1 WRF-CMAQ Two-way Coupled System with Aerosol

2 Feedback: Software Development and Preliminary Results

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12 Abstract

13 Air quality models such as the EPA Community Multiscale Air Quality (CMAQ) require 14 meteorological data as part of the input to drive the chemistry and transport simulation. The 15 Meteorology-Chemistry Interface Processor (MCIP) is used to convert meteorological data into 16 CMAQ-ready input. Key shortcoming of such one-way coupling include: excessive temporal 17 interpolation of coarsely saved meteorological input and lack of feedback of atmospheric 18 pollutant loading on simulated dynamics. We have developed a two-way coupled system to 19 address these issues. A single source code principle was used to construct this two-way 20 coupling system so that CMAQ can be consistently executed as a stand-alone model or part of 21 the coupled system without any code changes; this approach eliminates maintenance of separate 22 code versions for the coupled and uncoupled systems. The design also provides the flexibility to 23 permit users: (1) to adjust the call frequency of WRF and CMAQ to balance the accuracy of the 24 simulation versus computational intensity of the system, and (2) to execute the two-way 25 coupling system with feedbacks to study the effect of gases and aerosols on short wave 26 radiation and subsequent simulated dynamics. Details on the development and implementation 27 of this two-way coupled system are provided. When the coupled system is executed without 28 radiative feedback, computational time is virtually identical when using the Community 29 Atmospheric Model (CAM) radiation option and a slightly increased (~8.5%) when using the

1 Rapid Radiative Transfer Model for GCMs (RRTMG) radiation option in the coupled system 2 compared to the offline WRF-CMAQ system. Once the feedback mechanism is turned on, the execution time increases only slightly with CAM but increases about 60% with RRTMG due to 3 4 the use of a more detailed Mie calculation in this implementation of feedback mechanism. This 5 two-way model with radiative feedback shows noticeably reduced bias in simulated surface 6 shortwave radiation and 2-m temperatures as well improved correlation of simulated ambient 7 ozone and PM_{25} relative to observed values for a test case with significant tropospheric aerosol 8 loading from California wildfires.

9

10 **1 Introduction**

11 3-D chemical transport models (CTMs) that are used in air quality research and regulatory applications 12 are driven by 3-D meteorological fields provided by a priori runs of a meteorology model. Historically, 13 the CTMs and meteorological models were developed over several decades along independent tracks 14 with little regard for computational, numerical, or even scientific consistency between the two 15 modeling systems. In recent years, however, there have been several efforts to combine meteorological 16 and chemical transport models into single interactive systems (Grell and Baklanov, 2011). A primary 17 driver for this trend has been the need to include the direct and indirect feedback effects of gases and 18 While these feedback effects are mainly important for climate aerosols on radiative forcing. 19 applications, it is becoming evident that they have substantial effects on local meteorology and air 20 quality in polluted regions (Jacobson et al., 1996, Mathur et al., 1998, Xiu et al., 1999). Zhang (2008) 21 has provided an overview of several coupled meteorology-chemistry models including the WRF/chem 22 (Grell et al., 2005) model in which chemistry has been added into the Weather Research and 23 Forecasting model (Skamarock et al., 2008) at the science process level. Another approach is to 24 couple historically independent meteorology and chemical transport models into a single executable. 25 Advantages of this approach include maintaining consistency with existing separate loose coupled 26 meteorology-chemistry systems that are being continuously and extensively applied and evaluated. 27 Furthermore, the numerical and computational techniques employed in meteorology models and CTMs 28 differ considerably because of the greater need for strict mass conservation and positive-definiteness of 29 transported scalars in the CTM. Also, CTMs generally use fractional integration of various processes 30 while meteorology models use time split integration of all process rates.

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32 The development of the Community Multiscale Air Quality (CMAQ) (Byun and Schere, 2006)

1 modeling system was initiated in the 1990s by the U.S. Environmental Protection Agency (EPA); and 2 the system has continued to evolve (Foley et al., 2010). The model system has benefited from a diverse 3 user community with over 2000 users from 90 different countries. CMAO has been and continues to be 4 extensively used to provide guidance in rulemaking such as CAIR (Clean Air Interstate Rule, 5 http://www.epa.gov/cair/), by state and local agencies for air quality management analyses such as SIP 6 (State Implementation Plan), by academia and industry for studying relevant atmospheric processes and 7 model applications. CMAQ has also been adapted into the real-time US National Air Quality 8 Forecasting system (AQF) (Otte et al., 2005) and has been running operationally at National Weather 9 Service since 2003 and was recently deployed for forecasting air quality for the 2010 Shanghai World 10 Expo (Wang et al., 2010).

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12 In general, meteorological models are not built for air quality simulation purposes. Hence, the 13 meteorological model might not have the same map projection, coordinate system and grid format, and 14 layer structure as the air quality model. The CMAQ model uses the Meteorology-Chemistry Interface 15 Processor (MCIP) (Otte and Pleim, 2010) to bridge this gap by providing transformed CMAQ-ready 16 meteorological data. The transformation includes unit conversion, format conversion, vertical grid 17 resolution related interpolation, as well as calculations to create additional diagnostic variables that are required in CMAQ but not available in the meteorology model output. Typically, MCIP produces 18 19 hourly meteorological data, based on storage requirement considerations, as input to CMAQ.

