eastern Shandong peninsula and central-eastern Jiangsu to northern Zhejiang. From the large-scale weather pattern, the diffusion of weak cold air on December 10 gradually transported the polluted air mass in the upper reaches of the region to the YRD region. The pollution peaked in areas such as Lianyungang in northern Jiangsu on the evening of December 10. On December 11, the PM2.5 concentration peak appeared in central and southern Jiangsu as a result of northern weak air flow. The plume was further transported into Zhejiang province with the expansion in influenced areas as is shown in Figure 5. Therefore, the pollution process was mainly affected by the transport of polluted air mass caused by the southward motion of cold air.

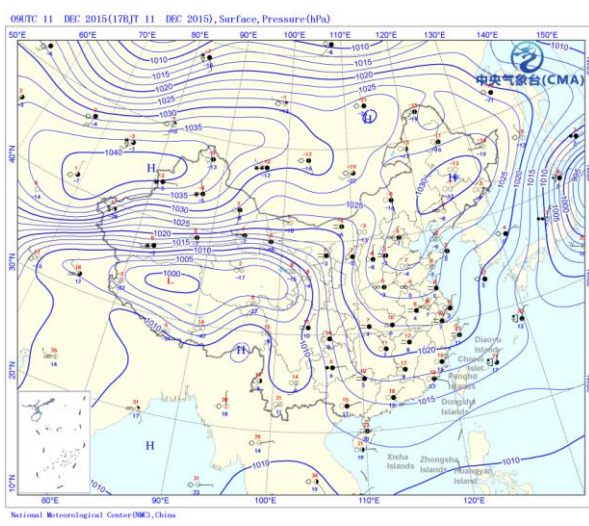
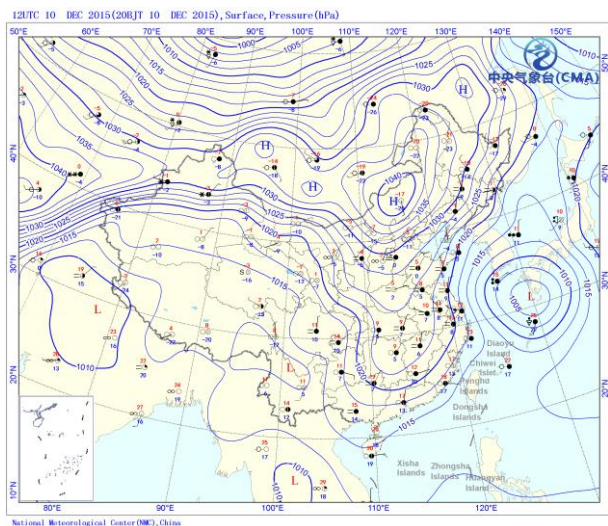
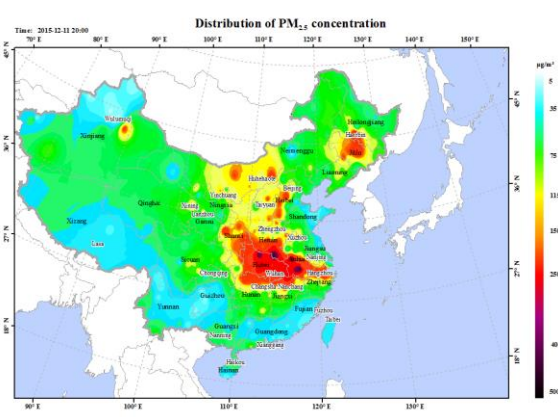
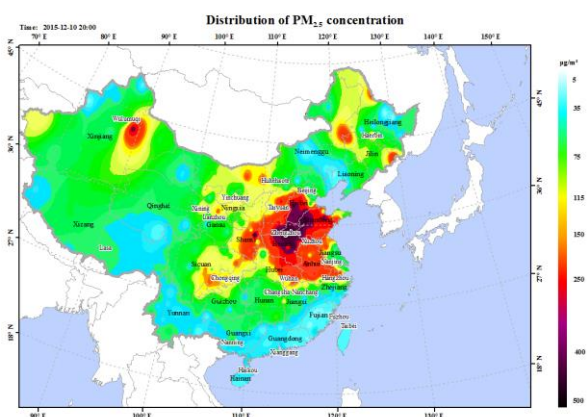
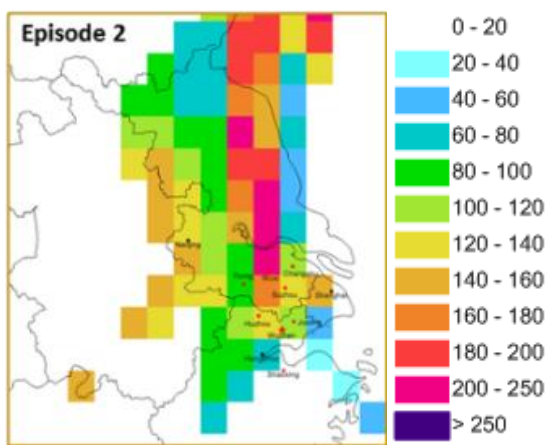
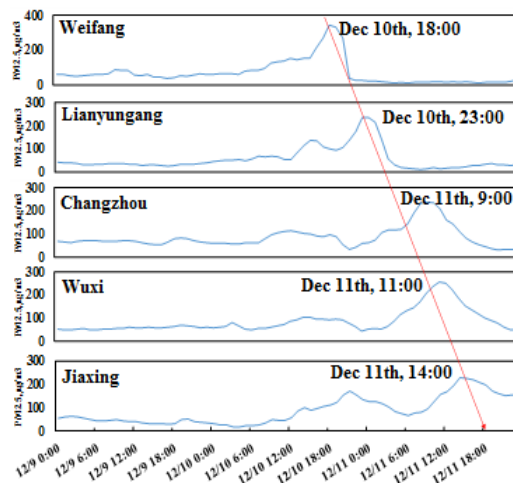



Fig. 5 Analysis of (a) the large-scale weather patterns, (b) distribution of PM$_{2.5}$ concentrations, (c) potential regional sources, (d) Observed PM$_{2.5}$ time series for select sites during December 10 to December 11, 2015

3.1.2 Heavy pollution process during the campaign with the transit and transport of strong cold air

The third period of interest was from December 13 to the early hours of December 16. Analysis of the potential source contribution areas suggests that the polluted air mass mainly came from the northwest direction, passing through south-eastern Shanxi, western Shandong, eastern Anhui and western Jiangsu to Zhejiang province. On December 14, affected by the cold air transport in the north, northern plumes hit Hebei, Henan and Anhui provinces,

with the highest degree of pollution on the 14th. On December 15, the further spread of cold air caused the transport of plumes into Jiangsu and Zhejiang. The northern part of Zhejiang province was in the centre of pollution on the 15th, which worsened the pollution and expanded the scope of pollution, as is shown in Figure 6. On December 16, under the control of the high-pressure system in northern Zhejiang, the air mass gradually moved eastward and the air quality improved in the morning. Therefore, for this time period, large-scale transport was the main factor leading to the increase in pollutant levels.

