A quasi chemistry-transport model mode for EMAC

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Abstract. A quasi chemistry-transport model mode (QCTM) is presented for the numerical chemistry-climate simulation system ECHAM/MESSy Atmospheric Chemistry (EMAC). It allows for a quantification of chemical signals through suppression of any feedback between chemistry and dynamics. Noise would otherwise interfere too strongly. The signal is calculated from the difference of two QCTM simulations, a reference simulation and a sensitivity simulation. These use offline chemical fields as a substitute of the feedbacks between chemistry and dynamics: (a) offline mixing ratios of radiatively active substances enter the radiation scheme, (b) offline mixing ratios of nitric acid enter the scheme for re-partitioning and sedimentation from polar stratospheric clouds, (c) and offline methane oxidation is the exclusive source of chemical water-vapor tendencies. Any set of offline fields suffices to suppress the feedbacks, though may be inconsistent with the simulation setup. An adequate set of offline climatologies can be produced from a non-QCTM simulation using the setup of the reference simulation. Test simulations reveal the particular importance of adequate offline fields associated with (a). Inconsistencies from (b) are negligible when using adequate fields of nitric acid. Acceptably small inconsistencies come from (c), but should vanish for an adequate prescription of chemical water vapor tendencies. Toggling between QCTM and non-QCTM is done via namelist switches and does not require a source code re-compilation.

1 Introduction

The recognition of feedback processes in the Earth system is essential to the understanding of past and future changes in the atmospheric chemical composition and climate. Global chemistry-climate models (CCMs) are well suited to investigate such feedbacks, since these models account for the non-linear coupling between atmospheric dynamics, chemistry, radiation, and other processes. However, the presence of feedbacks can hamper the identification of climate effects from, for instance, altered emissions of short-lived anthropogenic substances. This is the case if natural variability, here referred to as noise, disguises the expected signal.

For example, the objective of quantifying the expected impact from changes in nitrogen oxides ($\text{NO}_x=\text{NO}+\text{NO}_2$) emissions requires a comparison of at least two model simulations: a reference simulation with reference sources of $\text{NO}_x$ and a sensitivity simulation with sources that differ from the reference setup slightly. If the simulations include any feedback between dynamics and chemistry, then the meteorology of the two simulations will diverge despite binary identical initialization due to the inherent chaos in the system, a property commonly referred to as butterfly effect. Such behavior is inevitable, even for the smallest chemically triggered perturbation of the meteorological state. After a small number of time steps with perturbed conditions, the two simulations will differ completely in their meteorological weather patterns, communicating the meteorological variability to mixing ratios of chemical compounds such as $\text{NO}_x$ (e.g. Unger et al., 2008).

Note that time averages of dynamical variables may be hardly affected, but that the sequence of meteorological weather patterns can be completely different among the two simulations. Hence, day-to-day mixing ratio differences in the two simulations will be governed by noise (Fig. 1a), and estimation of statistical expectation values of chemical quantities require long and computationally expensive integrations. The problem is particularly strong when focusing on the quantification of effects from small perturbations of multiple emission sources such as road traffic, shipping, and aviation (e.g. Hoor et al., 2009; Unger et al., 2008).

Feedback processes between atmospheric chemistry and
Fig. 1. Mixing ratio \( x \) of an arbitrary chemical tracer from two hypothetical simulations (sketch a). The simulation setups differ only by a small chemical perturbation affecting \( x \), and share identical initialization. Note that the simulations have near-identical statistical expectation values with respect to dynamical quantities such as divergence and vorticity. However, the feedback between chemistry and dynamics causes a completely different sequence of dynamical patterns and associated tracer transport, and hence a low signal-to-noise ratio for the chemistry-related difference \( \Delta x \) among the two simulations. Without feedback between chemistry and dynamics, however, \( \Delta x \) relates directly to the chemical perturbation, and it displays a high signal-to-noise ratio (sketch b).

Dynamics are manifold. Every chemically and radiatively active compound causes such feedback. Typical examples are ozone \((O_3)\), methane \((CH_4)\), chlorofluorocarbons (CFCs), and nitrous oxide \((N_2O)\). A special species is water vapor \((H_2O)\) as it plays an important role not only in the radiation budget, but also in cloud physics and atmospheric chemistry; the latter, for example, as a source of hydroxyl-radical \((OH)\) formation or as a product of stratospheric \(CH_4\) degradation.

Also, there are substances which are not radiatively active, but which affect dynamical processes such as properties of clouds. The formation of polar stratospheric clouds (PSCs), for example, depends on species such as \(H_2O\) and nitric acid \((HNO_3)\). Any minor change in \(HNO_3\) mixing ratios will alter PSC formation, and, through the associated \(H_2O\) changes, causes a feedback between chemistry and dynamics.

Yet, chemically-triggered changes to the feedback processes themselves are often small enough to be neglected. It is hence feasible to account for the feedbacks between chemistry and dynamics with prescribed climatologies. With this restriction, chemical perturbations will not affect the meteorological patterns, turning chemistry-dynamics feedbacks off. In either of the two example simulations mentioned above, the patterns of atmospheric dynamics are then binary identical and any difference in chemical mixing ratios is directly attributable to the difference in \(NO_x\) emission rates among the two (Fig. 1b). As a result, the differences in chemical mixing ratios have a much higher signal-to-noise ratio.