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21 The flow of information in this one-way coupled system is (Figure 1a): run a meteorological model, 22 like the Fifth-Generation Pennsylvania State University-National Center for Atmospheric Research 23 Mesoscale Model (MM5) (Grell et al., 1994) or the Weather Research and Forecasting (WRF) model 24 (Skamarock 2008, Michalakes et al., 2005), process the meteorological model's output using MCIP, 25 then run the CMAQ air quality model using the MCIP output. This whole coupling approach has been 26 widely used in the research community as well as in the real-time National Air Quality Forecasting 27 system (Otte et al., 2005), but it has several potential shortcomings. First, the integration time step of 28 CMAQ is much finer than the typical hourly available meteorological data. Interpolation is used to 29 handle this issue; however, interpolation accuracy is a problem for meteorological variables, such as 30 wind direction and speed that are key variables for pollutant transport and dispersion. Additionally, at 31 fine horizontal resolutions (<10 km), the need for more frequent meteorological information (relative to 32 the typical hourly resolution) becomes critical for consistently representing transport (advective,

1 turbulent, and cloud) processes. Second, the lack of aerosol feedback from CMAQ to the 2 meteorological model is an important omission. For instance, CMAQ is able to compute the 3 concentration, composition, and size distribution of particulate matter (aerosol) in the atmosphere. The 4 presence of aerosols in the atmosphere affects the radiation which in turn affects the photolysis rates 5 which dictate atmospheric photo-chemistry, surface temperature that can affect thermally driven 6 atmospheric chemical reactions, planetary boundary layer (PBL) height which dictates dilution and 7 dispersion of pollutants, and even cloud formation. The response of the meteorological model to 8 aerosol loading can be significant under conditions of significant pollution loading. Without such 9 feedback, errors may be introduced into the meteorological model. Third, netCDF file format (32-bits precision) is used to store the intermediate data, (i.e. meteorological model and MCIP output), with 10 11 repeated reading and writing of data to/from intermediate files, resulting in reduced data accuracy due 12 to truncation error.

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14 To address these potential shortcomings, we coupled the WRF meteorology model and the CMAQ 15 model to create a two-way coupled modeling system to facilitate feedbacks between chemistry and 16 meteorology. Feedback can be categorized as direct or indirect effect: the former deals with how 17 aerosol affects the radiation and the latter considers how aerosols affects cloud formation and duration resulting from scattering and absorption as well as acts as CCN which impact cloud optical thickness 18 19 and cloud lifetime. In this article, we focus on the direct effect. Section 2 provides an overview of the 20 scientific components in the model. Section 3 describes the software considerations for developing a 21 flexible and efficient coupled modeling system, including domain decomposition and design issues. 22 Preliminary results are presented in Section 4 while Section 5 summarizes the main results and presents 23 a brief discussion of future work.

24

25 **2** Overview of Scientific Components of the Coupled model

The coupled modeling system consists of three components: the WRF meteorology model, the CMAQ model and the coupler. A high level view of the system is depicted in Figure 1b. A detail description of each component is provided below.

29

30 2.1 WRF

31 The Advanced Research WRF version 3 (WRF-ARW) is a state-of-the science mesoscale meteorology

32 model (Skamarock et al., 2008) that is typically configured with horizontal grid resolutions ranging

1 from 1-30 km, but WRF is also being used on Large Eddy Simulation Scale (dx 100m or smaller) 2 (Moeng 2007) as well as on global scales. The dynamical equations numerically solved by the WRF-3 ARW model are fully compressible, Euler nonhydrostatic, and are conservative for all scalar variables. 4 The prognostic variables are the three velocity components, perturbation potential temperature, 5 perturbation geopotential, and perturbation dry air surface pressure. Additional prognostic variables 6 depend on the model physics options and may include turbulent kinetic energy, water vapor mixing 7 ratio, and several cloud microphysical scalars such as cloud water/ice mixing ratio, rain/snow mixing 8 ratio, and graupel mixing ratio. Both the WRF-ARW and the CMAQ model can be configured to use 9 the exact same grid configurations and coordinate systems. Thus, no spatial interpolation of either 10 meteorological or chemical data is required.

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12 2.2 CMAQ

13 CMAQ version 4.7.1 (Foley et al., 2010) is a comprehensive atmospheric chemistry and transport 14 model that numerically integrates a set of independent chemical conservation of mass equations on a 15 series of 3D nested Eulerian grid meshes. The CMAQ model employs operator splitting to modularize 16 the various physical and chemical processes including: subgrid turbulent vertical transport, horizontal 17 and vertical advection, horizontal diffusion, cloud processes (i.e. aqueous chemistry, subgrid convective transport, wet deposition), gas-phase chemistry, and aerosol chemistry and dynamics. The 18 19 CMAQ system includes anthropogenic emission rates processed by the Sparse Matrix Operator Kernel 20 Emissions (SMOKE) [http://www.cep.unc.edu/empd/products/smoke]. Plume rise, biogenic emissions 21 and dry deposition are modeled by components of the CMAQ model. Both sources (emissions) and 22 sinks (deposition) are applied as mass tendencies in the vertical diffusion calculation.

23

24 **2.3 Coupler**

The coupler is used to link these two models together and serves as an inter-model translator. The design and functionality will be described in the next section. The coupler also includes software (aqprep) to transfer meteorological fields from WRF to CMAQ and to transfer aerosol predictions from CMAQ back to WRF (feedback).

