



Supplement of

**Operational chemical weather forecasting with the ECCO online
Regional Air Quality Deterministic Prediction System version 023
(RAQDPS023) – Part 1: system description**

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1 Supplement

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13 **S1 GEM-related meteorological components and attributes of RAQDPS023**

14 **S1.1 Dynamical core**

15 The governing equations of atmospheric dynamics used by the GDPS 8.0.0, RDPS 8.0.0, and RAQDPS023 have been
16 described by Côté et al. (1998) and Girard et al. (2014). Prognostic variables included the horizontal wind components,
17 vertical velocity component in the hybrid ζ -coordinate (Sect. 2.1), virtual temperature, geopotential, specific humidity,
18 and physics-related atmospheric tracers such as cloud condensate mixing ratio and turbulent kinetic energy (TKE).
19 GEM was configured to be hydrostatic in all three forecast systems (e.g., CMC-RDPS-8.0.0, 2021b). Time integration
20 of this dynamical core was performed using a two-time-level, iterative-implicit, three-dimensional semi-Lagrangian
21 scheme (Côté et al., 1998; Girard et al., 2014; Husain and Girard, 2017). This semi-implicit, semi-Lagrangian
22 integration is unconditionally stable for any time step, thus allowing the choice of time step to be based solely on
23 accuracy requirements (Robert, 1981; Ritchie et al., 2022). A dynamics integration time step of 450 s was used by the
24 GDPS 8.0.0 and one of 300 s was used by both the RDPS 8.0.0 and RAQDPS023. The improved trajectory calculations
25 for semi-Lagrangian advection described by Husain and Girard (2017) were employed, which involve trapezoidal
26 averaging of wind components along air-parcel trajectories and cubic interpolation at the departure positions. These
27 calculations remove numerical inconsistencies in the discretization and interpolation used for the source terms between
28 the dynamical and trajectory equations.

29 Semi-implicit, semi-Lagrangian time integration schemes are highly efficient by virtue of circumventing the stiffness
30 of the dynamic equations, but semi-Lagrangian advection schemes are not inherently shape-preserving or mass-
31 conserving (e.g., Staniforth and Côté, 1991; Hansen et al., 2011; de Grandpré et al., 2016). Shape preservation for two
32 meteorological tracers (water vapour, cloud water) was implemented in all three forecast systems using a two-step,

33 iterative locally-mass-conserving (ILMC) monotonicity correction for tracer advection (Sørensen et al., 2013). The
34 ILMC scheme is an *a posteriori* filter that detects the lack of shape preservation of a forecasted solution, including
35 negative concentrations, and corrects it by redistributing the overshoot or undershoot mass to the surrounding cells
36 while preserving the mass of the forecasted solution. By imposing shape preservation, the creation of artificial local
37 minima or maxima by the advection scheme was avoided. The Bermejo and Conde (2002) global mass fixer was then
38 applied in the three forecast systems to impose mass conservation, where the LAM boundary-flux estimation was used
39 for the two regional systems (Aranami et al., 2015). It is worth mentioning in this context that Staniforth and Côté
40 (1991) observed that semi-Lagrangian advection schemes handle sharp discontinuities (e.g., clouds) better than
41 Eulerian advection schemes and dispersion errors are localized around the discontinuity, which suggests that mass-
42 fixing adjustments will tend to be localized as well. McTaggart-Cowan et al. (2019a) noted that the use of these two
43 schemes in GEM improved the conservation of liquid water static energy and total water mixing ratio, with the result
44 that total global precipitation in the GDPS was reduced by approximately 5%. A surface pressure adjustment was also
45 made for moist-air mass conservation (McTaggart-Cowan et al., 2019a; CMC-RDPS-8.0.0, 2021b).

46 While semi-Lagrangian advection schemes have some inherent numerical diffusion (e.g., Staniforth and Côté, 1991),
47 artificial horizontal diffusion was also implemented to damp spurious numerical noise at the shortest spatial scales
48 (e.g., Mailhot et al., 2006). In all three systems a Del-4 operator was applied to momentum variables (10%) and a
49 Del-6 operator (1%) was applied to potential temperature, but tracers were not affected. As well, a Del-2 operator with
50 enhanced horizontal diffusion coefficients was applied to temperature and momentum variables in a sponge layer below
51 the model lid (top eight levels) to prevent wave energy reflection from the rigid model lid (Mailhot et al., 2006; CMC-
52 GDPS-8.0.0, 2021b; CMC-RDPS-8.0.0, 2021b).

53 **S1.2 Land-surface and water-surface schemes**

54 The ISBA (Interactions between Soil–Biosphere–Atmosphere) land-surface scheme was used in all three forecast
55 systems to parameterize land surface processes and atmosphere-surface exchanges of heat and moisture (Noilhan and
56 Planton, 1989; Bélair et al., 2003a,b). ISBA takes a mosaic approach that considers subgrid “tiles” of four major
57 surface types: land; glacier; water; and sea ice. A GEM surface grid cell may be composed of up to four tiles to account
58 for the presence of these different surface types. Ten prognostic variables are considered in the ISBA scheme for each
59 land surface tile, including two surface temperature variables, three soil moisture variables, one vegetation canopy
60 variable, and four variables describing snowpack characteristics. Each prognostic equation is solved based on an
61 energy or water budget formulation with a force-restore term to moderate diurnal variations. Initial values of land
62 surface temperature and moisture, deep-soil temperature and moisture, and snow depth, snow albedo and snow density
63 were derived from analyses and then predicted using the ISBA scheme. Emissivities of bare soil and snow were derived
64 from climatology and observations, respectively.