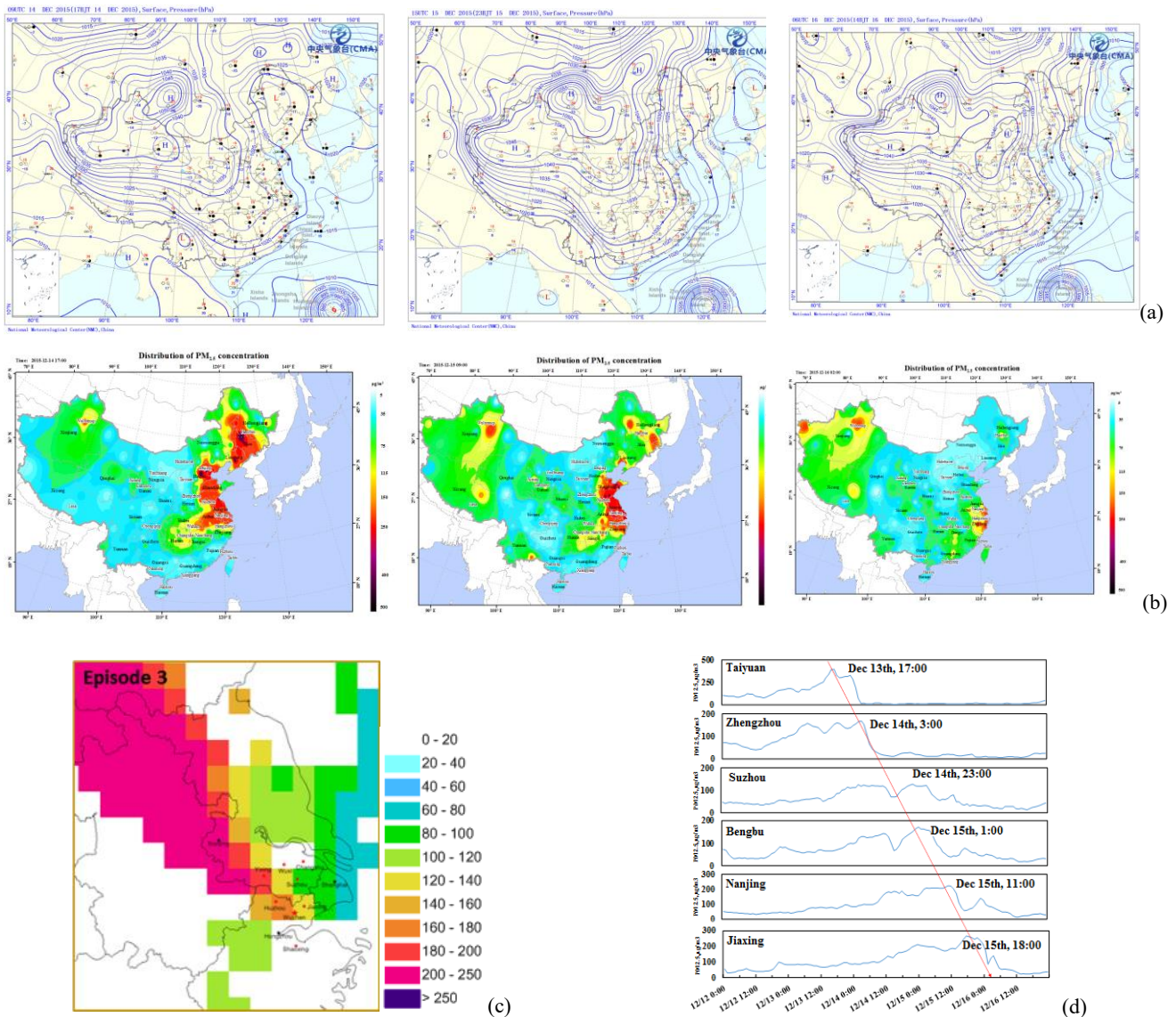



Fig. 6 Analysis of (a) the large-scale weather patterns, (b) distribution of PM$_{2.5}$ concentrations, (c) potential regional sources, (d) Observed PM$_{2.5}$ time series for select sites during December 14 to December 16, 2015

3.1.3 Pollution removal process caused by clean cold air during the conference

During the conference from December 16 to December 18, weather was affected by the large-scale southward transport of cold dry air in northern Zhejiang, resulting in lower temperature and relative humidity, as well as a significant improvement in the air quality. On the 17th and the 18th, under the control of a high pressure system in

northern Zhejiang, the sea level pressure increased, the humidity was lower and the wind speed was reduced. Because of the emission reduction effect of the control measures, the pollutant accumulation rate was likely slowed down and the air quality in northern Zhejiang was good overall. From the analysis of potential sources, PM$_{2.5}$ concentrations in Shandong, Jiangsu and Shanghai were significantly reduced. The PM$_{2.5}$ concentration during the conference was mainly controlled by local emissions, as is shown in Figure 7.

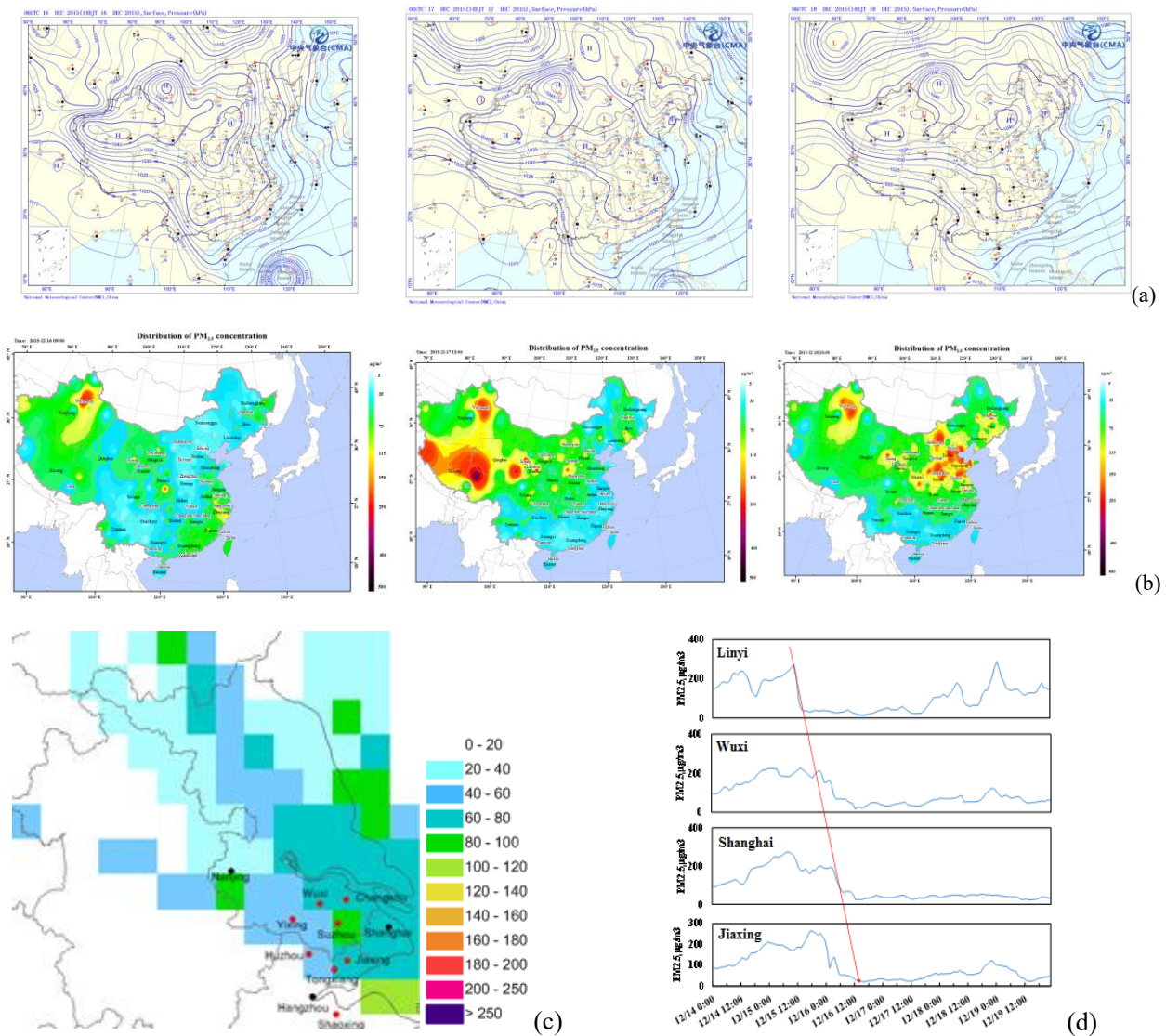



Fig. 7 Analysis of (a) the large-scale weather patterns, (b) distribution of PM$_{2.5}$ concentrations, (c) potential regional sources, (d) Observed PM$_{2.5}$ time series for select sites during December 16 to December 18, 2015

3.1.4 Pollution process after the campaign with local emission accumulation as the main contributor

The fourth period of interest was from December 20 to December 23. Analysis of the potential source contribution areas suggests that the polluted air mass mainly came from the southwest direction, passing through southern Hubei, southern Anhui and south-western Jiangsu to northern Zhejiang. On December 20, controlled by a stagnant air mass, Zhejiang province has a relatively low near-surface wind speed and little dispersion, resulting in the accumulation of local pollutants. On December 21, northern Zhejiang was located in the centre of a high pressure

system with conditions conducive to little mixing, and therefore polluted air mass occurred in some areas in northern Zhejiang. On December 22, affected by the warm and humid southwest air flow, Zhejiang had experienced some precipitation but the pollution in northern Zhejiang was not improved due to deep polluted air masses. In Hubei and Anhui located in the southwest of Jiaxing City, high pollution levels appeared from the evening of December 22 to the early hours of December 23, as is shown in Figure 8. On December 23, the further expansion of polluted air masses resulted in serious pollution in Jiangsu and northern Zhejiang. In general, under these heavily polluted conditions, the local accumulation of pollutants was mainly caused by stagnant conditions with little dispersion and transport within southwest air stream.