29

A subroutine called *aqprep* prepares meteorological fields in forms compatible for use in CMAQ's generalized coordinate formulation. The preparation includes extracting data such as pressure and wind field directly from WRF and calculating additional variables that are used in CMAQ such as the

vertical coordinate Jacobian and the fractional area of each land use category in each grid cell. In
 essence, aqprep includes the functionality currently embodied by the MCIP (Otte and Pleim, 2010)
 preprocesses in the offline WRF CMAQ system.

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5 An important benefit of two-way coupling between meteorology and air quality models is the ability to 6 use aerosol fields simulated by the air quality model to affect processes in the meteorology model. The 7 first feedback implemented in the WRF-CMAQ system is the direct effects by which chemical species 8 calculated in CMAQ are transferred to WRF for calculating their influence on radiation computed in 9 WRF. In addition to the data coupling described in the next section, implementation of direct feedback requires a new subroutine for the calculation of the aerosol optical properties: extinction optical depth, 10 11 single scattering albedo, asymmetry parameter, and forward scattering fraction, for short-wave spectral 12 bands (19 bands in Community Atmosphere Model (CAM) and 14 bands in Rapid Radiative Transfer 13 Model for GCMs (RRTMG)). The aerosol chemical species calculated by CMAQ are combined into 14 five groups: water-soluble, insoluble, sea-salt, black carbon, and water. The refractive indices for these 15 species are taken from the OPAC (Optical Properties of Aerosols and Clouds) (Hess et al., 1998) 16 database using linear interpolation to the central wavelength of the RRTMG wavelength intervals. 17 These direct feedbacks tend to reduce SW radiation reaching the ground in areas of high aerosol loading, thereby reducing daytime surface temperatures, as shown in Section 4. In addition, absorbing 18 19 aerosols, such as black carbon, tend to warm the air in layers with high concentrations. There are also 20 secondary effects on PBL heights and cloud properties.

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In WRF-Chem (Fast et al. 2006) the optical processes are done by calculating extinction, scattering and asymmetry factor by summing a parametric method originally developed for a modal approach by Ghan et al. (2001). This approach uses first calls a full Mie code to calculate optical properties over exponentially spaced intervals of x (2* pi * radius / wavelength) for a set of seven refractive indices. A polynomial fit is made for extinction, scattering, and asymmetry factor for each of the refractive indices. All subsequent calls extinction, scattering and asymmetry factor use the polynomial approximations. The integral properties are calculated by summing overall size bins.

30 In our approach, feedback effects from chemical species calculated by CMAQ are transferred to WRF 31 for calculating the influence of these species on the heat balance computed by WRF. The new aerosol 32 codes to be used with CAM and RRTMG calculate the aerosol extinction, single scattering albedo, and asymmetry factor for short-wave (SW) radiation and aerosol extinction for long-wave (LW) radiation.
 The aerosol chemical species calculated by CMAQ are combined into five groups: water-soluble,
 insoluble, sea-salt, black carbon, and water. The refractive indices for these species are taken from the

4 OPAC database using linear interpolation to the central wavelength of the CAM and RRTMG

5 wavelength intervals.

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The efficient Gauss-Hermite numerical-quadrature method calculates the extinction and scattering
coefficients along with the asymmetry factor by integrating the Bohren & Huffman Mie codes over the
log-normal size distributions representing the Aitken, accumulation, and coarse modes produced by
CMAQ.

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12 **3 Software Considerations for the Coupler**

An air quality model may utilize a different map projection, time integration, grid orientation, grid cell size, and/or vertical coordinate that is different from its meteorological driver. In order to facilitate communication between models to exchange relevant information that is usable by each individual model, a coupler is devised.

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18 **3.1 Modeling domain**

19 Both the WRF-ARW and CMAQ use the Arakawa C horizontal grid staggering, and in the coupler no 20 spatial interpolation of meteorological or chemical data is required. In addition, both WRF and CMAQ 21 use the same map projection, so the coupler inherits the map projection from WRF. The vertical 22 coordinate in WRF is a hydrostatic sigma-pressure, and CMAQ uses a modified, generalized form of 23 that coordinate. Unlike in the offline WRF-CMAQ system, the coupler must use the same number and 24 configuration of vertical layers (i.e., no layer collapsing, Otte and Pleim, 2010). Figure 2 illustrates the 25 typical domain configurations of the WRF-CMAQ coupled system; in this the chemistry-transport 26 calculations are performed for a sub-domain of the larger WRF domain. In typical WRF applications, 27 to provide a transition from externally specified lateral boundary conditions, a relaxation zone is 28 specified where the model is relaxed toward the large-scale forecast (Skamarock et al., 2008). The 29 coupler allows the users to specify how many grid cells to trim off for the chemistry-transport 30 calculations at run time, but five grid cells is the recommended minimum to avoid numerical artifacts 31 that commonly occur in the WRF boundary relaxation zone. Horizontal transport calculations in 32 CMAQ only require the specification of species concentrations (time varying or independent based on 1 user specification) at a one-cell thick boundary along the CMAQ domain. Thus, under conditions of 2 inflow, the concentrations specified at these upwind boundary cells are used to estimate the advective 3 flux (following the PPM formulation) into the domain. Users can define any CMAO domain as long as 4 that domain fits within the maximum CMAQ domain depicted in Figure 2. In addition, the user is 5 required to provide the value of delta_x and delta_y which defines the lower left corner of the CMAQ 6 domain relative to the WRF domain. Because the WRF model is the driver and CMAQ is called as a 7 subroutine within WRF, a global timer that is based on the WRF advection time step is used to 8 synchronize WRF and CMAQ in the coupled system.