65 The treatment of glacier surface tiles closely followed that for land surfaces and combined a surface-energy-budget
66 model with a two-layer force-restore term to predict glacier temperature profiles (McTaggart-Cowan et al., 2019a).
67 Updated ice emissivities were based on observations.

68 The treatment of water surface tiles depends on the forecast system. The GDPS 8.0.0 was coupled to an ocean model
69 whereas the RDPS 8.0.0 and RAQDPS023 obtained initial sea-surface temperatures from surface analyses and then
70 employed the one-dimensional thermodynamic mixed-layer model of Zeng and Beljaars (2005) to predict the diurnal
71 evolution of the water skin temperature and underlying mixed-layer temperature. This two-layer scheme provides a
72 more realistic description of diurnal variations in water surface temperature and hence surface fluxes over water. Note
73 that the GDPS 8.0.0 also used the mixed-layer scheme over lakes as did the RDPS 8.0.0 and RAQDPS023 (McTaggart-
74 Cowan et al., 2019a; CMC-GDPS-8.0.0, 2021b; CMC-RDPS-8.0.0, 2021b). The surface roughness length for
75 momentum over water surfaces was based on the Charnock (1955) scheme while roughness lengths over water for heat
76 and moisture came from Deacu et al. (2012).

77 Sea-ice (and frozen lake) surfaces were parameterized using a one-dimensional heat diffusion model (Semtner, 1976)
78 with a simple, three-level vertical discretization (McTaggart-Cowan et al., 2019a). Initial sea-ice thickness values were
79 derived from climatology but then evolved in time, whereas sea-ice fractions were derived from surface analyses, were
80 fixed in time, and were checked for consistency with the sea-ice thickness. A 3% lead fraction in sea ice was assumed
81 over salt water (McTaggart-Cowan et al., 2019a).

82 Other geophysical variables needed by the ISBA land-surface scheme include surface roughness length (except over
83 water), surface albedo, vegetation characteristics (vegetation type and fraction, leaf area index), soil texture and soil
84 thermal and hydraulic coefficients, and glacier fraction. For grid cells with heterogeneous land surface cover (e.g.,
85 some bare soil/some vegetation or some land/some water), some grid-scale geophysical fields such as roughness length,
86 albedo, and emissivity were determined as weighted averages of tile-specific geophysical fields.

87 **S1.3 Surface-layer scheme**

88 Monin-Obukhov surface-layer similarity theory was used in all three forecast systems to describe turbulent surface
89 fluxes of momentum, heat, and moisture and state variable profiles in the model surface layer, which extends from the
90 Earth's surface to the first prognostic model level (located at approximately 10 m AGL for temperature and moisture
91 and 20 m AGL for winds on the staggered Charney-Phillips vertical grid: Sect. 2.1). Separate profiles were calculated
92 over each of the four surface types (tiles) and were then aggregated as a weighted average to provide a grid-cell value
93 (McTaggart-Cowan et al., 2019a). Stability functions proposed by Delage and Girard (1992) and Beljaars and Holtslag
94 (1991) were used for unstable and stable stratifications, respectively (CMC-GDPS-8.0.0, 2021b; CMC-RDPS-8.0.0,
95 2021b).

96 For stable conditions the following minimum Obukhov lengths were imposed to prevent surface-atmosphere
97 decoupling (Foken, 2006; McTaggart-Cowan et al., 2019a): 20 m over soil surfaces; 5 m over glaciers; and 10 m over
98 water surfaces and sea ice (CMC-GDPS-8.0.0, 2021a; CMC-RDPS-8.0.0, 2021a). The higher value over land surfaces
99 is consistent with a dependence on surface roughness (van de Wiel et al., 2007; Zheng et al., 2017).

100 **S1.4 Boundary-layer scheme and vertical diffusion**

101 Vertical diffusion in the planetary boundary layer (PBL) and free troposphere was described in all three forecast
102 systems by the vertical diffusion coefficient K_z , which is parameterized using a 1.5-order turbulence closure scheme
103 based on a predictive equation for subgrid-scale (SGS) turbulent kinetic energy (Benoit et al., 1989; Bélair et al., 1999).
104 A key parameter of the scheme is the gradient Richardson number Ri . Two turbulence length scales are also needed
105 by this scheme. A turbulence-regime-specific mixing length λ , which reflects the significant eddy size, is needed to
106 calculate both TKE and K_z , and a dissipation length scale λ_e is also needed to calculate TKE. The required lower
107 boundary conditions are obtained from the surface-layer parameterization (see previous subsection). K_z from GEM
108 was also needed by MACH (**Sect. 3.1**).

109 Note that a number of enhancements have been made to the original formulation of Benoit et al. (1989). First, λ was
110 calculated in the GDPS 8.0.0, RDPS 8.0.0, and RAQDPS023 using a blended formulation (McTaggart-Cowan and
111 Zadra, 2015) based on a local scheme for λ (Blackadar, 1962) and a non-local scheme for λ (Bougeault and Lacarrère,
112 1989). A treatment for Richardson number hysteresis across the temporal transitions from stable-to-unstable conditions
113 and unstable-to-stable conditions that uses an Ri transition zone rather than a single critical value of Ri has been
114 implemented (McTaggart-Cowan and Zadra, 2015). A 900 s relaxation time scale was also imposed on λ to dampen
115 the effects of turbulent transitions, and an upper bound of 50 m was imposed on λ_e (Bélair et al., 1999; McTaggart-
116 Cowan et al., 2019a).