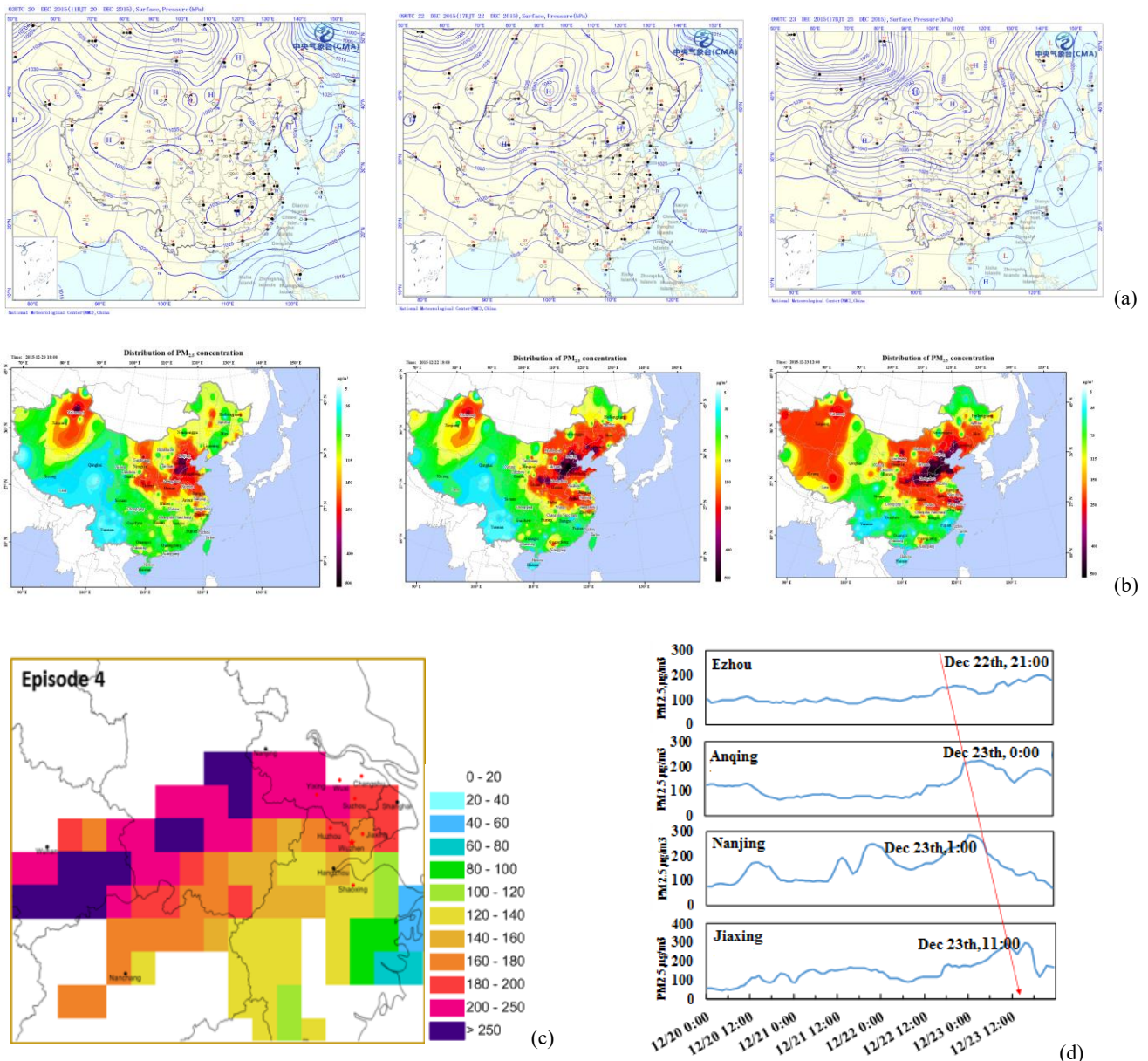



Fig. 8 Analysis of (a) the large-scale weather patterns, (b) distribution of PM2.5 concentrations, (c) potential regional sources, (d) Observed PM2.5 time series for select sites during December 20 to December 23

3.2 Air quality changes under the same meteorological conditions before and after the campaign

3.2.1 Air quality changes under static meteorological conditions before and during the campaign

During the air pollution control campaign for the conference, air quality in Jiaxing City fluctuated greatly due to the frequent southward motion of cold air from the north. Under static weather conditions, sources of atmospheric pollution mainly came from the accumulation of polluted air masses from local sources and sources in neighbouring areas. Therefore, in order to eliminate the influence of the transport process of the air mass, this study compared the air quality status before, during and after the campaign in Jiaxing City under stagnant weather conditions (wind speed less than 1m/s) and assessed the impact of control measures on ambient air quality in Jiaxing based on air quality observation data.

Figure 9 shows the concentration levels of criteria pollutants including SO_2, NO, CO, NO_2 and $PM_{2.5}$ in Jiaxing City before (December 1-7), during (December 8-19) and after the regulation (December 19-31) under stagnant weather conditions. It can be seen that pollutant concentrations during the campaign were less than those before the campaign, in which SO_2 had the most significant decline of 40.1%, NOx, CO, $PM_{2.5}$ and PM_{10} declined 8.0%, 2.6%, 12.5% and 16.3%, respectively, indicating that control measures have significantly improved the air quality in Jiaxing City, especially with respect to SO_2 and PM_{10}.

After the campaign, all the pollutant concentrations rebounded sharply. SO_2, NO, NO_2, CO, $PM_{2.5}$, PM_{10} increased 8.3%, 15.4%, 10.3%, 31.8%, 32.2% and 28.6%, respectively. Concentrations of some pollutants were even higher than those before the campaign, which suggests that the emission intensity of the sources had significantly increased after the campaign.

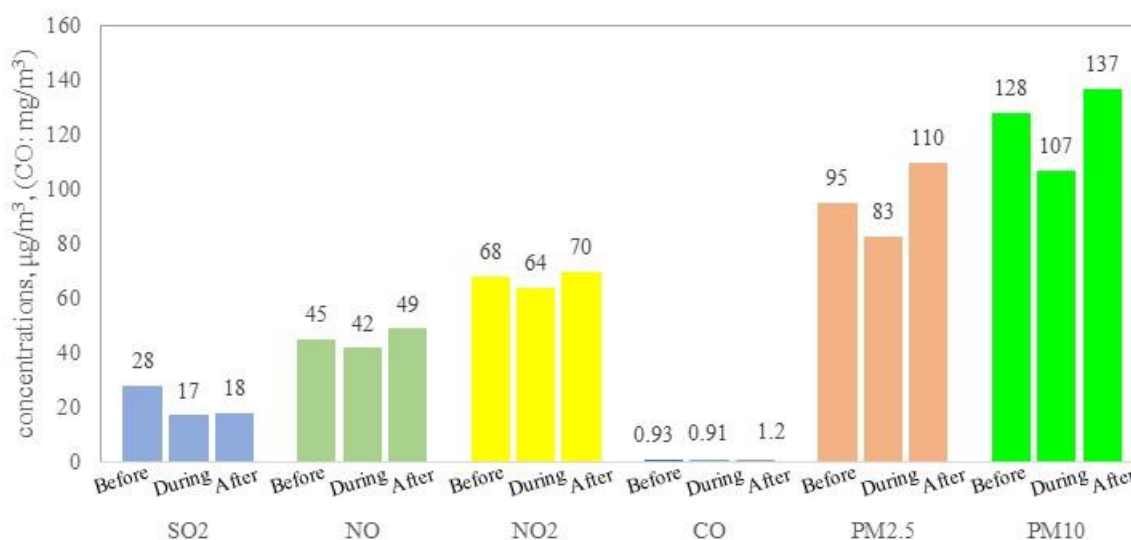
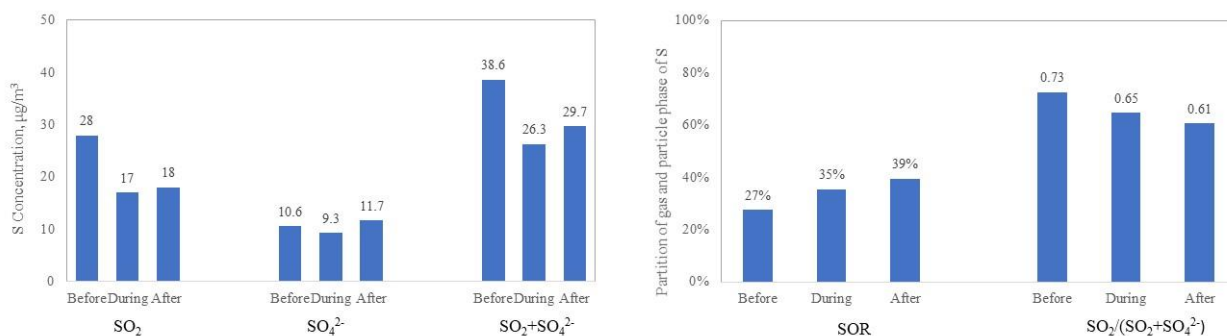
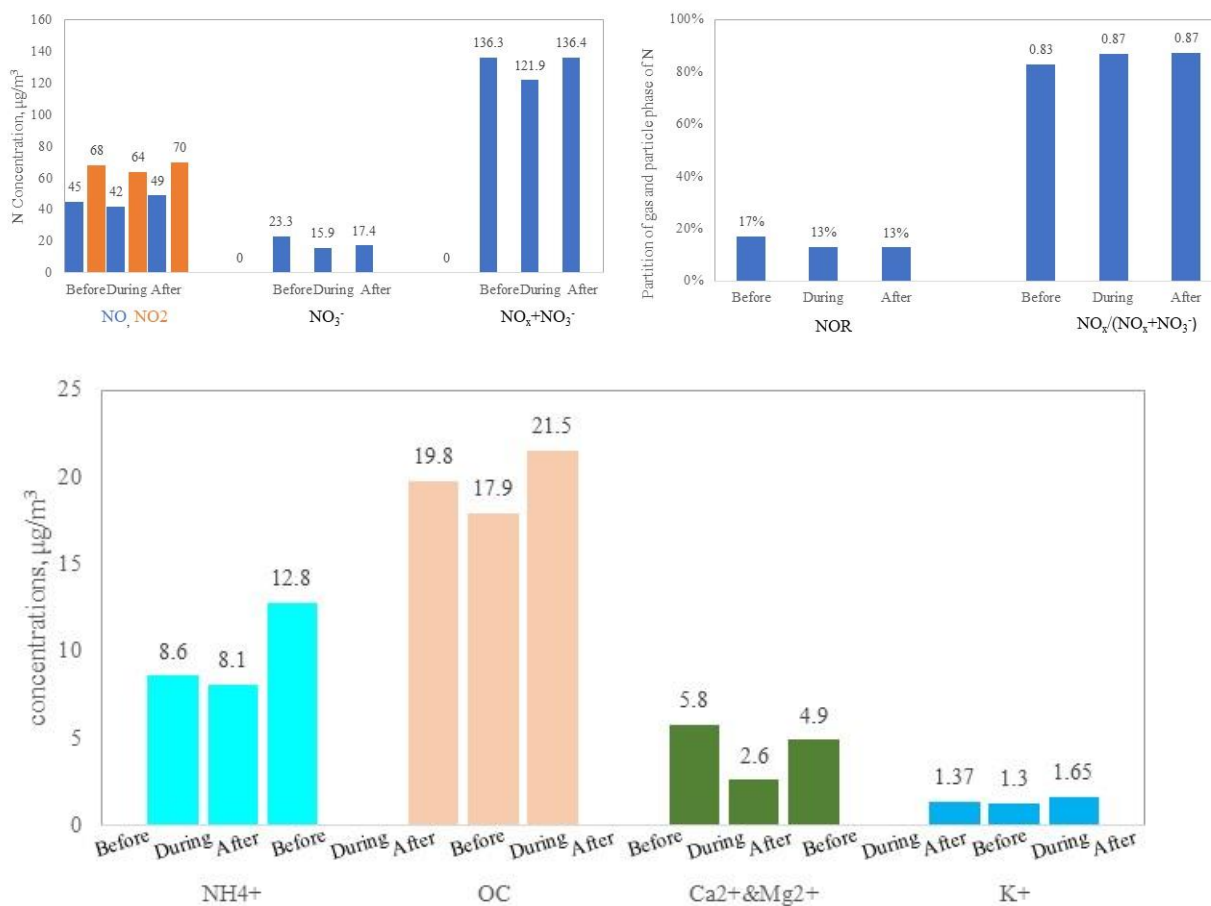