9

10 **3.2 Domain decomposition**

11 The main task of the coupler is to transfer needed data between these two models correctly. Both 12 models were designed to run in a parallel computing environment. WRF supports MPI and OpenMP 13 but the current version of CMAQ only supports MPI. They both use domain decomposition as the basic 14 parallelization approach. However, the details of the decomposition are quite different in both models. 15 Runtime System Library (RSL) (Michalakes, 1994) and RSL-lite (Michalakes, 1998), which both 16 handle high-level stencil and inter-domain communication, irregular domain decomposition, automatic 17 local/global index translation, distributed I/O, and dynamic load balancing, are used in WRF to parallelize the code. RSL-lite is a bit faster than RSL. Both RSL and RSL-lite were used in WRF 18 19 version-2 implementation, but RSL has been removed since version 3.

20

Besides performance, the main difference between RSL and RSL-lite is the domain decomposition algorithm. Since RSL has been removed, the description of the decomposition algorithm is focused on RSL-lite. The mapping between processor and sub domain is in row-wise fashion. The starting point and order of assigning the remainder row elements is at the bottom and then top, and then bottom until all the remaining elements are distributed (Fig. 3a). Similarly, for the column dimension, it starts at the left and then right and moving towards the center.

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In CMAQ, the domain decomposition uses the same processor and sub-domain mapping as in WRF (Fig. 3b), but the starting point and order of assigning the remainder column or row elements is different. The remainder elements starts from the bottom and moves toward the top for row dimension and starts from left and moves toward the right for column dimension, rather than alternating inward from the top/bottom and left/right.

2 When CMAQ is executed, users can choose a particular processor configuration based upon the 3 number of processors allocated. For instance, if the number of available processors is 16, the user can 4 choose between a 4x4, 8x2, 2x8, 16x1 or 1x16 processor configuration. In WRF, the processor 5 configuration with a "square" orientation, is the default but it is user-definable. In the WRF-CMAQ 6 coupled system, CMAQ's processor configuration is inherited from WRF. The coupler's main task is to 7 compute the mapping between the WRF and CMAQ domains with respect to each sub domain with 8 consideration of the position of the CMAQ domain relative to the WRF domain. This mapping 9 information will be used for data transfer between these two models in the forward and feedback steps.

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11 **3.3 Data exchange**

12 We have considered different tools, such as ESMF (Earth System Modeling Framework) 13 (http://www.earthsystemmodeling.org/), Cpl6 (Craig et al., 2005) and MCT (Larson et al., 2005), for 14 data exchange between the two models. CMAQ uses the IOAPI3 (Input/Output Applications 15 Programming Interface version 3, http://www.baronams.com/products/ioapi/AVAIL.html) to handle 16 physical file I/O. With the consideration of minimal code change, we chose to use IOAPI3 for the 17 coupling. The actual data transfer is performed in memory through IOAPI3 buffered files. IOAPI3 is third party software written by Baron Advanced Meteorological Systems (BAMS). It is written to 18 19 handle various types of files: volatile real files which are used in CMAQ to deal with I/O, using the 20 netCDF format; buffered virtual files which facilitate data exchange within the same program through 21 memory; coupling-mode virtual files which use the PVM3.4 mailbox mechanism to exchange data 22 among models executed concurrently; and native-binary real files which are the same as the volatile 23 real files except the file is stored in native binary format instead of netCDF.

24

25 The type of file, whether it is a volatile real file or a buffered virtual file, used in the application is 26 determined at run time. The capability of handling various file types within IOAPI3 is transparent to 27 the user application code; hence, code modification is not needed. The coupler will create/open the 28 same number of files as in the offline run that uses physical files. In the stand-alone CMAQ model, 29 each processor is able to access the entire file and only extract relevant data for the sub-domain portion 30 from the file. In addition, hourly meteorological input is interpolated to the current time step in various 31 science processes within CMAQ throughout the execution. In order to make the same code work for 32 buffered files while reducing memory consumption, each buffered file is exactly the same size as a subdomain and corresponds to that sub-domain only. Two time steps of data are stored in each circular
 buffered file.

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4 **3.4 System structure**

5 The coupler consists of two major components: *aqprep* and *feedback*. The prepared data is placed in the 6 corresponding buffered files which have similar attributes as the physical files used in the uncoupled 7 stand-alone CMAQ. In CMAQ, the same IOAPI3 calling interface as in the stand alone model is used 8 to access these buffered files. This design provides flexibility to read and write either buffered or disk 9 files, enabling consistent coupled and uncoupled modeling paradigm. Thus an inherent advantage of using the IOAPI3 to handle the file format in the coupler is that minimal code changes are needed in 10 11 CMAO. The feedback part is called within the aerosol module in CMAO and computes several 12 variables that are needed for direct aerosol feedback to the WRF radiation module. Various information 13 such as coarse mode diameter and Aitken mode natural log of standard deviation, is used to compute 14 soluble mass, elementary carbon mass and other parameters.