117 Lastly, the enhancement of turbulent diffusion near the top of the cloudy boundary layer due to updrafts, downdrafts,
118 and related entrainment associated with boundary-layer clouds, including stratus, stratocumulus, and shallow cumulus
119 clouds, was represented by an extension to the TKE scheme with statistical SGS cloudiness called the MoisTKE
120 scheme (Bechtold and Siebesma, 1998; Mailhot and Bélair, 2000; Bélair et al., 2005; Mailhot et al., 2006). Three key
121 parameters were considered: turbulent flux of total cloud water; SGS cloud fraction; and a nondimensional flux
122 enhancement factor (Bélair et al., 2005). These cloud effects were activated only if the surface buoyancy flux was
123 positive and the cloud layer resided below 1.5 times the PBL height (McTaggart-Cowan et al., 2019a).

124 PBL height, which is also an important quantity for air quality, was diagnosed every time step for use in the Kain-
125 Fritsch deep convection scheme (Sect. S1.7) as well as the MoisTKE scheme. A bulk Richardson number (Ri_b) method
126 was used in which PBL height was determined to be the lowest momentum vertical level at which Ri_b reaches a critical

127 value of 0.25 by scanning the R_{ib} profile upward from the surface (Seidel et al., 2012; McTaggart-Cowan et al., 2019a).
128 PBL height from GEM was also needed by MACH (Sect. 3.13).

129 **S1.5 Radiation scheme**

130 Solar and infrared radiation were parameterized in all three forecast systems using a correlated-k distribution technique
131 (Li and Barker, 2005). Nine prescribed longwave bands and four shortwave bands were considered. The updated
132 version of this scheme that is now used by GEM included an improved treatment of the water vapour continuum and
133 solar radiation absorption by methane (McTaggart-Cowan et al., 2019a). A monthly climatological O_3 distribution
134 constructed from several data sets was used along with three-dimensional climatologies of other trace gases (CH_4 , NO ,
135 CFC-11, CFC-12) appropriate for the year 2015 (McTaggart-Cowan et al., 2019a). A simple parameterization of
136 methane oxidation by the hydroxyl radical (OH) was also considered as a source of water vapour in the middle
137 atmosphere (McTaggart-Cowan et al., 2019a).

138 The GEM v5.1 code includes the addition of O_3 as a new prognostic tracer, which made it possible for the GDPS 8.0.0
139 to assimilate satellite O_3 measurements (CMC-GDPS-8.0.0, 2021a). In order to represent O_3 as a standalone chemical
140 constituent (i.e., no explicit reactive chemistry), a linearized approach was used to parameterize the photochemical
141 production and loss terms (McLinden et al., 2000). The use of this parameterization scheme, called LINOZ, in GEM
142 has been shown to be both robust and very fast (de Grandpré et al., 2016).

143 **S1.6 Grid-scale cloud and precipitation scheme**

144 The Sundqvist grid-scale condensation scheme (Sundqvist et al., 1989; Pudykiewicz et al., 1992) was used in the GDPS
145 8.0.0, RDPS 8.0.0, and RAQDPS023 to represent grid-scale cloud and precipitation processes. Four water phases were
146 considered: water vapour; cloud water; rain water; and snow water. One modification to the original scheme was to
147 use an updated cloud-to-precipitable water autoconversion rate (McTaggart-Cowan et al., 2019a; CMC-GDPS-8.0.0,
148 2021b; CMC-RDPS-8.0.0, 2021b). Lastly, this grid-scale cloud scheme was coupled to the three SGS cloud schemes
149 described next via the liquid and solid condensate mass detrained from convective updrafts in these cumulus
150 parameterization schemes, which were added to the grid-scale condensate mass (McTaggart-Cowan et al., 2019a).

151 **S1.7 Cumulus parameterization schemes**

152 Three separate cumulus parameterization schemes were used in the GDPS 8.0.0, RDPS 8.0.0, and RAQDPS023 to
153 represent three different modes of SGS unstable moist vertical motions, namely deep convection, shallow convection,
154 and mid-level convection. The term “deep convection” denotes convective clouds connected to the PBL with depths
155 of at least 3 km; the term “shallow convection” refers to the subset of convective clouds whose bases are located in the
156 PBL but which have updraft depths between 0.5 and 3 km that penetrate the PBL capping inversion (thus distinguishing
157 them from the boundary-layer clouds parameterized in the MoisTKE scheme described in Sect. S1.4); and the term
158 “mid-level convection” (or elevated convection) refers to convective clouds whose bases are located above the PBL.

159 All three schemes are formulated to transport momentum as well as heat and moisture vertically. Each scheme is mass-
160 flux-based, and their implementations share five important structural elements: trigger; updraft; downdraft; solver; and
161 closure (McTaggart-Cowan et al., 2019a). Calculations were performed sequentially for these three schemes, starting
162 with the deep convection scheme and finishing with the mid-level convection scheme.

163 For SGS deep convection a modified version of the Kain–Fritsch deep-convection scheme (Kain and Fritsch, 1990,
164 1992) was used, which has been designed for horizontal grid spacing approaching the “gray zone” over which deep
165 convection is partially resolved (i.e., horizontal grid spacing on the order of 1–10 km). The modifications included a
166 Lagrangian treatment of convective initiation, object-based clouds, and a vertical transfer of momentum. See
167 McTaggart-Cowan et al. (2019a) for a summary and McTaggart-Cowan et al. (2019b) for a detailed description of
168 these modifications. Note that the trigger-function parameters are sensitive to horizontal grid spacing, so different
169 values were specified for the GDPS 8.0.0 versus the RDPS 8.0.0 and RAQDPS023 (see Table 2 of McTaggart-Cowan
170 et al., 2019a).