We say a CCM is run in quasi chemistry-transport model mode (QCTM) when any impact from the model chemistry on the dynamics is switched off. The opposite case with active feedbacks between chemistry and dynamics is referred to as coupled mode, or coupled simulation. The term “quasi” serves to distinguish from pure chemistry-transport models which do not integrate the prognostic dynamical equations. The model setup determines also other model characteristics such as resolution, boundary conditions, and the selection of active submodels. Offline chemical characteristics are imposed on a simulation from outside, whereas online chemical characteristics are part of the simulation’s chemistry integration.

This documentation presents a QCTM implemented into the CCM ECHAM/MESSy Atmospheric Chemistry (EMAC), version 1.10. It was released in version 1.11. Section 2 introduces the model system EMAC, and Section 3 gives specific information on those submodels of EMAC which are important to the QCTM. Section 4 evaluates the QCTM. Following the conclusions of Section 5, the Appendix informs on how to use the QCTM.

2 Model system EMAC

EMAC is a numerical chemistry and climate simulation system that includes submodels describing tropospheric and middle atmosphere processes and their interaction with
oceans, land, and human influences (Jöckel et al., 2006). It uses the first version of the Modular Earth submodel System (MESSy1) to link multi-institutional computer codes. The core atmospheric model is the 5th generation European Centre Hamburg general circulation model ECHAM5 (Roeckner et al., 2006). More detailed information, including references, about the model system is available from http://www.messy-interface.org.

The core model of EMAC, ECHAM5, solves the atmospheric primitive equations horizontally via a spectral transform technique. In the vertical it uses the method of finite differences and for time integration a semi-implicit leap-frog scheme with time filter.

Most of the EMAC submodels act on a Gaussian transform grid and, e.g., handle parameterized physics, transport of atmospheric constituents, and atmospheric chemistry. Depending on the complexity of the chemistry setup, the chemical calculations usually consume a large portion of the total CPU time. Chemistry is dealt with by the submodel MECCA1 (Sander et al., 2005), which is based on the kinetic pre-processor (KPP) software (Damian et al., 2002) and considers the combined chemical reactions as a stiff system of ordinary differential equations. The system is solved via a Rosenbrock sparse-matrix technique with adaptive time step.

The parallelization is implemented based on the distributed memory concept with different approaches for the spectral and the grid-point representation, and a specific representation for the tracer advection algorithm, respectively. In spectral representation, variables are decomposed into their spherical harmonics, and in Gaussian representation the global domain is decomposed into blocks in latitudinal and longitudinal direction. For the advection an additional vertical decomposition is applied. For each variable in Gaussian representation two blocks at the same longitude and the same latitude are combined, with one block being in the Northern and one block in the Southern hemisphere. This improves the load balancing with respect to the distribution of day and night grid-boxes. The combined blocks are further re-arranged into vectors of selectable length to optimally exploit vector registers on vector architectures and cache sizes on scalar architectures.

3 Specific submodels and QCTM implementation

The implementation of the QCTM involves the EMAC submodels RAD4ALL, H2O, HETCHEM, and PSC, which are commented on here. Switching between QCTM and coupled mode does not require a re-compilation of the EMAC source code, so that the same executable can be used for both modes. Instead, it requires the modification of certain switches and adjustments in the respective namelists (see Table A1).

3.1 Submodel RAD4ALL

The submodel RAD4ALL (see Jöckel et al., 2006) is a re-implementation into the MESSY standard of the ECHAM5 radiation code (Roeckner et al., 2006). In coupled mode, RAD4ALL relies on online fields for the calculation of radiative transfer. These are with regard to the hydrological cycle: cloud cover, cloud water, cloud ice, prognostic specific humidity; and with regard to the radiatively active chemical species: CO₂, CH₄, O₃, N₂O, CFC-11, CFC-12, depending on the chemical setup.

For the QCTM, the mixing ratios of all radiatively active chemical species, except for the H₂O tracer, must be provided from offline climatologies. The relevant switches are located in the CPL namelist of the RAD4ALL submodel, and the respective offline fields are defined via the REGRID and RTGEVENTS namelists of RAD4ALL (see Table A1). Note the model chemistry still provides online mixing ratios of the radiatively active species, but these are not used by the RAD4ALL submodel if the QCTM is active.

3.2 Submodel H2O

The submodel H2O serves two purposes. First, it defines the gas-phase tracer H₂O as part of the model chemistry, in contrast to the specific humidity being inherent to the ECHAM5 core model. To clearly separate these two aspects, we will refer to the water vapor fields from the H2O submodel as H₂O tracer, and to the water vapor field from ECHAM5 as specific humidity. Optionally, the H2O submodel initializes the H₂O tracer in the stratosphere and mesosphere from satellite data. These domains are notoriously too dry in the base-model ECHAM5. Second, the H2O submodel controls the feedback between the H₂O tracer and specific humidity, and the contribution by CH₄ degradation. There are five different options, and these are controlled by the CPL namelist of the H2O submodel:

- \( 1, \text{H}_2\text{O}_\text{TENDENCY} = -2 \): The H₂O tracer is defined and initialized with the specific humidity, without correction from satellite data in the stratosphere and the mesosphere. No CH₄ oxidation from chemical submodels is taken into account. The H₂O tracer and the specific humidity are completely detached.