15

16 WRF integrates at a very fine time step, e.g. one minute for 12 km horizontal grid cell size. In CMAQ 17 each physical process, (e.g. transport and chemistry), has a different time step requirement that is based on individual process characteristic time scales and numerical stability criteria. As a stand-alone 18 19 model, CMAQ determines the minimum synchronization time step based on the horizontal wind speed 20 Courant condition in model layers lower than ~700 hPa which generally allows for CMAQ's 21 synchronization time step to be several times greater than the WRF time step. In coupled system, users 22 can choose the call frequency at run time as a ratio between the WRF and CMAQ time steps. For 23 instance, if the ratio is set to four and the WRF time step is 30 seconds, the CMAQ time step will be 24 two minutes. Consequently, the computational burden for the coupled system increases substantially as 25 the CMAQ calling frequency increases. The non-linear increase in computational intensity is related to 26 inherent non-linearity in atmospheric processes and numerical solution of the governing equations. 27 Figure 4 depicts the calling sequence for the coupled system. In general, CMAQ is called after an 28 apprep step except the very first time. This implementation ensures two steps of WRF data are always 29 available in case temporal interpolation of meteorological information is needed in CMAQ. The 30 feedback step takes place in CMAQ within the aerosol module. Since CMAQ is a subroutine of WRF it 31 occupies a portion of a WRF step. There are two versions of the radiation calculation within the two-32 way coupling model: one comes with the WRF code (Ra) and the other was modified to include feedback capability (Ra'). Ra' is not called until feedback information is available. That is why Ra is
 used for the very first step.

3

4 **3.5** Software modification in WRF and CMAQ to support the coupled system

5 In the coupled system, CMAQ is implemented as a subroutine in WRF. Since CMAQ is a community 6 model with a wide user community which has used the model in an offline mode, all the coupling 7 related functions are encapsulated in Fortran 90 modules that will not be invoked when running CMAQ 8 in a stand-alone offline mode. Thus, the same version of CMAQ can be consistently used both in an 9 offline or coupled system mode. The remainder of the necessary modifications to CMAQ to enable 10 online coupling are transparent to the user. Whether to include the coupling portion or not in CMAQ is 11 decided at compilation time. This fulfills the single source design principle and eliminates software 12 maintenance of separate model versions for on-line and off-line configurations.

13

The aerosol optical depth for nine selected wavelength bands estimated from the simulated aerosol distribution is added to the WRF output for examination and validation purposes. This requires nine new variables in the WRF Registry. Two routines in WRF were also modified for the coupled system: solve_em.F (to invoke CMAQ) and the radiation calculations (to add the nine variables which are passed into various parts of the radiation calculation and to process aerosol feedback). These modifications can be easily ported to newer versions of WRF as its science is updated.

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Run time switches have been implemented to disable the aerosol feedback in the coupled system as well as to run WRF in stand-alone mode (i.e., without calling CMAQ). These options provide flexibility to perform sensitivity studies on the effects of the feedback mechanism and the coupling

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4 Preliminary Performance of the coupled system

The performance of the WRF-CMAQ two-way coupled system is evaluated for both computational impact and scientific advancement. In order to be of general use to the CMAQ community, the coupled system must not add such a computational burden that it would become prohibitive for the average user to run. In addition, it is desirable that the coupled system scales well computationally as processors are added to the configuration. Of equal importance is the need for the coupled system to demonstrate a scientific advantage beyond what could be achieved by using the WRF and CMAQ models sequentially.

2 **4.1 Computational performance**

3 We conducted a series one-day simulation (1 August 2006) to examine and quantify the computational 4 performance of the coupled system. For these tests, we used a domain encompassing Eastern US, 5 discretized with 12-km horizontal grid spacing while the vertical extent ranging from the surface to 50 6 hPa was discretized using 34 layers of variable thickness. Here, the WRF domain size is 290×251 7 grid points, and the CMAQ domain size is 279×240 grid cells, which allows for the five-cell boundary 8 along the perimeter of the WRF domain to be excluded from CMAQ (Fig. 2). WRF-ARW version 3.1 9 was built with CMAQ version 4.7.1 to form the WRF-CMAQ two-way coupled system. Initial and lateral boundary conditions for WRF were derived from a combination of North American Mesoscale 10 11 (NAM) model analyses and forecasts at 3-h intervals that were developed by the National Center for 12 Environmental Prediction and obtained from the National Climatic Data Center. In WRF, the model 13 options included the WRF single-moment 6-class (WSM6) microphysics scheme (Hong et al., 2004), 14 version 2 of the Kain-Fritsch (KF2) cumulus cloud parameterization (Kain 2004), the Asymmetric 15 Convective Model version 2 (ACM2) for the planetary boundary layer (Pleim, 2007a,b), and the Pleim-16 Xiu land-surface model (Xiu and Pleim, 2001) with soil moisture and temperature nudging (Pleim and 17 Xiu, 2003; Pleim and Gilliam, 2009). Both the Rapid Radiative Transfer Model for GCMs (RRTMG) 18 (Clough et al., 2005) and Community Atmospheric Model (CAM) (Collins et al., 2004) radiation 19 schemes were tested and run in the coupled system to contrast the simulated impact of radiative 20 feedback from CMAQ to WRF using multiple radiation schemes. Also in WRF, analysis nudging was 21 included for temperature and humidity above the PBL and for winds at all model levels (Stauffer et al., 22 1991).