171 SGS shallow convection was represented with a mass-flux scheme based on Bechtold et al. (2001). Simple
172 representations of cloud entrainment (Siebesma et al., 2003) and constant fractional detrainment (Siebesma and
173 Cuijpers, 1995) were used in the GEM implementation along with a simple, non-iterative closure that assumes that the
174 shallow convective clouds in a grid cell are in a state of quasi-equilibrium with boundary-layer forcing (Bechtold et
175 al., 2014). Note that downdrafts and precipitation were not parameterized by this scheme.

176 Lastly, SGS mid-level convection was treated using components of the deep convection scheme if certain conditions
177 were met (McTaggart-Cowan et al., 2019a). One condition was that a minimum upward grid-scale mass flux was
178 present: the required threshold was higher for the GDPS 8.0.0 than the RDPS 8.0.0 / RAQDPS023 ($1 \times 10^7 \text{ kg s}^{-1}$ vs.
179 $7 \times 10^6 \text{ kg s}^{-1}$). Another condition was that there must be a layer at least 500 m AGL that can build a cloud at least 2 km
180 thick after the layer is raised to its lifting condensation level. Note that the deep and shallow convection schemes
181 would not be activated in a low convective-available-potential-energy (CAPE) environment, but the mid-level
182 convection scheme could be triggered even in low-CAPE environments.

183 **S1.8 Gravity wave drag and low-level orographic blocking schemes**

184 SGS orography is associated with two processes that must be parameterized: (i) vertical momentum transport due to
185 the breaking of mountain-induced gravity waves (“mountain wave drag”) and (ii) blocking of low-level upstream flows
186 by mountain barriers. Parameterizations due to McFarlane (1987) for mountain wave drag and Lott and Miller (1997)
187 for low-level blocking (with an enhanced drag coefficient: Wells et al., 2008; Vosper et al., 2009) were used by all
188 three forecast systems (McTaggart-Cowan et al., 2019a; CMC-GDPS-8.0.0, 2021b; CMC-RDPS-8.0.0, 2021b). In
189 addition, the Hines (1997a,b) Doppler spread parametrization for non-orographic gravity-wave drag due to imbalances
190 and geostrophic adjustment associated with a variety of transient mesoscale meteorological features was also used to

191 simulate the upper-level circulation accurately (Charron et al., 2002; McTaggart-Cowan et al., 2019a; CMC-GDPS-
192 8.0.0, 2021b; CMC-RDPS-8.0.0, 2021b)

193 The geophysical variables required by these schemes include gridded orography plus SGS orographic parameters for
194 gravity wave drag and low-level blocking. Note that it is a standard procedure to apply a smoothing operator to the
195 model orography field in order to minimize stationary noise generation due to small-scale resolved orographic features.
196 A numerical filter with a wavelength cutoff at $3\Delta x$ for the GDPS 8.0.0 and $5\Delta x$ for the RDPS 8.0.0 and RAQDPS023
197 was applied. SGS orographic parameters were computed using an analytically-extended reference orography spectrum
198 and a 5 km cut-off for horizontal terrain scales (Beljaars et al., 2004; McTaggart-Cowan et al., 2019a; CMC-GDPS-
199 8.0.0, 2021b; CMC-RDPS-8.0.0, 2021b).

200 **S2 Overview of anthropogenic national emissions inventories pre-processing and processing**

201 A number of adjustments had to be made to the projected national emissions inventories before they underwent the
202 processing required to create the SET4.0.0 emissions input files. For the projected 2020 Canadian Air Pollutant
203 Emissions Inventory (APEI), emissions from two emissions sectors, prescribed burning and agricultural-land wind
204 erosion, were removed from the inventory because of insufficient information to allocate these emissions to specific
205 days of the year or times of day. Projected 2020 aircraft in-flight cruise emissions were also removed because they
206 would not materially affect short-term surface pollutant concentrations due to the altitude (6-10 km above sea level) at
207 which they were emitted, whereas projected 2020 aircraft landing and takeoff emissions were kept. Projected 2020
208 residential wood combustion (RWC) emission values were replaced by 2010 APEI values due to concerns over some
209 aspects of the revised estimation methodology used to calculate RWC emissions for the 2015 APEI (ECCC, 2013,
210 2016). In addition, projected 2020 emission values for one power plant and one large industrial facility that had been
211 active in 2015 but that were known to have closed by 2020 were removed from the major-point-source emissions file
212 (Moran et al., 2021).

213 Adjustments were also made to Canadian projected 2020 fugitive dust emissions from paved and unpaved roads and
214 from construction activities. Fugitive dust emissions are the largest component of anthropogenic $PM_{2.5}$ and PM_{10}
215 emissions, of which the main contributing emission sectors are vehicular traffic on paved and unpaved road surfaces
216 and agricultural and construction activities. Emissions estimation methodologies are always subject to re-evaluation
217 and improvement, and a new methodology to estimate fugitive dust emissions from paved and unpaved road surfaces
218 was introduced by ECCC in 2020 for the 2018 APEI (ECCC, 2020). In order to account for these methodological
219 changes in the projected 2020 APEI, whose base 2015 fugitive road dust emissions had been calculated using an older
220 methodology (ECCC, 2016), the paved-road fugitive dust emissions from the projected 2020 inventory for each
221 province and territory were scaled by a factor of 0.15 (i.e., decreased by 85%) and unpaved-road fugitive dust emissions
222 were scaled by a factor of 1.11 (i.e., increased by 11%).