Fig. 9 Comparison between air pollutant concentrations at Shanxi station before, during, and after the campaign under stagnant meteorological conditions

There are also some differences in concentrations of major chemical components of $PM_{2.5}$ in Jiaxing City before (December 1-7), during (December 8-19) and after the campaign (December 19-31) under static weather

conditions, as shown in Figure 9. The concentrations of major chemical components of $PM_{2.5}$ during the campaign were less than those before the campaign, which is consistent with the conclusion about changes in criteria pollutant concentrations. On average, SO_4^{2-}, NH_4^+, NO_3^-, OC, mineral soluble irons (Ca^{2+} and Mg^{2+}) and K^+ declined 11.8%, 5.1%, 32.1%, 9.8%, 56.8% and 5.1%, respectively. Comparisons between the distribution of $PM_{2.5}$ chemical components before and during the campaign under static conditions suggest that Ca^{2+} and Mg^{2+} decreased most significantly during the control period, which indicates that the suspension of construction operations which result in dust emissions and the rising frequency of rinsing and cleaning paved roads, significantly reduced dust emissions. During the campaign, NO_3^- significantly decreased, indicating that vehicle control measures successfully reduced NO_x emissions and subsequently the formation of inorganic aerosols. The significant decrease in SO_4^{2-} also shows that restricting or suspending the operation of coal-burning power plants and industries in local and neighbouring cities played a very positive role.

Both the chemical compositions of $PM_{2.5}$ and chemical processes associated with $PM_{2.5}$ production change if we compare observed data during and after the regulation. As is shown from figure 10, the SO_2 concentrations after control is a little bit higher than during control (+5.9%). However, the SO_4^{2-} after control is much higher than during control (25.8%). This is probably due to two reasons: firstly, SO_2 emissions and primary sulfate emissions increased after the control measures were terminated; secondly, previous studies have reported that increased NOx emissions could accelerate the formation of secondary sulfate (Cheng et al., 2016). This can be clearly seen from the SOR. A different trend is observed for NO_2 and NO_3^-, with the NO_2 concentrations after control being much higher than during control (+9.4%), while the increase of NO_3^- (+9.45%) is about the same. Sulfate originates from both primary emissions and secondary formation, but nitrate is mostly secondary. The NOR during and after regulation is about the same and most of the N is in the gas phase as indicated by $NOx/(NOx+NO_3^-)$ (0.87). Therefore, the increase of NO_3^- is smaller than SO_4^{2-}. The $PM_{2.5}$ concentration after control sharply rebounded by 31.8%, indicating that both primary emissions and secondary formation are activated.

352

353
354 Fig. 10 Comparison between PM$_{2.5}$ chemical components at Shanxi station before and after the campaign under static meteorological
355 conditions

356 **3.2.2 Air quality changes under the same air mass trajectory before and during the campaign**

357   In order to distinguish the impact of meteorological conditions on air quality in Jiaxing City and better analyze

358 the effects of control measures on air quality during the conference, this study has combined meteorological

359 conditions with backward air flow trajectory analysis and carried out a comparative study by selecting a relatively

360 similar pollution period before and during the campaign. The first period occurred before the campaign from 12:00

361 December 2 to 20:00 December 4, while the second period occurred during the campaign from 9:00 December 16

362 to 5:00 December 18. Both of these periods were relatively unaffected by long-range transport of plumes into the

363 study area, and have similar backward airflow trajectories and meteorological conditions. Table 3 and Figure 11

364 compare average mass concentrations of pollutants (SO$_2$, NO$_x$, PM$_{2.5}$ and PM$_{10}$) during these two periods. As can

365 be seen from the figure, SO$_2$, PM$_{2.5}$ and PM$_{10}$ decreased during the campaign by roughly 46%, 13% and 27%,

366 respectively, while NO$_x$ exhibited only a small decrease. This shows that without the impact of long-range transport,

367 emission reduction measures carried out by local and surrounding cities play a significant role in defining the air

368 quality in Jiaxing.

369
370   Table 3 Concentrations of major pollutants under similar meteorological conditions before and during the campaign

Period	Time	Wind speed m/s	Wind direction °	Relative humidity %	Temperature °C	Pressure hPa	Visibility km	SO₂ µg/m³	NO₂ µg/m³	PM₁₀ µg/m³	PM₂.₅ µg/m³
Before the campaign	12.2 12:00-12.4 20:00	3.1	268.0	59.2	8.2	102.6	22.8	39.1	44.4	89.5	49.4
During the campaign	12.16 9:00-12.18 5:00	3.4	247.5	53.0	2.6	103.2	32.1	22.4	39.3	65.3	42.8

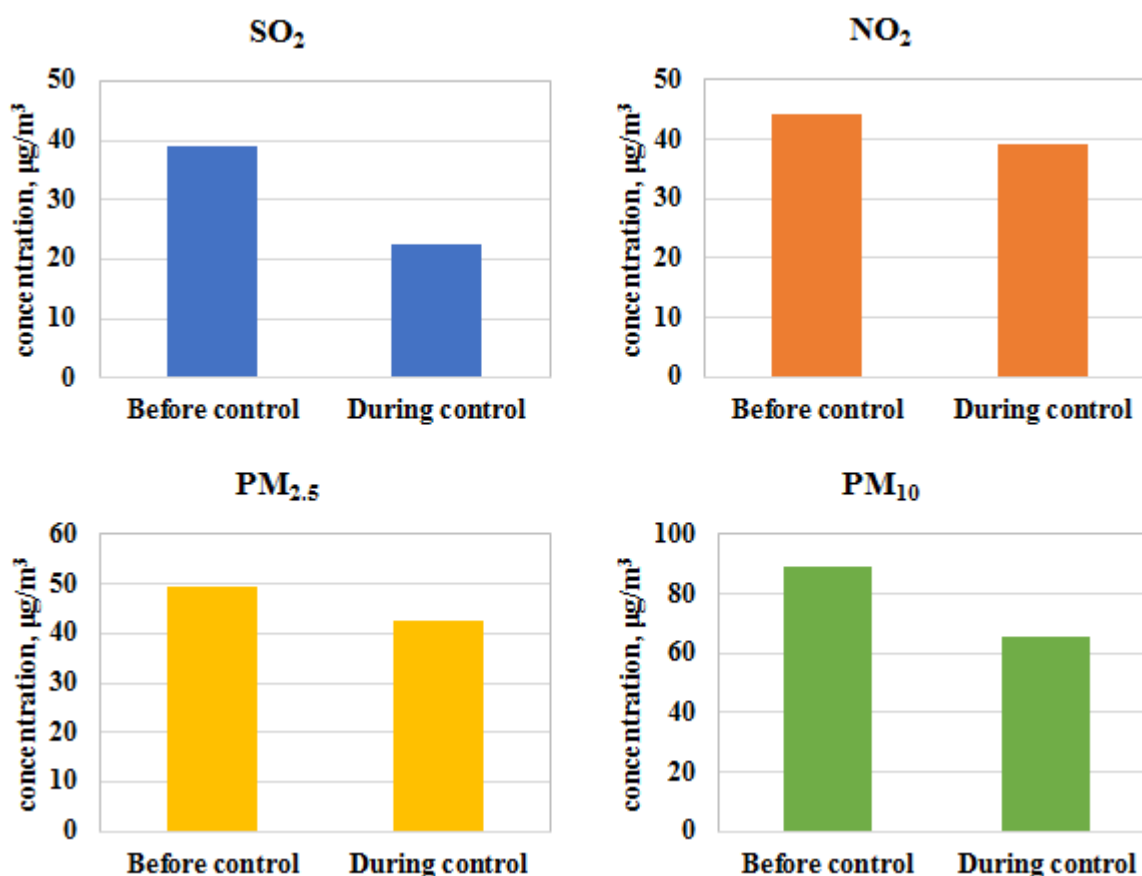



Fig. 11 Comparison between concentrations of major air pollutants in Jiaxing before and after the campaign under same meteorological conditions

There were two regional pollution episodes that occurred during the campaign. The first was on December 10-12 caused by the southward motion of northern weak cold air. Polluted air masses from south-eastern Shandong peninsula passed through central eastern Jiangsu and into northern Zhejiang, affecting the air quality in Jiaxing. During this period, the average daily PM₂.₅ concentration in Jiaxing was 145.7 µg/m³, higher than the regional average, and its major chemical components were nitrate (31%), sulphate (18%), ammonium (13%) and organic carbon (13%), with obvious regional secondary pollution characteristics.