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24 In CMAQ, the CB05 gas-phase chemical mechanism and the modal aerosol model known as AERO5 25 (Carlton et al., 2010) were used. The same subgrid vertical turbulent transport of meteorological and 26 chemical species was used in both WRF and CMAQ following the ACM2 PBL scheme. The sub-grid 27 convective cloud scheme in CMAQ, which is responsible for convective transport of chemical species, 28 aqueous chemistry, and wet scavenging, is a simple bulk scheme based on the convective cloud model 29 in the Regional Acid Deposition Model (RADM; Chang et al., 1987) but with convective transport 30 based on the Asymmetric Convective Model (Pleim and Chang, 1992). Since the CMAQ cloud scheme 31 uses the convective precipitation rate to diagnose sub-grid mass fluxes, the location and timing of 32 precipitating convective clouds are consistent with WRF. A new convective cloud scheme based on the

1 Grell scheme in WRF (Grell and Devenyi, 2002) is being tested to improve consistency across 2 chemical and meteorological components of the system. Note that WRF and CMAQ use different 3 scalar advection schemes that are both monotonic and positive definite for meteorological and chemical 4 species. However, differences in numerical formulations and time steps allow subtile differences in the 5 3-d mass fields to accumulate over time. Mass conservation and consistency between chemical 6 concentrations and air density is ensured in CMAQ by adjustment of the vertical velocities according to 7 a layer-by-layer solution of the 3-d mass continuity equation at every time step. In this design 8 chemical species are advected in CMAQ by an efficient scheme that has very little numerical diffusion: 9 the piecewise parabolic method (PPM) (Colella and Woodward 1984). A potential drawback of this 10 approach is the inconsistencies between advective transport of microphysical scalars in WRF and 11 advection of gas and aerosol species in CMAQ. While such discrepancies are likely very small they 12 could be important for modeling aerosol indirect effects which result from interactions between 13 aerosols and cloud microphysics. The significance of these inconsistencies will be assessed as 14 implementation and testing of indirect aerosol processes continues.

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For these tests, a one-minute WRF time step was used, and CMAQ was called every five WRF time
steps (ratio of 5:1). The simulations were run on 32 processors on a Linux cluster.

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19 Table 1 presents the execution time of the offline WRF-CMAQ system, the WRF-CMAQ two-way 20 coupled system with and without radiative feedback. When the coupled system is executed without 21 radiative feedback (but with increasing the temporal frequency of the WRF meteorological fields 22 available for CMAQ), computational time is virtually identical when using the CAM radiation option 23 and a slightly increased (~8.5%) when using the RRTMG radiation option compared to the offline 24 WRF-CMAQ system. Once the feedback mechanism is turned on, the execution time increases only 25 slightly with CAM but increases about 60% with RRTMG. The numerical techniques used to compute aerosol optical characteristics (extinction, scattering, and asymmetry factor) with the Mie approach 26 27 used in RRTMG are much more computationally intensive than the Mie approximation (Evans and 28 Fournier, 1990) used in the CAM implementation. However, this new scheme used for RRTMG is 29 more accurate and robust over a wider range of refractive indices.

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The individual models used in the coupled system, WRF and CMAQ, are fully parallelized. As a result, the scalability of the coupled system is inherited from both components. By doubling the processors from 32 to 64 on this domain, a speedup of ~1.6 was achieved for both CAM and RRTMG configurations, regardless of whether the radiative feedback was enabled (Table 2). Increasing the number of processors by a factor of four (32 to 128) resulted in a speedup of ~2.3-2.7. The addition of radiative feedback in the coupled system does not adversely affect the scalability using either CAM or RRTMG, but results in a greater relative speedup on more processors.

6

7 **4.2 Scientific performance**

8 To examine the scientific performance of the coupled system with radiative feedback, we conducted a 9 ten-day simulation (20 Jun - 29 Jun 2008) of a wildfire event in California. Widespread wildfires (Fig. 10 5) resulted in significant particulate matter (PM) pollution during mid-late Jun 2008 in California and 11 surrounding states. The coupled model using the options discussed in Section 4.1, was applied to a 12 domain which covers California and portion of the surroundings states (Fig. 6); the RRTMG radiation 13 scheme was used and the vertical extent up to 100mb was discretized using 22 layers. This simulation 14 the fire emission uses latest data 15 (ftp://ftp.epa.gov/EmisInventory/2005v4/2007emis/smartfire and bluesky enabled methodology 2006 2008.pdf). Model 16 results from simulations with and without radiative feedback were compared to ground-level 17 meteorology measurements from the Meteorological Assimilation Data Ingest System (MADIS), 18 radiation measurements from the Integrated Surface Irradiance Study (ISIS) Network, and 19 concentration measurements from the U.S. Environmental Protection Agency's Air Quality System 20 (AQS) network.