- \( 1, \text{H}_2\text{O}_\text{TENDENCY} = -1 \): Same as above, but the H₂O tracer is initialized in the stratosphere and the mesosphere from satellite data.

- \( 1, \text{H}_2\text{O}_\text{TENDENCY} = 0 \): The H₂O tracer is defined and initialized in the stratosphere and the mesosphere from satellite data. In addition, the tendencies of the H₂O tracer from all other submodels are synchronized with those of the specific humidity. This option considers the impact from water vapor on the dynamics by online CH₄ oxidation calculated by a full-chemistry submodel.
- **I\_H2O\_TENDENCY = 1**: The H\_2\_O tracer is defined and initialized in the stratosphere and the mesosphere from satellite data. In addition, a source of H\_2\_O from climatological offline CH\_4 oxidation is added to the overall tendency. The H\_2\_O tracer and the specific humidity are synchronized to consider the impact on the dynamics by the climatological CH\_4 oxidation. This option is implemented for model setups without CH\_4 oxidation calculated by a chemistry submodel.

- **I\_H2O\_TENDENCY = 2**: Same as for option I\_H2O\_TENDENCY = 1, but only the H\_2\_O-tracer tendency from the climatological offline CH\_4 oxidation is used to modify the specific humidity during the synchronization.

For the QCTM, the switch I\_H2O\_TENDENCY must be set to 2 as this eliminates the feedback on the dynamics by online chemical tendencies of water vapor. The offline water vapor tendency in options 1 and 2 is calculated based on a satellite climatology of methane mixing ratios and pre-calculated climatological conversion rates. The latter are taken from a coupled simulation. The default involves monthly average conversion rates from a MAECHAM\_4/CH\_4 simulation (Steil et al., 2003). These correspond to photolysis and oxidation of CH\_4 by OH, O1D, and Cl. Alternative input data may be selected through the submodel namelists.

### 3.3 Submodel HETCHEM

The submodel HETCHEM computes heterogeneous reaction rates for surfaces associated with solid nitric acid trihydrate (NAT), ice, super-cooled ternary solutions, and aerosols. As to the present study, a climatology of H\_2\_SO\_4 mixing ratios and pre-calculated climatological conversion rates. The latter are taken from a coupled simulation. The default involves monthly average conversion rates from a MAECHAM\_4/CH\_4 simulation (Steil et al., 2003). These correspond to photolysis and oxidation of CH\_4 by OH, O1D, and Cl. Alternative input data may be selected through the submodel namelists.

### 3.4 Submodel PSC

PSCs affect the stratospheric chemistry and chemical composition by heterogeneous reactions occurring on both liquid and solid cloud particles, and by sedimentation of the solid cloud particles. The submodel PSC takes these effects into account within the EMAC framework. It further determines heterogeneous reaction rates. Also, it establishes the vertical redistribution of H\_2\_O and HNO\_3 due to sedimentation of ice, referred to as dehydration, and due to sedimentation of NAT, referred to as denitrification.

The removal of PSC-related feedback is not as intuitive as for the submodels H\_2\_O, HETCHEM, and RAD4ALL, so that the PSC submodel deserves a somewhat more extensive explanation.

In every time-step, the PSC submodel is initialized with integration results from the previous timestep. Among these are temperature, local pressure, amounts of H\_2\_O in its three thermodynamic phases, the combined amounts of gaseous and liquid HNO\_3, and the amount of HNO\_3 as NAT.

Then, the PSC submodel determines which thermodynamical phases are stable under the current local conditions of temperature, pressure, total-H\_2\_O content, and amounts of gaseous and liquid as well as NAT. There are three thermodynamic phases: a) binary or ternary liquid particles only, b) mixture of solid-NAT and liquid particles, c) mixture of solid-NAT, ice, and liquid particles. Sulphuric acid tetrahydrate is not accounted for.

The thermodynamic partitioning of H\_2\_O and HNO\_3 is accomplished in two steps. First, the PSC submodel separates the combined gaseous and liquid phases from the solid phase, generating ice in case of c), and solid NAT in case of b) and c). It also determines the size of the solid ice and NAT particles and the user can choose among two different schemes. The simpler one assumes that ice and NAT particles have all the same size (Buchholz, 2005); the more sophisticated one considers a discrete size distribution for the NAT particles (Kirner et al., 2010). As a next step, the submodel solves for the amount of super-cooled ternary solution and calculates a composition-dependent size distribution of these liquid particles.

Thereafter, the PSC submodel determines the sedimentation velocities of the solid H\_2\_O and NAT particles based on temperature, local pressure, and particle size. Liquid particles are regarded as too small for sedimentation. The product of sedimentation velocity and length of the integration time step gives the vertical re-distribution of the solid particles. This, combined with the particle number density, defines the new amounts per model box of ice and solid NAT due to sedimentation. If the particles fall out of the height region where PSCs can occur, then the particle mass is transferred to the corresponding gas phase species.

The entire numerical scheme presented above is restricted to those atmospheric regions where PSCs are allowed to exist. The user prescribes an equatorward latitude limit as well as upper and lower pressure bounds to avoid PSCs where observations do not support their existence. Inside these pre-defined regions, the PSC submodel performs any PSC-related calculation only where the current local equilibrium temperature of ice is smaller than that of NAT.