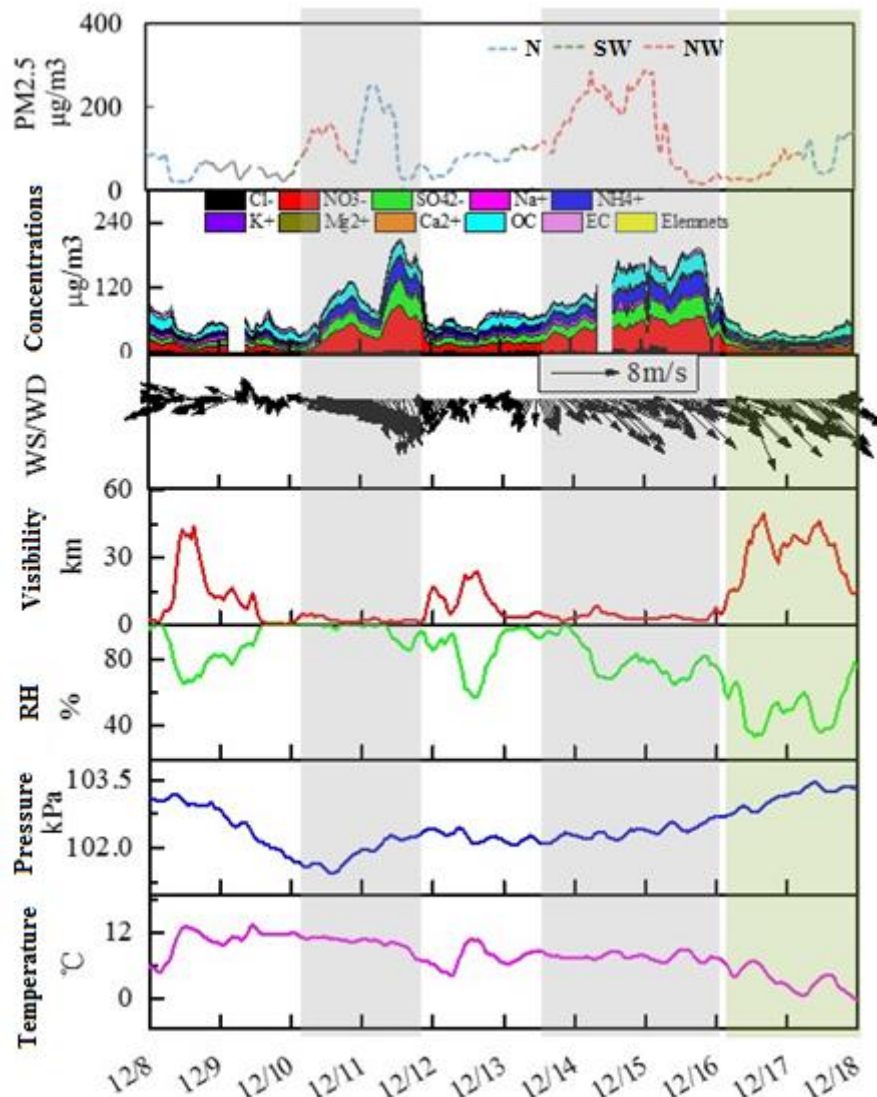



Fig.12 Changes in air quality and meteorological parameters in Jiaxing City during the campaign

The second episode occurred from December 14-15, and was caused by the transit of northwestly strong cold air. Polluted air masses came from the northwest direction, moved rapidly to the southeast, passed through Shanxi, Hebei, west Shandong, east Anhui and west Jiangsu and ultimately into Zhejiang province. The air masses left China through south-eastern Zhejiang on the early morning of the 16th. The YRD region was strongly affected by the transport of the polluted air mass, with heavy polluted air masses appearing and lasting for about one day over the YRD region from north to south. PM$_{2.5}$ peaked in Jiaxing on the 15th with a daily average of 201.6 µg/m^3. The main chemical components of PM$_{2.5}$ during the episode were nitrate (25%), sulphate (14%), ammonium (12%) and organic carbon (13%), which is consistent with an aged air mass as well as regional secondary pollution characteristics.

The regional linkage was initiated from December 16 to December 18, combined with favourable mixing conditions brought by the cold front. The overall air quality in the YRD region during this time period was good, with an average daily PM$_{2.5}$ concentration in Jiaxing of 45 µg/m^3. The major chemical components during this

395 cleaner period were organic carbon (26%), nitrate (16%), ammonium (12%), sulphate (9%) and other components

396 (37%), with some newly formed particles and no obvious regional transport, suggesting that air pollutants were

397 mainly derived from local emissions.

3.3 Emissions reduction estimation during the campaign

399   The air quality assurance campaign for the 2nd World Internet Conference was from December 8 to December

400 18. In order to ensure the air quality during the conference, three provinces and Shanghai municipality in the YRD

401 region carried out joint control measures. Based on the implementation of control measures in all areas during the

402 conference and whether each area had effectively implemented control measures during December 8-18, regional

403 emission reductions have been assessed. It is estimated that emission reductions of SO_2, NO_x, $PM_{2.5}$ and VOCs

404 caused by production restriction in regional industrial enterprises are 2867.8 tons, 3064.7 tons, 2165.5 tons and

405 5055.4 tons, respectively. Emission reductions of various pollutants caused by the restrictions on motor vehicle

406 traffic are estimated to be 4.7 tons of SO_2, 326.9 tons of NOx, 36.1 tons of $PM_{2.5}$ and 452.5 tons of VOCs. Emission

407 reduction of $PM_{2.5}$ caused by dust control was estimated to be 266.0 tons. Therefore, it can be seen that emission

408 reductions mainly come from industrial sources, while motor vehicle restrictions contributed greatly to emission

409 reductions of NO_x and VOCs, and dust control contributed 10% to emission reductions of $PM_{2.5}$.

410   When looking at specific industries, the power plants contributed most to the emission reductions of SO_2 and

411 NO_x at 49.7% and 46.9%, respectively, followed by the chemical industry, building materials industry, steel industry

412 and petrochemical industry with a total contribution from all four sectors to emission reductions of SO_2 and NO_x of

413 42.0% and 47.2%, respectively. For $PM_{2.5}$, the building materials industry contributed the most at 62.0%, followed

414 by steel and processing industry, power industry and non-ferrous smelting and process industry with a contribution

415 of 14.3%, 13.1% and 8.1%, respectively. For VOCs, the emission reduction sectors are mainly chemical,

416 petrochemical and machinery manufacturing sectors with a total contribution of 65.7% and individual contributions

417 of 25.1%, 23.2% and 17.4%, respectively. In addition, metal products processing, building materials and steel and

418 processing sectors also contributed significantly to emission reductions of 13.4%, 8.0% and 6.5%, respectively.

419   In terms of the regional distribution of emission reductions, Jiaxing, Hangzhou, Suzhou and Shaoxing have the

420 largest contribution of around 80%. These four cities contribute 87% to the total emission reduction of $PM_{2.5}$.

421   Combing all control measures, total emission reductions of SO_2, NO_x, $PM_{2.5}$ and VOCs are estimated to be

422 2872.5 tons, 3391.6 tons, 2467.6 tons and 5507.9 tons, respectively, which accounts for 10%, 9%, 10% and 11%,

423 respectively, of the total urban emissions. It is worth mentioning that if we consider the emergency emission

424 reduction measures for heavy pollution during the campaign, the amount of emission reduction for all pollutants

and the proportion of their emission reductions would be even larger. Table 4 shows the percentage and the amount

of emission reductions for pollutants under various control measures.