21

22 Figure 6 presents an illustration of the direct aerosol feedbacks simulated by the coupled WRF-CMAQ 23 modeling system during the time period of these wild fires in California. In general relatively high 24 aerosol optical depths are noted in regions of high surface and boundary-layer particulate matter 25 pollution (Figures 6a and b). Shown in Figure 6c is the difference in surface shortwave radiation 26 between a run with aerosol feedback and one without, while Figure 6d presents an illustration of a 27 similar difference in the modeled planetary boundary layer (PBL) height at the same time. As 28 illustrated, aerosol direct radiative effects associated with scattering and absorption of incoming 29 radiation, result in a reduction of short-wave radiation reaching the surface (Figure 6c), which then 30 translate to reduction in temperature at the surface as well as a reduction in PBL height (Figure 6d). 31 These effects are particularly pronounced in regions with high aerosol loading with simulated reductions of over 250 Wm⁻² in instantaneous surface shortwave radiation and corresponding 32

- 1 reductions of over 500m in PBL heights.
- 2

3 Figure 7 presents comparisons of the surface shortwave radiation simulated for the cases with and 4 without direct aerosol feedback with measurements at an Integrated Surface Irradiance Study (ISIS) 5 site in Hanford, California (see Fig. 5, red star). The observations show that there was significant reduction of incident shortwave radiation at the surface (peak observed values of 900-1000 Wm⁻²) on 6 24, 26 and 27 Jun due to the smoke plumes from the wildfires. Without aerosol feedback effects, the 7 model overestimates shortwave radiation by ~100-200 Wm⁻². When the aerosol feedbacks were 8 9 included, the model bias was significantly reduced, though a slight overestimation still persists. Thus 10 including the aerosol feedback in the coupled system is important for better simulating the shortwave 11 radiation fields in WRF.

12

13 The presence of aerosols from the wildfires reduces the shortwave radiation at the surface, which also 14 acts to reduce the maximum daytime temperatures near the surface. Figure 8 shows a comparison of 2-15 m temperatures averaged from four sites in the Sacramento Valley, (Oroville, Red Bluff, Redding, and 16 Sacramento, see blue triangles in Figure 5) with model simulations for the cases with and without 17 radiation feedback, for the ten-day simulation period. Typically the 2-m temperature was 18 overestimated in the simulation that did not consider any aerosol feedback effects. This overestimation 19 was reduced in the simulation with the aerosol feedbacks. For example, on 25, 26, 27 and 29 June, 20 when the wildfires were most actively affecting these sites, inclusion of the aerosol feedback reduced 21 or eliminated the persistent over prediction evident in the simulation without feedback.

22

23 Figure 9 presents comparisons of the day time (8am to 6pm local time) model and observed ambient 24 levels of ozone and PM_{2.5} for all sites and data pairs; model results for both simulations with and 25 without the feedback effects are shown. Figure 10 presents similar comparisons but only for data pairs 26 where the simulated AOD>0.5. While somewhat arbitrary this criteria helps examine the model performance for cases of significant aerosol loading, and consequently wherein radiative feedback 27 28 effects on temperature and PBL heights could in turn influence the subsequent chemistry-transport 29 simulation. Though modest, the simulation including the aerosol feedback effects exhibits slight higher 30 correlation coefficients than the one without.

31

32 **5 Summary and Future Work**

1 A two-way coupled meteorological and air quality modeling system has been developed by linking the 2 WRF and CMAQ models. The system represents advancement over the traditional offline WRF-3 CMAO system because the aerosols predicted by CMAO are able to impact the clouds, radiation, and precipitation simulated by WRF in a consistent online coupled manner. In addition, because CMAQ is 4 5 called directly from WRF, the temporal interpolation of meteorological fields from WRF is eliminated 6 thereby improving consistency in the use of meteorological information in the chemistry-transport 7 calculations. A coupler is developed to efficiently link the two model systems. The coupler handles 8 communication between WRF and CMAQ, performs translation of the WRF fields to drive CMAQ, 9 and provides aerosols feedback information from CMAQ to the WRF. The coupler is encapsulated in Fortran 90 modules so the details of the two-way coupling are transparent to the users. This software 10 11 design also enables WRF and CMAO to be detached and executed as stand-alone models as in the 12 traditional offline paradigm. The single-source coding approach minimizes software maintenance so 13 scientific updates to both WRF and CMAQ can be readily incorporated into the coupled WRF-CMAQ 14 system.

15

16 In addition to scientific and software maintenance issues, the coupled modeling system was designed to 17 maximize user flexibility for research and applications by imposing minimal restrictions on domain specifications and physics options in both WRF and CMAQ. Furthermore, the coupler allows users to 18 19 choose the call frequency of CMAQ to balance the computational burden against the scientific 20 accuracy, depending on the availability of computational resources. The coupled modeling system also 21 includes a run-time switch to disable aerosol feedback and emulate the traditional offline paradigm 22 albeit with greater frequency of communication of meteorological information from WRF to the 23 CMAQ model; this option can be used for further sensitivity tests examining the potential effects of 24 temporal interpolation of meteorological data in the traditional offline paradigm.

25

When aerosol feedback is disabled, the computational time for of the coupled model is virtually identical to the offline WRF-CMAQ system. When the radiative feedback is enabled, there is slight increase in execution time (compared to the case without feedback) using the CAM radiation scheme. However, adding radiative feedback with the RRTMG scheme results in an increase in run time of about 60%, which is largely attributed to the more computationally intensive Mie calculation used in the implementation of the feedback effects with the RRTM scheme. Improving the computational efficiency of the more accurate Mie scheme and its coupling with the RRTM is currently being investigated. In general, the coupled WRF-CMAQ modeling system scales well as the number of
 processors increase, regardless of the radiation model chosen in WRF or whether or not simulation of
 feedback effects is enabled.