The PSC submodel causes feedback between dynamics and chemistry through the following processes:

1. It defines heterogeneous chemical reactions via the HETCHEM submodel. These, in turn, alter the distribution of radiatively active substances through the chemistry submodel MECCA1. The RAD4ALL radiation submodel then communicates the chemical perturbation
to the dynamics. In case of water vapor, the hydrological cycle represents an additional communication path.

2. Sedimentation-caused mixing ratio changes of $H_2O$ depend on $HNO_3$ through ice particle size and partitioning. Again, hydrological cycle and radiation close the feedback towards the model dynamics.

3. The amount of $HNO_3$ determines which thermodynamic phases are stable and thus impacts on the PSC-relevant regions (see above). For these regions, the CLOUD submodel suppresses the occurrence of any cirrus clouds, hence generating feedback. Additionally, the PSC-relevant regions re-enter the PSC submodel in the next time-step.

A removal of the PSC-related feedbacks by avoiding PSCs at all would be trivial to do. However, PSCs are important to the Earth system, e.g. through their impact on stratospheric ozone mixing ratios and dynamics, and should be accounted for. Rather than having no PSCs at all, it is hence better to remove only the PSC-related feedbacks.

The QCTM settings in the submodels HETCHEM and H2O (see Sections 3.2, 3.3 and Table A1) remove any feedback associated with item 1 in the above list. Items 2 and 3 represent feedback through non-chemical mechanisms which involve $HNO_3$. Figure 2 illustrates how the feedback removal is implemented in this case. The main idea is to replace online total-$HNO_3$ mixing ratios by an offline climatology where necessary in the code, while affecting PSC-related effects such as sedimentation as little as possible.

Partitioning within PSCs is therefore evaluated twice:

1. using an offline total-$HNO_3$ climatology in order to make the following quantities and characteristics independent of the KPP-solved chemistry: water vapor tendencies, particle radii, and thermodynamic phase. The water vapor tendencies and particle radii go into the sedimentation scheme. In a similar way, the thermodynamic phase goes into the cirrus-related computations by the CLOUD submodel.

2. using online total-$HNO_3$ fields from the KPP chemistry solver. As an important consequence of the double calculation, PSC-related partitioning/sedimentation induces KPP-dependent $HNO_3$ tendencies, which remain in the chemistry domain and do not affect the dynamics.

Table A1 informs about the namelists which make the partitioning QCTM conform.

4 Evaluation of EMAC in QCTM

This Section presents and discusses EMAC simulations in QCTM and coupled mode. Section 4.1 demonstrates that the QCTM has been successfully implemented into EMAC. It gives an example where sensitivity simulations and reference simulations in coupled mode fail to provide meaningful results due to a low signal-to-noise ratio. Based on this exemplary situation, the QCTM is shown to perform and to tackle the underlying scientific problem.

Section 4.2 evaluates the QCTM-related effects on dynamical and chemical characteristics of the atmosphere. This is accomplished through a comparison of simulation pairs that differ only by the settings required to switch between QCTM and coupled mode.

4.1 Demonstration of performance

Figure 3 shows the relative difference in $O_3$ mixing ratios between a sensitivity simulation and its reference simulation in QCTM (a and c) and fully coupled mode (b and d). The sensitivity simulation differs from the reference simulation by 5% stronger emissions from biomass burning of CO, non-methane volatile organic compounds (NMVOCs), and $NO_x$. The setup for the simulations is T42L90MA with EVAL chemistry (Jöckel et al., 2006), and includes the submodels CLOUD, CONVECT, CVTRANS, DRYDEP, H2O, HETCHEM, JV AL, LNOX, MECCA1, OFFLEM, ONLEM, PSC, QBO, RAD4ALL, SCAV, TNUDGE, and TROPOP. There is no nudging of the dynamical variables. The model setup is for conditions of the year 2000 and we analyze the seventh month, July, after initialization. This implies non-equilibrium conditions, but is sufficient for demonstration purposes.

In the presence of $NO_x$ and sunlight, oxidation of CO and NMVOCs produces $O_3$ in a complicated nonlinear way (Unger et al., 2008). The chemical scheme which has been evaluated by Jöckel et al. (2006) and which we adopted in our simulations accounts for the nonlinear ozone production. The nonlinearity implies considerable complexity of the impact on tropospheric chemical composition by altered $NO_x$, CO, and NMVOC emissions; Yet, the perturbation in emissions between sensitivity and reference simulation is small (5%), and applied to CO, NMVOC, and $NO_x$ simultaneously. It is hence reasonable to assume throughout the following discussion that their tropospheric mixing ratios, together with those of $O_3$, in general should be higher for the sensitivity simulation with its stronger emissions.