223 One useful check on any Canadian CAC annual emissions inventory is to compare per capita Canadian and U.S.
224 emissions for the same emissions source sectors for the same year or closest year (since U.S. National Emissions
225 Inventories (NEIs) are only available every three years). Given that differences in population have been controlled for
226 by considering per capita values, one expects the per capita emissions values to be broadly similar given socioeconomic
227 and emission control similarities between the two countries. When this comparison was made for the scaled 2015
228 Canadian fugitive dust emissions from paved and unpaved road surfaces and from agricultural activities and residential
229 construction, the agreement with the closest U.S. CAC inventory, the 2014 U.S. NEI ([https://www.epa.gov/air-
230 emissions-inventories/2014-national-emissions-inventory-nei-data](https://www.epa.gov/air-emissions-inventories/2014-national-emissions-inventory-nei-data)), was reasonable. For 2015 fugitive dust emissions
231 from heavy construction, however, per capita Canadian emissions were 40 times higher than 2014 per capita U.S.
232 emissions. Moreover, the estimation methodology used to estimate 2015 Canadian heavy construction emissions had
233 not changed since 2012 and heavy construction emissions in the APEI had been held constant since 2013, which
234 suggested a high level of uncertainty for these emissions. It was thus decided to scale Canadian projected 2020 fugitive
235 dust emissions for heavy construction by a factor of 0.05 so that 2015 per capita Canadian emissions from this sector
236 would be within a factor of two of the 2014 per capita U.S. emissions.

237 The Canadian projected 2020 APEI then underwent two final preparatory steps by ECCC before the emissions
238 processing step. First, the point source file was separated into two files, a major-point-source file and a minor-point-
239 source file, where a 20-m stack height threshold was applied to assign individual point sources to the major- or minor-
240 point-source files based on stack height. Similarly, the area source file was separated by emissions source type into
241 multiple files, including on-road mobile sources, off-road mobile sources, marine sources, agricultural sources, other
242 fugitive dust area sources, and other non-fugitive-dust area sources. This separation by source type made it possible
243 to apply special treatments such as meteorological modulation to an emissions subset (Sect. 3.12), and also to perform
244 quality-control checks more easily after the emissions processing step. The second final preparatory step was to convert
245 the Canadian projected 2020 APEI into a different format that would be readable by the emissions processing system.
246 This step included augmentation of the APEI to include some non-standard data such as point source stack
247 characteristics (Sect. 3.13) and facility operating schedules (e.g., Sassi et al., 2021).

248 Several minor adjustments were also required to the projected 2023 U.S. NEI for emissions from commercial marine
249 vessels. These emissions had been pre-processed by the U.S. EPA for three non-overlapping modelling grids used by
250 the EPA: (i) emissions within U.S. federal waters for the continental U.S. on a 12-km horizontal grid; (ii) emissions
251 for Alaska on a 9-km grid; and (iii) emissions for other Pacific and Atlantic Ocean areas on a 4-km grid. However,
252 there were spatial gaps among these three emissions regions, which were filled by inserting emissions from another
253 marine emissions data set that was used for Phase 4 of the Air Quality Model International Initiative (AQMEII-4)
254 project (Galmarini et al., 2021). The projected 2023 U.S. inventory also contained marine emissions over some
255 Canadian waters: these emissions were removed to avoid double counting with the Canadian inventory. Note that the

256 EPA had created an emissions-processing version of the projected 2023 U.S. NEI and had also separated the surface
257 emissions into multiple files by source type for easier processing and tracking (U.S. EPA, 2019).

258 **S2.1 Emissions processing**

259 Once the above inventory adjustments had been made, the three projected national inventories were processed using
260 version 4.7 of the SMOKE (Sparse Matrix Operation Kernel Estimation) emissions processing system software (see
261 <https://www.emascenter.org/smoke/>) to generate the SET4.0.0 set of major-point-source emissions files and gridded
262 surface emissions files. Two types of emissions files were needed because the separate major-point-source files
263 contained emissions for large individual sources such as power plants and smelters for which plume rise was calculated
264 by GEM-MACH (Sect. 3.13), whereas the surface emissions were the sum of area sources, on-road mobile sources,
265 off-road mobile sources, and minor point sources for which plume rise was not relevant. Note that there were also two
266 types of surface emissions identified in the SET4.0.0 set, one type identifying fugitive-dust area emissions and the
267 other type identifying all other surface emissions. This made it possible to apply meteorological modulation only to
268 fugitive-dust area emissions (cf. Sect. 3.12).

269 The SMOKE emissions processing system must perform four major processing tasks: (i) temporal allocation, where
270 annual/monthly inventory emissions were allocated to specific hours of specific days and months of the year based on
271 a set of source-sector-specific monthly, day-of-the-week (DOW), and diurnal temporal profiles; (ii) spatial allocation,
272 where emissions reported by jurisdiction such as a province or county were allocated to the group of model grid cells
273 that overlay that jurisdiction in as realistic a pattern as possible using source-sector-specific gridded spatial surrogate
274 fields; (iii) chemical speciation, where four of the CACs contained in the inventories (NO_x , VOC, $\text{PM}_{2.5}$, PM_{10}) were
275 separated into more detailed model species using source-sector-specific speciation profiles; and (iv) PM_x size
276 disaggregation ($x = 2.5$ or 10), where PM_x emissions from the inventory were allocated to different model PM size
277 sections or modes using source-sector-specific PM size distribution profiles (e.g., Dickson and Oliver, 1991; Houyoux
278 et al., 2000; Moran et al., 2013; Matthias et al., 2018; Zhang et al., 2018; Boutzis et al., 2020).