Table 4 Emission reduction estimations for various control measures

Province	City	Sector	Amount of emission reduction (tons)				Percentage of reduction			
			SO$_2$	NOx	PM$_{2.5}$	VOCs	SO$_2$	NOx	PM$_{2.5}$	VOCs
Zhejiang	Jiaxing	Industries and enterprises	925.6	709.5	462.3	1872.7	56%	58%	64%	80%
	Huzhou		414.8	585.6	602.5	514.0	46%	37%	47%	53%
	Hangzhou		657.2	654.1	476.2	1043.2	36%	42%	59%	33%
	Ningbo		59.1	65.3	107.5	84.0	32%	30%	37%	33%
	Shaoxing		365.9	414.8	403.9	678.7	34%	38%	62%	31%
Shanghai	Shanghai		253.6	368.7	83.6	796.1	9%	7%	6%	8%
Jiangsu	Suzhou		89.4	34.9	10.2	11.4	3%	1%	1%	1%
	Wuxi		94.4	163.0	10.2	55.3	12%	10%	1%	5%
Anhui	Xuancheng		7.8	68.8	9.1	0.0	15%	42%	28%	0%
Sub-total			**2867.8**	**3064.7**	**2165.5**	**5055.4**	**23%**	**19%**	**27%**	**19%**
Zhejiang	Jiaxing	Motor vehicles	2.3	157.7	16.4	211.3	46%	53%	38%	25%
	Huzhou		0.7	48.4	6.2	81.0	23%	24%	19%	12%
	Hangzhouy		1.7	120.8	13.5	160.2	8%	15%	20%	20%
Sub-total			**4.7**	**326.9**	**36.1**	**452.5**	**15%**	**25%**	**25%**	**19%**
Zhejiang	Jiaxing	Dust control	/	/	119.5	/	/	/	100%	/
	Huzhou		/	/	11.1	/	/	/	10%	/
	Hangzhou		/	/	26.6	/	/	/	10%	/
	Ningbo		/	/	28.8	/	/	/	5%	/
	Shaoxing		/	/	5.8	/	/	/	5%	/
Shanghai	Shanghai		/	/	69.3	/	/	/	6%	/
Jiangsu	Suzhou		/	/	2.7	/	/	/	1%	/
	Wuxi		/	/	1.8	/	/	/	1%	/
Anhui	Xuancheng		/	/	0.4	/	/	/	1%	/
Sub-total			/	/	**266.0**	/	/	/	**9%**	/
In total			**2872.5**	**3391.6**	**2467.6**	**5507.9**	**10%**	**9%**	**10%**	**11%**

3.4 Quantitative estimates of the contribution of control measures to air quality improvement

3.4.1 PM$_{2.5}$ concentration improvement in Jiaxing

The WRF-CMAQ air quality model, combined with observations, was used to evaluate the improvment of

PM$_{2.5}$ in Jiaxing due to the emission reductions achieved through the campaign. This analysis utilized two model

simulations to assess the impact of the emission reductions: 1) a baseline scenario, which utilized an uncontrolled

emission inventory (i.e., the emissions that would have occurred without the campaign), and 2) an emission

inventory, which reflects the emission reductions achieved by the campaign. Figure 13 shows the time series of

PM$_{2.5}$ observed concentrations and the percent change in PM$_{2.5}$ after the air quality control measures were

implemented. It can be seen that the PM$_{2.5}$ decline ratio in Jiaxing varies with time. The PM$_{2.5}$ decline ratio was

most significant during December 8-9 with a maximum reduction of 56%. During the campaign from December 8 to December 18, average $PM_{2.5}$ concentrations decreased by 10.5 μg/m³ with an average decrease of 14.4%. However, Although there are many control strategies implemented, the effects during 12/14-12/16 are low. As described in section 3.1.2, the prevailing wind direction during this period is NW, and Jiaxing experienced a heavy pollution process with the transit and transport of strong cold air. Therefore, we can not see obvious effect without strong upwind precursor emissions reductions.

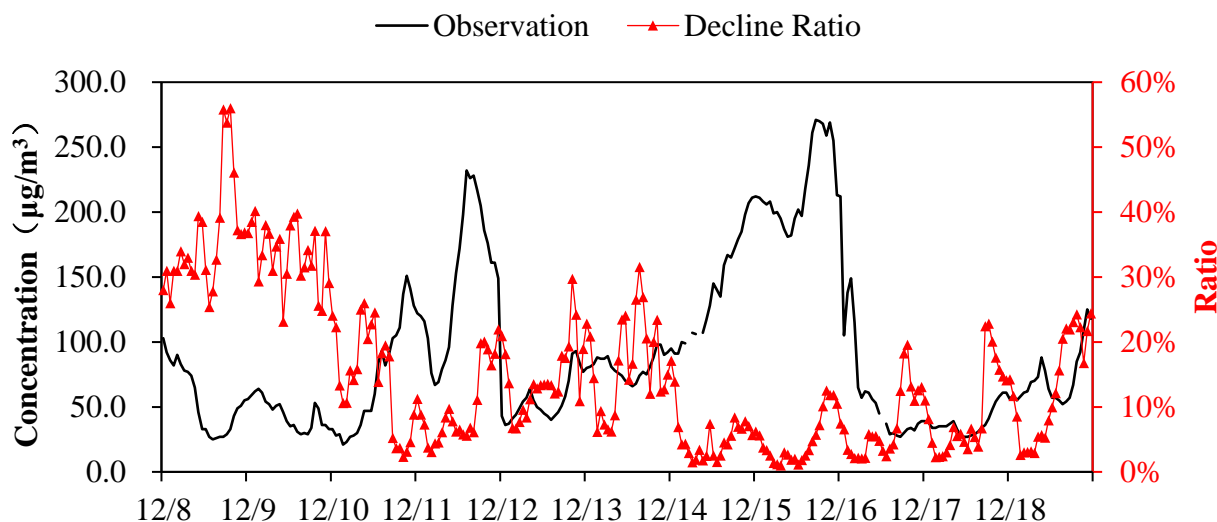
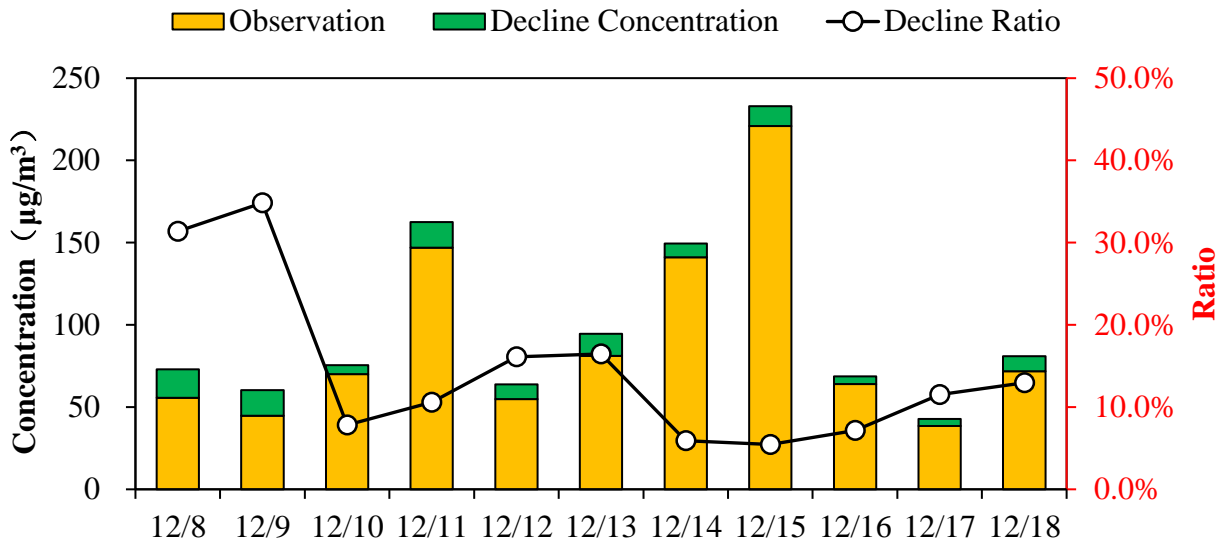



Fig. 13 Time series of observed $PM_{2.5}$ and the percentage reduction resulting from the implementation of air quality control measures

Figure 14 shows the reduction in daily average $PM_{2.5}$ concentrations in Jiaxing resulting from the emission reductions associated with the Action Plan for Air Quality Control during the World Internet Conference. As can be seen from the figure, the improvement in $PM_{2.5}$ before the conference (December 8 and 9) was relatively significant, with a daily average decline of roughly 31% and 35%, respectively, which corresponds to a decrease of around 17 μg/m³. The reduction in $PM_{2.5}$ during December 14-15, two of the days with some of the highest observed $PM_{2.5}$, was relatively low at around 6%, while daily average $PM_{2.5}$ concentrations on those days decreased by around 10.0 μg/m³. The magnitude of emission reductions during those two time periods was basically the same, so it's likely that the observed difference in $PM_{2.5}$ levels was the result of meteorological differences, and in particular, enhanced transport of polluted air into Jiaxing from December 14 to 15. Overall, under the influence of regional control measures for emission reductions from December 8 to December 18, $PM_{2.5}$ daily average concentration decreased by 5.5%-34.8% with an average of 14.6% or 10 μg/m³.

458
459
Fig.14 Percentage reduction in PM$_{2.5}$ resulting from the control measures

460   The decline ratio changes with meteorological conditions even under the same emissions reduction situation,

461   because meteorological conditions influence dispersion from primary emissions, regional transport and secondary

462   formation. The magnitude of emission reductions during those two time periods was basically the same, so it is

463   possible that the observed difference in PM$_{2.5}$ levels was the result of meteorological differences. Overall, the

464   residual PM$_{\textbf{2.5}}$ may come from three aspects: (1) although stringent control measures have been implemented, there

465   are still some precursor emissions in the city, which accumulated and formed secondary particles under favorable

466   meteorological conditions; (2) enhanced transport under specific meteorological conditions, especially upwind

467   emissions; (3) in view of the uncertainties of model performance (underestimation of PM$_{\textbf{2.5}}$, especially

468   underestimation of SOA) described in previous sections, it should be noted that the secondary formation may

469   probably be underestimated, causing modeled decline ratio lower than observed.