In the QCTM simulation, there is a clear positive anomaly in the troposphere of $O_3$ (Figures 3a and c), $NO_x$ (Figure 4), CO, and the NMVOCs (not shown). The anomaly is solely due to the chemistry as the dynamical states of sensitivity and reference simulation are binary identical (not shown). It is emphasized that the amplitude of the mixing ratio anomaly is similar for snapshot (Figure 3a) and monthly average (Figure 3c). The reason is that the sensitivity simulation, compared to the reference simulation, has biomass burning emission rates which are enhanced by a constant percentage (5%) and which include no day-to-day variability. However, the anomaly pattern is spatially less variable for the monthly mean, due to the
Fig. 2. Flowchart of how the QCTM is implemented into the EMAC submodel PSC. The partitioning is calculated twice using a) offline, and b) online mixing ratios of HNO$_3$ (see text for more detailed explanations).

smoothing of the day-to-day tracer transport variability. Note in this context that the spatio-temporal transport variability is binary identical for sensitivity and reference simulation in QCTM, in contrast to that in coupled mode. Finally, the O$_3$ response is consistent with the general situation. Enhanced African biomass burning causes higher near-surface mixing ratios of NO$_x$ (Figure 4), CO, and NMVOC (not shown) just north of the Equator. The enhanced mixing ratios are transported upwards by convection, and southward by the Hadley circulation, increasing the O$_3$ mixing ratios on their way.

The coupled mode, in contrast, does not give a consistent result (Figures 3b and d). Instead of a widespread positive pattern, there are noisy patches of both positive and negative anomalies, which also applies to NO$_x$, CO, and the NMVOCs (not shown). Such behavior indicates transport-related interference with the chemical signal as the dynamical states have diverged between the two simulations despite binary identical initialization, as sketched in Figure 1a. This is due to the feedback between chemistry and dynamics (see Section 1). Note that the anomalies are significantly weaker for the monthly averaging (Figure 3d) than for the instantaneous snapshot (Figure 3b). Such behavior is likely to result from the averaging which acts as a low-pass filter on the spectrum of atmospheric dynamics, and hence on tracer mixing ratios. Applying statistical analysis to quantify, or even identify, a chemical signal would be a fruitless endeavor due to the overwhelming noise.

4.2 Impact on simulation climatologies

The removal of chemistry-climate feedbacks from EMAC represents an interference with the model chemistry and dynamics and it is important to evaluate the associated side effects. In this context, we compare simulations in coupled mode against a simulation in QCTM. We assume that such a comparison also serves to check whether the side effects cause significant inconsistencies for the QCTM simulation itself.

The setup for all simulations is T42L90MA with EVAL chemistry (Jöckel et al., 2006), and includes the submodels CLOUD, CONVECT, CVTRANS, DRYDEP, H2O,
Fig. 3. Latitude-height section of the relative difference in O₃ mixing ratios between sensitivity and reference simulation in QCTM (left) and fully coupled mode (right). The sensitivity simulation has 5% stronger emissions from biomass burning of NOₓ, CO, and NMVOC, affecting O₃ mixing ratios chemically (see text). Results are shown for the Greenwich Meridian for a single month under July conditions: 00:00 UTC on July 1st (top), monthly mean (bottom). Isoline intervals are 0.3% in QCTM (left), but 10% (top right) and 0.5% (bottom right) in coupled mode. Positive values (red) denote higher O₃ mixing ratios for the sensitivity simulation.

Fig. 4. Same as Figure 3c, but for NOₓ. Isoline intervals are 0.6%.

HETCHEM, JVAL, LNOX, MECCA1, OFFLEM, ONLEM, PSC, QBO, RAD4ALL, SCAV, TIMEPOS, TNUDGE, and TROPOP. Each experiment covers five years with nudging of tropospheric dynamics towards ECMWF re-analysis fields.
ratios of the radiatively active species, except for CO in coupled mode, the radiation scheme uses online mixing. The nudging reduces the noise in the simulation inter-comparisons and makes these better fulfill the normality assumption inherent to the paired t-test (Fomichev et al., 2007). Mixing ratios of total HNO₃ from coup_full are averaged on a monthly basis before fed into decoup_full and coup_rad.

<table>
<thead>
<tr>
<th>experiment</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>coup_full</td>
<td>fully coupled mode</td>
</tr>
<tr>
<td>coup_PSC</td>
<td>coupling only due to PSCs other settings as for decoup_full</td>
</tr>
<tr>
<td>coup_rad</td>
<td>coupling only due to radiation other settings as for decoup_full</td>
</tr>
</tbody>
</table>

For the period 1999 to 2003. Figure 5a reveals significant discrepancies in temperature between two simulations, one in fully coupled mode and one in QCTM (see Table 1). The largest discrepancies occur above the 1 hPa level, culminating near the uppermost model layer in a +30 K temperature difference, a +50% difference in relative humidity, and a -60% difference in ozone mixing ratios. A further distinct feature is a widespread pattern of opposite sign showing negative temperature differences in the height region between 1 hPa and 100 hPa.

Two additional test simulations are performed in order to unravel the mechanisms behind these discrepancies. These simulations are identical to the QCTM simulation mentioned above except that in each case a single coupling pathway remains switched on. One simulation includes the coupling due to radiative effects, the other the coupling associated with PSCs (see Table 1).