279 Each SET4.0.0 emissions file generated by SMOKE was specific to a particular day of the week and month of the year,
280 and each file contained 24 hours of gridded emission fields of model species (where NO_x emissions were split into
281 emissions of NO , NO_2 , and HONO, $\text{PM}_{2.5}$ and PM_{10} emissions were each split into emissions of six model PM chemical
282 components, and VOC emissions were split into emissions of 11 model lumped VOC species for the ADOM-2 gas-
283 phase chemistry mechanism) (Stroud et al., 2008; Zhang et al., 2018; see Table 3).

284 For the first emissions processing task, temporal allocation, SMOKE used 82 monthly temporal profiles, 22 DOW
285 profiles, and 29 diurnal profiles to process the Canadian emissions, 9,971 monthly profiles, 12 DOW profiles, and 95
286 diurnal profiles to process the U.S. emissions, and 161 monthly profiles, 15 DOW profiles, and 49 diurnal profiles to
287 process the Mexican emissions (CMC-RAQDPS-023, 2021). For the second task, spatial allocation, SMOKE
288 employed 55 source-sector-specific gridded spatial surrogate fields to allocate emissions for each Canadian province

289 or territory to one or more grid cells on the RAQDPS023 10-km continental Yin grid (Fig. 2), 42 gridded spatial
290 surrogate fields for each U.S. county, and 15 gridded spatial surrogate fields for each Mexican county (CMC-RAQDPS-
291 023, 2021). The gridded surface emissions from these different jurisdictions and source types were then summed for
292 each model grid cell. Major-point-source emissions, on the other hand, were assigned to a single model grid cell based
293 on their latitude-longitude coordinates while remaining in separate records for subsequent plume-rise calculations
294 (Sect. 3.13). For the third task, chemical speciation, SMOKE used 331 VOC and 85 PM speciation profiles from the
295 U.S. EPA's SPECIATE 4.5 database (<https://www.epa.gov/air-emissions-modeling/speciate-1>; Reff et al., 2009;
296 Simon et al., 2010). Only two speciation profiles were used to speciate NO_x emissions: one profile for on-road and
297 off-road sources, which included a small HONO component as well as NO and NO₂ components, and a generic profile
298 with only NO and NO₂ components for point and area sources. Lastly, the fourth task, PM size disaggregation, was
299 handled simply by assigning PM_{2.5} emissions to the first RAQDPS023 PM size bin and assigning PM_{coarse} emissions,
300 calculated as the difference between PM₁₀ and PM_{2.5} emissions, to the second RAQDPS023 PM size bin. Following
301 the chemical speciation and PM size disaggregation steps the SET4.0.0 output files produced by SMOKE contained
302 hourly gridded emission fields of 18 gases and 12 particle size bin/chemical components for a total of 30 emitted model
303 species: SO₂, (H₂)SO₄, NO, NO₂, HONO, CO, C₃H₈, ALKA, ETHE, ALKE, TOLU, AROM, HCHO, ALD₂, MEK,
304 CRES, ISOP, NH₃, SU₁, SU₂, NI₁, NI₂, AM₁, AM₂, EC₁, EC₂, PC₁, PC₂, CM₁, and CM₂.

305 It should be noted that two more adjustments were made to all fugitive dust emissions *after* SMOKE processing. One
306 adjustment was done immediately as a (static) post-processing step, where a transportable-fraction (TF) scaling factor
307 field was applied to account for "home-grid-cell" SGS particle settling and particle removal by interception and
308 impaction by vegetation (e.g., Pace, 2005). The TF scaling factor ranged between 0 and 1 and was dependent on the
309 fractional coverage of different land-use classes in each grid cell, where forest classes had a TF value of zero and
310 agricultural classes had a value of 0.75. As described in Sect. 3.12 the other (dynamic) adjustment was to apply
311 meteorological modulation at each chemistry time step during each RAQDPS023 simulation to account for the day-
312 specific dampening effects of precipitation and snow cover on fugitive dust emissions. In order to accommodate the
313 model calculations required for meteorological modulation and plume rise (Sects. 3.12 and 3.13), three sets of
314 emissions fields must be read by the RAQDPS023 each hour from two separate day-specific files: (i) major-point
315 source emissions; and (ii) surface emissions files consisting of fugitive dust emissions (type F) and other surface
316 emissions (type E). SET4.0.0 thus consisted of two input files per day of a typical week of each month, or 168 files in
317 total. More information related to the SMOKE processing required to create GEM-MACH emissions files may be
318 found in Zhang et al. (2018) and Sassi et al. (2021).

319 **S3 Description of Canadian Forest Fire Emissions Prediction System (CFFEPS) version 4.1**

320 CFFEPS v4.1 consists of a fire-growth module, a fire emissions module, and a thermodynamic-based module to
321 estimate the vertical penetration height of a smoke plume (Chen et al., 2019a,b). Outputs from the Canadian Wildland

322 Fire Information System (CWFIS), which is an operational product developed and managed by the Canadian Forest
323 Service (CFS), underpin the CFFEPS fire-growth module. The CWFIS is itself composed of three major subsystems:
324 (i) the Canadian Forest Fire Weather Index (FWI) subsystem, which accounts for the influence of four surface weather
325 elements (dry-bulb temperature, relative humidity, wind speed, and 24 hour accumulated precipitation) on the relative
326 risk for wildfire occurrence; (ii) the Fire Monitoring, Mapping, and Modeling (Fire M3) subsystem, which identifies
327 and locates active fires (“hotspots”) using NRT satellite imagery; and (iii) the Canadian Forest Fire Behaviour
328 Prediction (FBP) subsystem, which estimates fire fuel consumption and fire intensity
329 (<https://cwfis.cfs.nrcan.gc.ca/home>).