470   **3.4.2 PM$_{\textbf{2.5}}$ concentration improvement across regions**

471   Figure 15 shows the spatial distribution of PM$_{2.5}$ concentrations in the Yangtze River Delta region from

472   December 8 to December 18 in the baseline scenario and the campaign scenario. As can be seen from the figure,

473   southern Jiangsu, Shanghai and northern Zhejiang in the central YRD region had relatively high PM$_{2.5}$

474   concentrations, which is consistent with the typically more serious pollution levels in autumn and winter in the YRD

475   region. Under the influence of regional control measures, PM$_{2.5}$ average concentrations declined significantly in

476   Jiaxing, Hangzhou and Huzhou, especially at the junction of these three cities, with a slight improvement in central

477   southern Zhejiang as well. The average percentage PM$_{2.5}$ decline ratio in Jiaxing, Hangzhou and Huzhou was about

478   6%-20%. Meanwhile, given that the prevailing winds are north-westerly in winter, there was also some

479   improvement in central and southern Zhejiang.

(a)   (b)   (c)

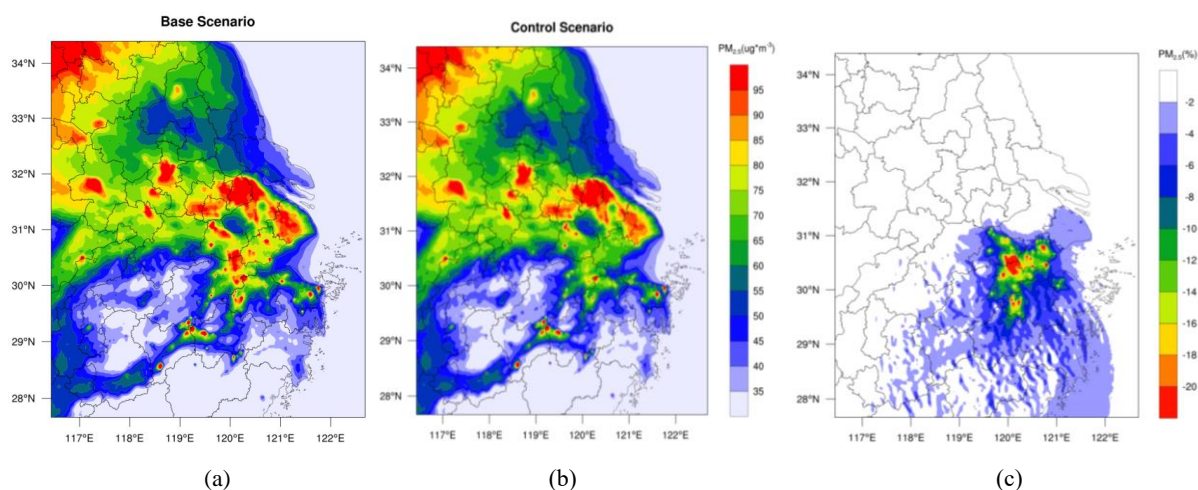



Fig. 15 Spatial distribution of PM$_{2.5}$ concentrations in the Yangtze River Delta region under the baseline scenario (a) and the campaign scenario (b), and the percentage reduction in PM$_{2.5}$ throughout the YRD region (c)

3.4.3 Regional contributions of PM$_{2.5}$ concentration improvement in Jiaxing

Figure 16(a) shows the percentage reduction in PM$_{2.5}$ daily average concentrations from December 13 to December 18 after control measures were implemented in Jiaxing and regionally. The reduction in PM$_{2.5}$ was the results of both local controls, as well as regional controls which reduced pollution in the air masses transported into Jiaxing. Overall, modelling suggests that the regional controls reduced PM$_{2.5}$ levels in Jiaxing between 5.5%-16.5% (9.9% average), while local control measures contributed 4.5%-14.4%, with an average of 8.8%.

Figure 16(b) shows the average contribution of local emissions reductions in Jiaxing and in the YRD region over the entire campaign (Dec.13-18), as well as the corresponding improvement in PM$_{2.5}$ levels in Jiaxing. During this period, PM$_{2.5}$ daily average concentration declined by 4-13 μg/m^3, while there were differences in the contribution of regional remission reductions and local emission reductions in Jiaxing during different periods. Overall, local control measure in Jiaxing had the largest impact on PM$_{2.5}$ levels and accounted for 89% of the decline in PM$_{2.5}$, while regional control measures contributed the remaining 11%.

(a)   (b)

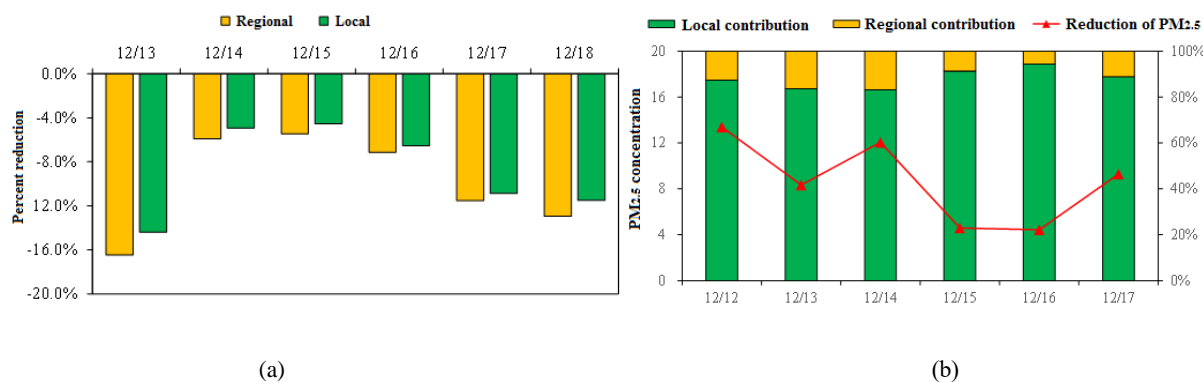



Fig. 16 Percentage reduction in daily average PM$_{2.5}$ concentrations from December 13 to December 18 after implementation of the control measures across the region and in Jiaxing (a) and Contribution of local and regional emissions reductions in Jiaxing, and the resulting improvement of daily average PM$_{2.5}$ concentrations in Jiaxing (b)

3.5 Optimisation scenario analysis of regional linkage control measures

3.5.1 Optimization scenario settings

In order to further analyse the optimisation potential of air quality control measures for major events and enhance the effectiveness of the control measure scheme design, three control measure optimisation scenarios have been set on the basis of the evaluation scenario (Base) after the implementation of air quality control measures during the conference. These scenarios include local emission reductions in Jiaxing under stagnant meteorological conditions, where local emission accumulation is the main contributor to the pollution process (Sce.1), and the emission reduction scenario where transport of polluted air masses into Jiaxing is a major contributor to the $PM_{2.5}$ levels in Jiaxing. In order to investigate the transport processes further, the latter scenario was further divided into a scenario 24 hours in advance (Sce.2) and a scenario 48 hours in advance (Sce.3). Table 5 describes the details of each scenario.

Table 5 Control measure optimization scenario settings

Scenario name	Scenario settings	Emission reduction regions	Emission reduction measures	Starting time
Base	Regional emission reduction	All the cities and areas involved in the campaign scheme	All control measures mentioned in the campaign scheme	December 8
Sce.1	Local emission reduction in Jiaxing	Jiaxing	Control measures in Jiaxing mentioned in the campaign scheme	December 8
Sce.2	Emission reduction through transport channels 24 hours in advance	Cities located in the northwest transport channel of Jiaxing	Cut down industrial sources by 30%	December 13
Sce.3	Emission reduction through transport channels 48 hours in advance	Cities located in the northwest transport channel of Jiaxing	Cut down industrial sources by 30%	December 12

Figure 17 shows the cities that primarily influence the polluted air masses transported into Jiaxing, where the transport channels were determined through backward trajectory analysis. These cities include Huzhou in Zhejiang province, Suzhou, Wuxi, Changzhou, Nanjing, Zhenjiang, Huai'an, Suqian and Suzhou in Jiangsu province and Suzhou, Huaibei, Bozhou, Bengbu, Chuzhou and Ma'anshan in Anhui province. Each of these cities took measures to reduce emissions by limiting production from industry industries by 30%.

519
520   Fig. 17 Cities involved in the transport channel and the emission reduction channel

521   The WRF-CMAQ modelling system was used to analyse and compare the air quality improvement effect under

522 different pollution process in four scenarios.