Obviously, the removal of radiative-feedback effects explains the lion’s share of the temperature discrepancies above the 100 hPa level (see Figure 5b); for the following reason: in coupled mode, the radiation scheme uses online mixing ratios of the radiatively active species, except for CO₂ and CFC-11, whereas it relies on offline mixing ratios in QCTM. It is by accident that the mixing ratios of CFC-11 for the coupled setup are offline. Note that the offline mixing ratios of CO₂ and CFC-11 are identical for both setups. For the particular case presented here, the QCTM setup adopts horizontally uniform, height-decreasing, mixing ratios of CFC-12, CFC-11, N₂O, and CH₄. The mixing ratios of CO₂ are uniform, and those of O₃ are taken from Fortuin and Kelder (1998). The O₃ climatology by Fortuin and Kelder (1998) extends up to the 0.3 hPa pressure level, well below the highest model layer which is centered at the 0.01 hPa pressure level. For the model layers above the 0.3 hPa pressure level we use height-constant O₃ mixing ratios from the uppermost layer of the climatology, about 1.6×10⁻⁶ mol/mol. Finally, the radiation scheme employs the prognostic specific humidity which, in coupled mode, depends on the KPP chemistry solver via the H₂O tracer, but not in QCTM (see Section 3.2). The relative differences for N₂O, CFC-12, and CH₄, offline versus online, share the same basic features in the stratosphere as well as the mesosphere. To give an example, Figure 6a depicts the N₂O difference. The offline case displays positive deviations at all latitudes above the 1–0.1 hPa height region, the exact height being species dependent. The disparity increases strongly towards the uppermost model layer where, at the same time, the mixing ratios are small. The absolute deviations are hence weak, of the order of 10⁻¹⁴ mol/mol for CFC-12, 10⁻¹⁵ mol/mol for N₂O, and 10⁻¹⁶ mol/mol for CH₄. Below the 1–0.1 hPa height region, there are positive deviations poleward of the sub-tropical to middle latitudes, but negative deviations more equatorward. Again, the location of the transition between positive and negative deviations is species dependent. The pattern below the 1–0.1 hPa height region is most likely associated with upwelling in the tropical pipe which transports air from the tropical troposphere upwards.

At first glance, the deviation pattern of temperature above the 100 hPa level (Figure 5a) appears to match its counterparts in case of N₂O (Figure 6a), CFC-12, and CH₄. If the deviation patterns of these species were to explain the temperature response, however, a positive mixing ratio deviation would have to increase the long-wave radiative heating. For realistic mixing ratio profiles it is well known that the opposite is true above the 100 hPa level in case of N₂O and CH₄, and above the 30 hPa level in case of CFC-12 (e.g. Clough and Iacono, 1995).

Figure 6b shows the relative difference in case of O₃. As for the three species mentioned above, there is a positive deviation which increases towards the uppermost model layer, and it attains values of well beyond +600%. The deviation is also considerable in absolute numbers. It reaches +1×10⁻⁶ mol/mol, compared to −0.5×10⁻⁶ mol/mol at the 30 hPa level in the tropics where the largest mixing ratios are.

Enhanced ozone mixing ratios cause a stronger local short-wave heating (e.g. WMO, 2007). This might well explain the distinct positive temperature deviation in the mesosphere,
Fig. 5. Zonally averaged annual-mean temperature deviation (isolines [K]) for the five simulation years 1999-2003. Shading refers to the 99% (dark-gray) and 95% (light-gray) significance levels from a paired t-test with four degrees of freedom. a): simulation *decoup full* minus *coup full*. b): *coup rad* minus *coup full*. Table 1 presents the simulation setups.

Fig. 6. Same as Figure 5, but only for the relative deviation of simulation *decoup full* versus *coup full* (see Table 1). a): ∆N2O [%], b): ∆O3 [%], and c): ∆H2O [%] (gaseous only). In case of *decoup full* there are two different sets of the radiatively active species O3, N2O, CFC-12, and CH4. The radiation submodel RAD4ALL relies on their respective offline mixing ratio climatologies, whereas chemistry and transport impact the online mixing ratios.

the region above the 1 hPa level (Figure 5a), and outweigh the intensification in long-wave cooling from N2O, CFC-12, and CH4. Conversely, the positive deviation in mesospheric O3 mixing ratios could also explain the widespread negative
deviation in stratospheric temperatures, in the height region between 100 and 1 hPa. The mechanism responsible is that stronger absorption in the mesosphere means less short-wave radiation is available to heat the stratosphere below.

To circumvent the aforementioned discrepancies the radiation scheme of the QCTM simulation has to be driven with appropriate offline climatologies of \( \text{O}_3 \), \( \text{N}_2\text{O} \), CFC-12, and \( \text{CH}_4 \). These should be obtained from the same coupled reference simulation that was used to derive the offline climatology of \( \text{HNO}_3 \) (see Table 1). However, this is beyond the scope of this documentation.

Figure 5b demonstrates the temperature effect without the radiation-induced contribution by non-\( \text{H}_2\text{O} \) species such as \( \text{O}_3 \), \( \text{N}_2\text{O} \), CFC-12, and \( \text{CH}_4 \). Hence, the remaining effect is associated with the removal of \( \text{H}_2\text{O} \)-related feedback mechanisms (Section 3.4): the substitution of chemical \( \text{H}_2\text{O} \) tendencies by climatological \( \text{CH}_4 \) degradation, the omission of heterogeneous-chemical \( \text{H}_2\text{O} \) tendencies, and the PSC-related sedimentation of \( \text{H}_2\text{O} \) made to “see” climatological \( \text{HNO}_3 \) offline mixing ratios.