330 Input data needed by the FWI subsystem include observing station measurements of 2 m dry-bulb temperature and
331 relative humidity, 2 m dew point, 10 m wind speed, surface pressure, and past 24 hour rainfall from about 2500
332 Canadian and U.S. weather stations plus 72 hour ECCC weather forecasts for the same station locations. The six
333 primary outputs of the FWI subsystem are numerical moisture codes for the forest-floor fine-fuel layer (i.e., leaf litter,
334 needles, cured grass, small twigs; 1.2 cm deep), duff layer (layer of partially/fully decomposed organic materials
335 beneath the fine-fuel layer; 7 cm deep), and deep duff layer (beneath the duff layer; ~18 cm deep), an initial spread
336 index (a function of surface wind speed and fine-fuel moisture content), total subsurface fuel available for combustion,
337 and a combined FWI numerical value (Van Wagner, 1987; <https://cwfis.cfs.nrcan.gc.ca/background/summary/fwi>).

338 As of 2021 the fire hotspot data needed by the Fire M3 subsystem are obtained from the 4- μm and 11- μm infrared
339 channel imagery of two radiometer instruments carried on four U.S. polar-orbiting satellites: (i) the Moderate
340 Resolution Imaging Spectroradiometer (MODIS) on two NASA satellites; and (ii) the Visible Infrared Imaging
341 Radiometer Suite (VIIRS) on two more NASA satellites (<https://cwfis.cfs.nrcan.gc.ca/background/dsm/fm3>). One to
342 two retrievals covering all of Canada, including remote areas, are available from each satellite each day. However,
343 hotspot detection may not be possible due to cloud or thick smoke obscuration of the surface or if fire size or intensity
344 is small, so not every fire, even large fires, will be detected every day. In addition, internal data filtering and
345 measurements at other wavelengths are applied to remove false positives (e.g., industrial sources, roofs of large
346 buildings, cloud edges). To support ECCC the CFS has extended CWFIS capabilities to provide fire hotspots over the
347 U.S. as well as Canada throughout the year instead of just the May to September period, which historically was the
348 typical wildfire season in Canada. Note that the satellite measurements have a time lag or latency between one and
349 seven hours due to the time required for image acquisition and distribution, hence the NRT descriptor.

350 The third major CWFIS component is the FBP subsystem, which uses FWI system outputs together with fuel type and
351 topography to predict fire rate of spread (ROS; m min^{-1}), fire intensity (kW m^{-1}), crown fraction burned and fire type
352 (crown fire or surface fire), and total fuel consumption ($\text{kg dry biomass m}^{-2}$) at hotspot locations
353 (<https://cwfis.cfs.nrcan.gc.ca/background/summary/fbp>). Sixteen fuel types such as Boreal Spruce and Ponderosa
354 Pine–Douglas Fir are considered (e.g., <https://cwfis.cfs.nrcan.gc.ca/background/fueltypes/c2>).

355 **S3.1 Forest fire emission forecasts**

356 CFFEPS v4.1 includes the source code for the FWI and FBP subsystems, which allows meteorological forecast fields
357 from GEM to be considered for every hour and not just for noontime as in the CFS operational version of the CWFIS.
358 Hourly biomass burning (BB) emissions are then estimated in CFFEPS v4.1 using a bottom-up, process-based
359 approach, where emissions per fire (or hotspot) per day are the product of estimated burn area per hotspot, estimated
360 fuel consumption per unit area, and emission factors for different chemical species and combustion stages (Chen et al.,
361 2019a,b). Hotspot and associated fuel type information for the previous 24 hours is provided by the CWFIS Fire M3
362 subsystem. Initial daily burn area per hotspot is a prescribed value by province and fuel type that is updated periodically
363 by CFS based on their analysis of CWFIS historical averages of reported burn areas by eco-region and fuel type; it is
364 assumed to remain constant for each day of the forecast (i.e., persistence). Fuel consumption values are estimated by
365 fuel type via the FBP subsystem driven with forecast meteorology. The incorporation of the FBP subsystem allows
366 CFFEPS v4.1 to account for spatially variable biomass/fuel inputs using forest composition data from Canada's
367 National Forest Inventory (Beaudoin et al., 2014, 2018) (Chen and Menelaou, 2021).

368 To calculate fire emissions, CFFEPS v4.1 uses FBP fuel combustion estimates to account for emissions from three
369 combustion stages: flaming, smoldering, and residual combustion (Chen et al., 2019b). A look-up table is used to
370 allocate fuel consumption between these different stages (Table 2 of Chen et al., 2019b), which have different burn
371 durations: 15 minutes for flaming followed by several hours for smoldering and several hours for residual combustion,
372 depending on the thickness of the forest-floor organic duff layer (Chen et al., 2019b). Different emission factors, which
373 are constant parameters representing grams of emissions per kilogram of fuel consumed, are applied to estimate
374 emissions from a wildfire for each combustion stage for eight chemical species: CO, CH₄, NO_x, NH₃, SO₂, non-
375 methane hydrocarbons (NMHC), PM_{2.5}, and PM₁₀ (Table 3 of Chen et al., 2019b). Chemical speciation is further
376 applied for NMHC emissions (cf. Sect. S2) to calculate ADOM-2 VOC species emissions using the U.S. EPA
377 SPECIATEv4.5 speciation profiles for flaming (#95425) and smoldering and residual combustion (#95428)
378 (<https://www.epa.gov/air-emissions-modeling/speciate-1>; Chen et al., 2019a). Similarly, PM_{2.5} and PM₁₀ emissions
379 from CFFEPS are subdivided into six PM chemical components with the following EPA SPECIATEv4.5 wildfire PM
380 speciation profile: POM (78.5%), EC (9.5%), CM (9.7%), SU (1.3%), NI (0.1%), and AM (0.9%) (Chen et al., 2019a).