523 **3.5.2 Analysis of optimization scenario effects**

524   In order to evaluate the effect of the different starting time for the same control measures, and the same starting

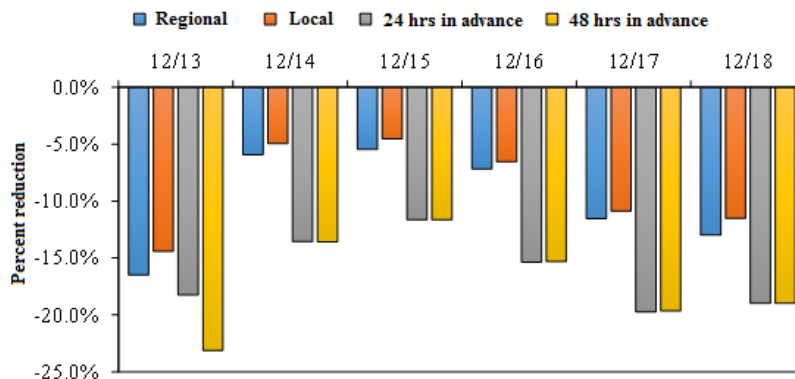
525 time for local and regional control measures, we investigated four scenarios. Figure 18 shows the percentage

526 reduction in daily average $PM_{2.5}$ concentrations in Jiaxing City from December 13 to December 18 under the

527 regional emission reduction scenario, the Jiaxing local emission reduction scenario and the transport channel

528 emission reduction scenario. Overall, there are differences in the distribution of $PM_{2.5}$ under the different scenarios.

529 The air quality improvement due to the regional emission reductions was higher than that of local emission

530 reductions in Jiaxing, and lower than that of channel emission reductions.

531
532   Fig. 18 Decline rates of PM2.5 daily average concentrations in Jiaxing under different scenarios

(1)Effect of local emission reductions in Jiaxing

By comparing the effect of local emission reductions in Jiaxing (Sce.1) and the effect of regional emission reductions (Base), we can see that $PM_{2.5}$ daily average concentrations in Jiaxing declined by around 5.5%-16.5% under the regional emission reduction plan (regional emission plan including the local emissions control) from December 13 to December 18 and by around 4.5%-14.4% under the local emission reduction plan. Local emission reductions in Jiaxing contributed 83%-94% to the emission reduction effect. Therefore, local emission reduction in Jiaxing is the key factor in improving the local air quality.

Compared with the channel emission reduction scenario 24 hours in advance (11.6%-18.2%), local emission reductions also contributed more than 50% to the improvement effect on December 13, 17 and 18. Therefore, local emission reductions contributed most to the air quality improvement effect in Jiaxing, indicating that local areas are still the most important control areas during the campaign.

(2)Effect of emission reductions through transport channels

As mentioned above, during the large-scale transport of heavily polluted air masses into the Yangtze River Delta region from December 14 to December 15, the $PM_{2.5}$ pollution in Jiaxing was significantly affected. Under the local emission reduction scenario (Sce.1) and the regional linkage emission reduction scenario (Base), $PM_{2.5}$ daily average concentrations in Jiaxing decline by only 4.5%-5.9%. If a 30% reduction in emissions from industrial sources in the upwind transport channel is implemented, $PM_{2.5}$ daily average concentrations in Jiaxing declined by 11.6%-13.6%, while local emission reductions contributed less than 40% to the improvement of $PM_{2.5}$. Therefore, to reduce $PM_{2.5}$ under these large-scale transport conditions, in addition to intensifying local emission reduction efforts, it is more effective to prevent and control such pollution by adopting emission reductions of industrial sources over key transport channels, especially for elevated sources.

In this study, the main transport channel involved is the northwest transport channel in control areas, which basically represents the typical winter transport channel in the region. In this study, the main transport channel involved is the northwest transport channel in control areas, which basically represents the typical winter transport channel in the region. Air quality improvement due to regional emission reductions was slightly larger than that of local emission reductions in Jiaxing, and smaller than that of channel emission reductions. This suggests that emissions reduction in the downwind cities does not have much effect on Jiaxing's air quality. In contrast, emissions reduction based on predicted transport pathway in advance are much more effective than local emissions reduction as well as regional emission reductions. Therefore, a well-designed management plan for the main transport channel is necessary to ensure optimized air quality improvement in autumn and winter, in addition to reducing local

563 emissions.

564 (3)**Effect of the starting time for channel emission reductions**

565 According to the comparisons between the emission reduction scenario 24 hours in advance (Sce.2) and the

566 emission reduction scenario 48 hours in advance (Sce.3) during the large-scale $PM_{2.5}$ transport, we can see that if

567 we take December 13 as the target and adopt channel emission reductions 48 hours in advance, $PM_{2.5}$ daily average

568 concentrations will decline by 23.1% when compared to the baseline scenario, which is significantly better than the

569 improvement achieved by the emission reduction scenario 24 hours in advance (18.2%). Therefore, early measures

570 to reduce emissions will lead to the improvement of air quality.

571 If we focus on the conference period (December 16-18), $PM_{2.5}$ daily average concentrations will both decline

572 by 15.3%-19.7% under the two channel emission reduction scenarios, indicating a close improvement effect.

573 Therefore, during the pollution process when local emissions are the main contributor, local emission reductions

574 should be the top priority with no difference between channel reductions 24 hours in advance and 48 hours in

575 advance. If transport is the main contributor to the pollution, adopting channel reductions 48 hours in advance can

576 bring about more improvement effect than 24 hours in advance.

577 **4 Conclusions**

578 **(1) The effect of restricting production in industrial enterprises is remarkable.** The power industry and

579 related industrial enterprises in Jiaxing cut down SO_2 and NO_x emissions by over 50%, while the building materials

580 industry, smelting industry and other industrial enterprises cut down $PM_{2.5}$ emissions by 63%, contributing greatly

581 to the reduction of primary $PM_{2.5}$ concentrations. The petrochemical industry, chemical industry and other related

582 industrial enterprises cut down VOCs emission by 66% in total, contributing greatly to the reduction of $PM_{2.5}$ formed

583 through the conversion of precursor species. The observation data of $PM_{2.5}$ components suggest that the relative

584 contribution of secondary components dropped significantly during the conference. Production restriction or

585 suspension for industrial enterprises is the main contributor to emission reductions for various pollutants during the

586 campaign, which resulted in the largest improvement in air quality.

587 **(2) Motor vehicle pollutant emissions declined significantly.** In Jiaxing, motor vehicle restrictions were fully

588 implemented during heavy pollution days, temporary traffic control was implemented during certain periods, and

589 enterprises and institutions had a three-day vacation during the conference. Emission reduction rates for various

590 pollutants from motor vehicle emissions were around 40%-50%. Motor vehicle emission reduction measures

591 contributed to the total emission reductions of nitrogen oxides by 18.2%, fine particles by 3.4% and volatile organic

592 compounds by 10.1%.

593   **(3) The effect of dust control measures is remarkable.** During the conference, most of the construction sites

594 in Jiaxing were suspended from operation. Increased frequency for road cleaning activities greatly lowered the dust

595 emissions. Speciation of the measured $PM_{2.5}$ suggest that the mass concentration of crust material, decreased by 14%

596 compared to measurements after the conference. Specially, under static conditions, mineral soluble irons (Ca^{2+} and

597 Mg^{2+}) declined 56.8% before and during the campaign. This suggests that the suspension of construction operations

598 and increased frequency of rinsing and cleaning of paved roads significantly reduced dust emissions.

599   **(4) Regional linkage between surrounding areas played an important role.** $PM_{2.5}$ is a typical regional air

600 pollutant, with obvious regional transport characteristics. In accordance with the requirements of the campaign

601 scheme, eight cities around Jiaxing have actively implemented emissions reduction measures. During the campaign,

602 $PM_{2.5}$ concentrations in eight surrounding cities and south-eastern Zhejiang also declined with obvious regional

603 synergies.

604   It is worth noting that the implementation of control measures has also had a negative impact on the economy

605 and the society in the short term while improving the air quality. For example, production restriction or suspension

606 on a large number industrial enterprises were taken at great economic costs, and motor vehicle restriction had a

607 large impact on the society.

608   **(5) Suggestions on emission reduction plans:** Local emission reductions shall be supplemented by regional

609 linkage. Assessment results show that local emission reductions play a key role in ensuring air quality. Therefore,

610 it is recommended that a synergistic emission reduction plan between adjacent areas with local pollution emission

611 reductions as the core part should be established and strengthened, and emission reduction plans for different types

612 of pollution through a stronger regional linkage should be reserved. Strengthen the pollution reduction in the upper

613 reaches along the transport channel. It is especially crucial to enhance pollution emission reductions in the upper

614 reaches of the channel since long-distance transport of plumes is a problem. This is especially true for key industrial

615 sources and elevated sources. Considering that polluted air mass transport is more frequent in winter, it is necessary

616 to develop emission reduction plans for different plume transport channels, combined with forecasting and warning

617 mechanisms which could be initiated on time.

618

619 **Author contribution**

620 L. Li designed this study and wrote the paper. H. L.Wang co-designed the study and provided valuable advice on

621 the data anlysis. C. Huang developed the regional emissions inventory. S. H. Zhu performed observational data

622 analysis and J. Y. An carried out the CMAQ and CAMx modelling work. R. S. Yan performed the WRF modelling.

623 M. Zhou and L. P. Qiao helped observation and data quality control. X. D. Tian and L. J. Shen carried out the

624 measurements and provided the observed data. L. Huang and Y. J Wang helped to revise the paper. Jeremy C Avise

and Joshua S Fu helped revise and polish the manuscript and gave advices on paper writing.

Competing interests

The authors declare that they have no conflict of interest.

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