The are two obvious structures of temperature deviation. The tropical/subtropical mesosphere displays a deviation of \(-0.6\text{K}\), and the entire troposphere a deviation of about \(-1\text{K}\). Both structures appear to result from deviations of \( \text{H}_2\text{O} \) and \( \text{O}_3 \) (Figures 7a and 7b): in the mesosphere, the positive \( \text{H}_2\text{O} \) deviation increases long-wave radiative cooling, and the negative \( \text{O}_3 \) deviation decreases short-wave radiative heating. The \( \text{H}_2\text{O} \) deviation by itself results mainly from the substitution of the online chemical \( \text{H}_2\text{O} \) tendencies by climatological \( \text{CH}_4 \) degradation, and the \( \text{O}_3 \) deviation occurs as an effect to this. In the troposphere, a negative \( \text{H}_2\text{O} \) deviation is known to reduce the long-wave heating, producing the negative temperature deviation in Figure 5b. The tropospheric \( \text{H}_2\text{O} \) deviation by itself is likely to arise from the omission of heterogeneous-chemical \( \text{H}_2\text{O} \) tendencies as \( \text{CH}_4 \) degradation plays a minor role in the water vapor budget there.

Finally, the removal of non-\( \text{H}_2\text{O} \) radiative feedback artificially compensates for the tropospheric deviation in temperature (Figure 5a) and \( \text{H}_2\text{O} \) (not shown).

Again, a remedy for these \( \text{H}_2\text{O} \)-related discrepancies would be to feed the QCTM simulation with appropriate climatological offline tendencies of \( \text{H}_2\text{O} \). These should originate from the coupled reference simulation (see Table 1) and describe the tendencies of \( \text{H}_2\text{O} \) from \( \text{CH}_4 \) oxidation.

Figure 8a depicts the temperature deviation due to removal of feedback from PSC-related \( \text{H}_2\text{O} \) sedimentation. The pattern of statistical significance is typical for type I random errors from multiple testing in the presence of spatial autocorrelation (von Storch and Zwiers, 1999). Hence, the deviations are not considered as robust, particularly for the southern polar lower stratosphere where PSCs most frequently occur. The same holds for mixing ratios of total \( \text{HNO}_3 \) (Figure 8b) and radiatively active substances such as \( \text{H}_2\text{O} \) and \( \text{O}_3 \) (not shown). The lack of deviation in temperature and tracer mixing ratios demonstrates that the QCTM approach is reliable.

5 Conclusions

A quasi chemistry-transport model mode, QCTM, has been implemented into the EMAC chemistry-climate model system. The QCTM is demonstrated to eliminate any feedback between chemistry and dynamics, which is a pre-requisite for meaningful sensitivity studies on the impact from small chemical perturbations. Such studies have been hampered by the feedback-triggered noise from the chaotic climate system.

Specifically, the elimination of radiation-induced feedbacks is achieved by feeding offline mixing ratios of radiatively active substances to the radiation scheme. An equivalent approach is not applicable in case of \( \text{H}_2\text{O} \) as it feeds
Fig. 8. Same as Figure 5, but for the deviation of simulation coup_psc versus decoup_full (see Table 1). a): $\Delta T$ [K], b): $\Delta$HNO3 [%] (sum of gaseous, liquid, solid).

Back on the model dynamics not only via the radiative transfer, but also via the hydrological cycle and other, PSC related, processes. In this case, the tendencies of H$_2$O are disconnected from the KPP chemistry solver.

1. H$_2$O tendencies from homogeneous chemistry are limited to climatological CH$_4$ oxidation.
2. H$_2$O tendencies due to thermodynamic re-partitioning are neglected.
3. H$_2$O tendencies from PSC-related sedimentation are calculated using prescribed mixing ratios of HNO$_3$.

Activation of the QCTM is accomplished by adjustment of certain namelist switches (see Table A1).

It is also required to feed the radiation scheme with appropriate offline fields of the radiatively active species, except H$_2$O, and the PSC scheme with appropriate offline fields of total HNO$_3$. These fields should be compatible with the general chemical and dynamical characteristics of the QCTM simulation. Specifically, it is highly recommended to perform a coupled simulation with similar setup, derive mixing ratio climatologies of the relevant substances, and feed these into the associated QCTM simulation.

The two simulations in different mode, yet similar setup, are designed to yield consistent results. Hence, a comparison of the two simulations serves to evaluate the QCTM:

1. The offline fields of radiatively active species dominate the QCTM climatology. Associated inconsistencies can be large, especially in the mesosphere and the stratosphere, so that a prescription of adequate offline fields is important.
2. The removal of PSC-related feedback is consistent.
3. Acceptably small inconsistencies are associated with the disconnection of the H$_2$O tendencies from online chemistry.

Elimination of the inconsistencies under item three would require a prescription of the offline H$_2$O tendencies from the coupled reference simulation. These should climatologically represent the H$_2$O tendencies from homogenous chemistry and thermodynamic re-partitioning.