4

5 To demonstrate the improvements in simulated atmospheric dynamical and chemical features with the 6 inclusion of aerosol radiative effects, we conducted a ten-day simulation of a wildfire event in 7 California, a case characterized by significant tropospheric aerosol loading. Including radiative 8 feedbacks in the model noticeably reduced the bias in simulated surface shortwave radiation and 2-m 9 temperatures as well improved the correlation of simulated ambient ozone and PM_{2.5} relative to 10 observed values. This preliminary analysis suggests that for cases with high aerosol loading (such as 11 from wildfires, or in regions with significant anthropogenic pollution), including the radiative effects of 12 aerosols improves the accuracy of both the meteorology and air quality simulations.

13

14 Further model evaluation studies are continuing, including efforts to examine the direct aerosol effects 15 using a closed set of aerosol and radiation observations from the DOE/ARM Southern Great Plains site. 16 Ozone has absorption bands in the long wave radiation bands and can thus absorb outgoing radiation. 17 Efforts are underway to implement ozone feedback in the coupled WRF-CMAQ system and study the impact of ozone on long wave radiation using the RRTMG long wave radiation scheme. An initial 18 19 implementation of indirect aerosol forcing has also recently been completed and is under further testing 20 and evaluation. To simulate the most realistic representation of the dynamical state of the atmosphere, 21 four dimensional data assimilation (FDDA) is often employed in retrospective WRF applications. 22 Depending on the strength of the nudging coefficients used, FDDA could dampen the effects of 23 radiative feedbacks in the simulations and such effects need to be further quantified and understood. 24 The overall CPU usage of the coupled WRF-CMAQ system is equivalent to the total CPU time 25 associated with running the offline WRF-MCIP-CMAQ modeling system. A speedup of 2.3 when 26 going from 32 to 128 processors for the Eastern U.S. domain, however, suggests that there is room for 27 further improvement in the parallel performance of the WRF-CMAQ system. Part of this performance 28 hit is due to the input and output files (about ten each) as well as intermediate run time diagnostics. 29 Improvements to the I/O in the modeling system are currently under investigation and should result in 30 overall better performance scaling. Improvements in many of these aspects are currently being 31 investigated and will be made available in future versions of the 2-way coupled WRF-CMAQ 32 modeling system.

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	Execution time		
	CAM	RRTMG	
WRF only	0:19:59	0:18:50	
MCIP	0:02:31	0:02:31	
Offline CMAQ	1:18:28	1:19:05	
Offline WRF-CMAQ system, Total time	1:40:58	1:40:26	
Coupling system w/o feedback and call frequency ratio 5:1	1:41:12	1:48:59	
Coupling system w/ feedback and call frequency ratio 5:1	1:43:39	2:54:25	

1 Table 1: Performance of the uncoupled and coupled system (hh:mm:ss) on a linux cluster

Processor	САМ				RRTMG			
configuration	w/o feedback	speedup	w/ feedback	speedup	w/o feedback	Speedup	w/ feedback	speedup
4x8	2:05:06		2:08:21		2:13:17		3:19:25	
8x8	1:19:46	1.57	1:21:57	1.57	1:24:12	1.58	1:58:21	1.68
8x16	0:55:28	2.26	0:55:12	2.33	0:56:38	2.35	1:14:14	2.69

1 Table 2: Computational performance (hh:mm:ss) and scalability on a linux cluster



- 2 Figure 1. Schematic of high-level modules in (a) offline WRF-CMAQ system and (b) two-way coupled
- 3 WRF-CMAQ model.





Figure 2. WRF and CMAQ domain orientation in coupled system.



4 Figure 3. Decomposition with a 4x4 processor configuration for RSL-lite (a) and CMAQ (b).



W – WRF step F – apprep step B – feedback step C – CMAQ step Ra – radiation calculation

1

Figure 4. Calling sequence of the coupled system with 4:1 call frequency and radiation is updated every 6 timesteps



- 2 Figure 5. Locations of the June 2008 California wildfires (captured by NASA's Aqua Satellite,
- 3 http://www.nasa.gov/topics/earth/features/fire_and_smoke.html) illustrated by the red dots. Also
- 4 shown are the locations of observations sites whose data is shown in Figures 7 and 8.



Figure 6. Illustration of the direct radiative effects of aerosols simulated by the coupled WRF-CMAQ
model at 22 UTC on June 25, 2008.



2 Figure 7. Total surface short wave radiation comparison between measurements at Hanford, CA (blue),

3 and with (yellow) and without (red) direct aerosol feedback



2 Figure 8. Surface (2-m) temperature comparison of averaged measurements (blue) from four sites

- 3 (Oroville, Red Bluff, Redding and Sacramento) with model simulation with (yellow) and without (red)
- 4 direct aerosol feedback



3 Figure 9. Scatter plots of model, with (blue) and without (red) feedback, and observed daytime (a) O_3

4 and (b) PM_{25} ambient levels. Also shown are the slope and correlation coefficient (R)

1



3 Figure 10. Daytime O_3 and PM25 model performance with AOD greater than or equal to 0.5 (same

4 colour code as in Fig. 9)

1