381 Fire emission forecasts were calculated by CFFEPS v4.1 prior to each RAQDPS-FW023 forecast simulation using the
382 latest hotspot information for the past 24 hours from the operational CFS version of CWFIS. To obtain 72 hour
383 forecasts of hourly BB emission values, current hotspots were assumed to persist for the next 72 hours and hourly
384 changes in biomass fuel consumption were parameterized for a 24-hour period based on burn duration for the three
385 combustion stages, fire ROS, and the average diurnal variation of the moisture content of the forest-floor litter layer,
386 which increases as moisture content decreases (Chen et al., 2019b).

387 In the previous version of CFFEPS, v2.06, the diurnal distribution of daily burn area, which is used to allocate daily to
388 hourly emissions, was estimated by a fixed weighted-normalization approach using a predefined diurnal profile, with
389 peak fire growth at 16:00 local standard time (LST) and minimal near constant growth between 0:00 to 8:00 LST (Chen
390 et al., 2019b). In CFFEPS v4.1, however, the diurnal distribution of daily burn area was determined based on a
391 weighted moving average of the ratio of hourly fire ROS as estimated by the FBP subsystem to the sum of hourly ROS
392 values for the previous 24 hours (Chen and Menelaou, 2021). This new scheme should be more realistic as it is based
393 on hourly forecast meteorology rather than a static temporal profile.

394 **S3.2 Plume rise**

395 Plume-rise schemes for smokestacks like the Briggs (1984) scheme (cf. Sect. 3.13) have also been used to estimate
396 wildfire plume rise, but wildland fires are much broader than smokestacks and they lack any initial vertical velocity;
397 instead, they generate their own buoyant plume. CFFEPS v4.1 uses an alternate thermodynamic approach described
398 by Anderson et al. (2011) to parameterize wildfire plume rise. Hourly plume injection height is estimated based on a
399 fire energy balance such that the buoyancy of a smoke plume is thermodynamically balanced with the forecast vertical
400 temperature profile of the atmospheric column above the fire while accounting for fuel moisture evaporation and
401 condensation.

402 In earlier versions of CFFEPS, the plume-rise scheme was based on lifting the smoke plume dry adiabatically at a
403 constant rate at hotspot locations and then equilibrating the buoyant plume with the local environmental lapse rate
404 derived from a simplified vertical temperature profile determined by temperatures on five pressure levels: surface, 850,
405 700, 500, and 250 hPa (Chen et al., 2019b). CFFEPS v4.1 instead uses the higher-resolution forecast vertical
406 temperature profiles that are available on the RAQDPS-FW023 84-layer vertical grid (Sect. 2.1). The thermodynamic
407 calculation of plume heat flux in CFFEPS v4.1 was also updated to account for the heat lost due to the latent heat of
408 evaporation of fuel moisture during combustion. At the same time, as the plume passes the atmospheric lifting
409 condensation level, the latent heat of condensation of plume water vapour is added to the energy of the plume aloft to
410 obtain the final injection height within the atmospheric column. These detailed physical processes are considered
411 hourly for each hotspot as the fire thermal energy and the atmospheric vertical temperature profile change, and the final
412 hourly plume injection height is used to prescribe the vertical distribution of emissions within the column below the
413 plume injection height (Chen et al., 2019b; Chen and Menelaou, 2021).

414 **S3.3 Outputs and file structure**

415 CFFEPS v4.1 outputs are saved in one file similar in format to the anthropogenic major-point-source emissions input
416 file. The outputs are estimated hourly plume injection height (ZPLM meters), plume vertical distribution ratio
417 (RSMOKE), and hourly speciated emissions (g s^{-1}) over the 72-hour forecast period at each hotspot location as
418 specified by latitude/longitude. There are 21 ADOM-2 gas-phase and 12 particle-phase emitted species whose
419 emissions are output (identified by a leading “E”): ECO, ECO2, ENO, ENO2, ENH3, ESO2, ESO4, EA2, EA3, EALD,

420 EARO, EC38, ECRE, EETH, EHCH, EISO, EMEK, ETOL, EOTH, EC1, EC2, EAM1, EAM2, ECM1, ECM2, EEC1,
421 EEC2, ENI1, ENI2, EPC1, EPC2, ESU1, and ESU2 (cf. Table 3). Note that ECO2, EC1 (methane), and EC2 (ethane)
422 are not needed by the ADOM-2 gas-phase chemistry mechanism but may be used in other chemistry parameterizations.
423 This BB emissions file was input directly to the RAQDPS-FW023, which then allocated the emissions to model surface
424 and elevated grid cells.

425 **S3.4 Limitations**

426 CFFEPS v4.1 has a number of limitations that should be kept in mind (Chen et al., 2019a). First, as already noted
427 above some fires may not be detected due to (i) their small size, (ii) their initiation after the last available satellite
428 overpass, or (iii) the presence of thick cloud or heavy smoke above the fire. Second, detected fires are assumed to
429 continue burning but to remain the same size over the duration of the forecast. Third, fire emissions are treated as point
430 sources and hence are assigned to a single grid cell. And fourth, wildfire combustion heat and emitted smoke are
431 assumed not to affect meteorological fields, which may not be a good assumption for large, intense fires.

432

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