Appendix A

Use of EMAC in QCTM

This section exemplifies the use of EMAC in quasi chemistry-transport model mode (QCTM), and it is helpful to recapitulate some definitions. EMAC is run in QCTM when any impact from the model chemistry on the dynamics is switched off. The opposite case with active feedbacks between chemistry and dynamics is referred to as coupled mode. The model setup determines other model characteristics such as resolution, boundary conditions, and selection of active submodels. An offline field of chemical characteristics such as resolution, boundary conditions, and selection of active submodels. An offline field of chemical characteristics is imposed on a simulation from outside, whereas an online field is part of the simulation’s chemistry integration. A reference simulation has a reference chemical setup, from which the chemical setup of a sensitivity simulation differs slightly.

In case of one reference simulation and a single sensitivity simulation it is usually required to perform three different simulations. All simulations should use the same executable. The reference simulation is implemented as two different simulations: once in coupled mode, once in QCTM. The coupled simulation serves to generate climatologies for input into the QCTM reference and sensitivity simulations. Hence, the coupled reference setup and the QCTM reference setup should be as similar as possible, and should usually differ only by the settings required to switch between coupled mode and QCTM (see Table A1). It is up to the user
to decide on the length of the coupled simulation and on the
details of the climatologies such as base period and temporal
resolution.

The coupled simulation is accomplished first. Straightfor-
ward summation of the standard output variables HNO$_3$ and
HNO$_3$nat serves to establish a climatology of total-HNO$_3$
mixing ratios. It goes into the QCTM simulations as input
variable predef.HNO3$_{tot}$ (see Figure 2) and removes the
PSC-related feedbacks listed in Section 3.4 under items 2
and 3.

Also from the coupled simulation, climatologies of those
radiatively-active species should be made which enter the
radiative transfer as online fields (see rad4all.nml). This is
usually the case for O$_3$, N$_2$O, CFC-11, CFC-12, and CH$_4$.
Again, the climatologies are fed into the two QCTM simu-
lations as explained in Table A1. It is emphasized that the mea-
sures recommended in this paragraph do not apply to water
vapor, which enters the radiative transfer as an online field in
both QCTM and coupled mode.

The QCTM works as well for any arbitrary prescription
of O$_3$, N$_2$O, CFC-11, CFC-12, CH$_4$, and predef.HNO3$_{tot}$.
Yet, the offline fields made from an appropriate coupled sim-
ulation bring the climatological characteristics of coupled
mode and QCTM close together. Even more important, it
ensures consistency between the offline fields and the clima-
tological characteristics of the QCTM simulation.

Finally, the desired response to small chemical perturba-
tions follows from the difference between the two QCTM
simulations, reference and sensitivity. In case of more than
one sensitivity simulation it is not necessary to make further
coupled reference simulations as the one applies to all sensi-
tivities.

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project POMODORO.
Table A1. Namelist settings required to activate the QCTM for EMAC version 1.11 and higher. Regridding is referred to as spatial interpolation of an offline field to the ECHAM5 Gaussian grid (see Jöckel, 2006). Triggering is referred to as the timed import of the regridded field into a running simulation (see Kerkweg et al., 2006).

### offline climatologies of radiatively active species

<table>
<thead>
<tr>
<th>&amp;CPL in rad4all.nml</th>
<th>user-defined mixing ratio for ich4_rad4all = 2 or 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>ch4vmr_rad4all = const. value</td>
<td>whole atmosphere: mixing ratio set to ch4vmr_rad4all</td>
</tr>
<tr>
<td>ich4_rad4all = 2 or</td>
<td>troposphere: ch4vmr_rad4all; decay with elevation above</td>
</tr>
<tr>
<td>ich4_rad4all = 3 or</td>
<td>offline CH₄ climatology</td>
</tr>
<tr>
<td>ich4_rad4all = 4</td>
<td>(preferred setting if climatology from coupled reference simulation)</td>
</tr>
</tbody>
</table>

| &REGRID in rad4all.nml              | offline CH₄ climatology: path and regridding         |
| var = “CH4razil=CH4”                |                                                      |

| &RGTEVENTS in rad4all.nml           | offline CH₄ climatology: triggering                  |
| RG_TRIG(5) = . . .                  |                                                      |

### offline climatology of chemical H₂O tendencies

| &CPL in h2o.nml                     | offline climatological CH₄ oxidation only           |
| L_H2O_TENDENCY = 2                 |                                                      |

| &CPL in hetchem.nml                 | no H₂O tendencies from thermodynamic re-partitioning |
| L_COUPLE_H2O = F                    |                                                      |

### PSC-related partitioning of H₂O based on offline climatology of total HNO₃

<table>
<thead>
<tr>
<th>&amp;CPL in psc.nml</th>
<th>H₂O partitioning based on offline climatology of total HNO₃</th>
</tr>
</thead>
<tbody>
<tr>
<td>l_feedback = F</td>
<td>(preferably from coupled reference simulation)</td>
</tr>
</tbody>
</table>

| &REGRID in offlem/psc_offline.nml   | offline total-HNO₃ climatology: path and regridding         |
| var = ’pref_HNO3=HNO3;tot;INT’      |                                                                |

| &RGTEVENTS in offlem.nml            | offline total-HNO₃ climatology: triggering                  |
| RG_TRIG(40) = . . .                 |                                                                |

| &CTRL in psc.nml                    | QCTM not tested for microphysical kinetic NAT scheme         |
| KinPar=F                            |                                                                |
References


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