

Numerical Simulation of a Clear Air Turbulence (CAT) event over Northern India using WRF modelling system

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Abstract

Atmospheric turbulence is a primary meteorological hazard to en-route air traffic. The role of Clear Air Turbulence (CAT) for various processes in the atmosphere is still ambiguous. An Air India flight AI462 encountered severe CAT on 19 April 2018. The present study simulates the CAT event and is focused on understanding and investigating favourable conditions for the occurrence of CAT. Weather Research and Forecasting (WRF) Model V4.0.3 has been used to simulate the turbulence. The 6-hourly NCEP FNL Operational Global Analysis data at $0.25^\circ \times 0.25^\circ$ resolution is taken as input to provide the model's initial and lateral boundary conditions. For simulating the atmospheric environments at the time of the event, Yonsei University Scheme, WSM 3-Class Simple Ice Scheme, Kain-Fritsch (New-Eta) Scheme, Rapid Radiative Transfer Model (RRTM) Scheme, and Revised MM5 Monin-Obukhov Scheme are used. This study shows that Vertical Velocity, Geopotential Thickness, Wind Shear and Bulk Richardson Number Shear are correlated with CAT as the model predicted both upward and downward velocity very close to each other between 400 hPa and 550 hPa levels along with strong geopotential thickness gradient and strong wind shear gradient near the accident location. This could lead to CAT. Model simulated variations in CAT Index with change in time and pressure levels. CAT dissipates as we go higher in the atmosphere above 550 hPa.

Keywords: Clear Air Turbulence, vertical wind shear, CAT Index, Ellrod's Index, aviation meteorology, aircraft hazards.

1 Introduction

Turbulence is a primary meteorological hazard to en-route air traffic. At high altitudes, aircraft may encounter turbulence unexpectedly without any significant cloudiness. Clear Air

31 Turbulence (CAT) occurs when severe turbulence occurs in an environment with no clouds.
32 It creates a ferocious buffeting effect in aircraft. The role of CAT for various processes in the
33 atmosphere and vice versa is still ambiguous. CAT occurs in a statically stable shear layer,
34 typically found amid the high troposphere and low stratosphere. CAT is non-convective
35 turbulence not within the Planetary Boundary Layer (PBL). CAT is “aircraft turbulence that
36 occurs at 5.6 km or higher altitudes, either in cloud-free condition or within stratiform
37 clouds” (Ellrod et al. 2003). According to the Meteorological College office (1997), CAT is
38 horizontally 80-500 km along wind direction and 20-100 km in the across-wind direction
39 having vertical dimensions 500-1000 m. “CAT exists in the atmosphere for about 30 minutes
40 to 24 hrs” (Stefan et al. 2020).

41 “The physical impact of CAT on crew and passengers varies from discomfort to
42 injuries, loss of flight control, and in some sporadic instances, fatalities have befallen.
43 Repeated turbulence encounters that occur during the lifetime of an aircraft might advance to
44 metal fatigue and, in sporadic cases, structural failure” (Ellrod et al. 2003). Commercial
45 airlines have economic loss also due to meteorological hazards as there is a significant
46 increase in fuel consumption during turbulent flights. So, knowing the unknown about CAT
47 is important for Aviation Safety. Favourable conditions for the occurrence of violent
48 turbulence or CAT, leading to low-level aircraft hazards, are passage of an active cold front;
49 preceding the time of thunderstorm; mountainous terrain; arced parts of jet-stream which
50 might be much more likely to contain turbulence than beeline jet-stream parts; convergence
51 area of polar and subtropical jet-stream; and the presence of vertical wind shear, horizontal
52 shear, convergence, deformation zone, strong thermal wind gradient, and steep lapse rate.
53 From the actual first flight, pilots are acquainted with mid-flight turbulence. In the 1940s, a
54 previously unidentified phenomenon was discovered as fighter aircraft reached the interface
55 between the troposphere and the stratosphere. This phenomenon was termed Clear Air
56 Turbulence (CAT) as antecedent encounters were experienced in cloudless regions.
57 Throughout the latter half of the 20th century, CAT attracted several organized research
58 efforts because the aircraft was designed to fly at significant heights and speeds. As an
59 outcome, our acquaintance with CAT has developed substantially. Between 1967–2010, “the
60 relative contributions of meteorological phenomena to weather-related aircraft accidents
61 shows that turbulence was associated with 66% of the cruise flight accidents and 56% of
62 accidents occurred during descent whereas, CAT accounted for 13% of accidents during
63 cruise flights, and 7% of accidents during descent” (Mazon et al. 2018). Studies concluded
64 that “the frequency of CAT would increase significantly in the next 50 years due to

65 strengthening of jet stream velocities” (Williams and Joshi 2013). “CAT is more frequent in
66 the tropopause region” (Dutton and Panofsky 1970) near the jet stream as “jet stream contains
67 about three times more CAT than the rest of the atmosphere” (Reiter 1963). About 60% of
68 CAT incidents are associated with the jet stream.

69 Two widely accepted mechanisms supporting CAT formation are Kelvin Helmholtz
70 Instability (KHI) and Mountain Waves. In most cases, KHI triggers CAT as it uses the
71 mechanical energy from the vertical wind shear. Thus, forecasting CAT is challenging
72 without understanding vertical shear (KHI). “Mountain waves are also a source of CAT.
73 There have been several cases where severe CAT was encountered in areas with no
74 significant KHI favourable conditions” (Hopkins 1977). Another CAT production mechanism
75 is Internal Gravity Wave (IGW). The excitation of the IGW also affects the manifestation and
76 strength of CAT from different sources and resonant non-linear relations amongst different
77 IGW modes and among IGW and KHI in turbulent layers. “Low values of Richardson
78 number (i.e., $Ri < 1$), discontinuity in lapse rate, significant cyclonic horizontal shear, and
79 large vertical velocity ($\sim 1\text{m/s}$) are triggering factors for CAT” (Venkatesh et al. 2014).

80 Forecasting and predicting CAT is a challenge for Meteorologists, as it is generally a
81 microscale phenomenon having a small temporal and spatial dimension, making it
82 challenging to attain consistent and inclusive observations. In the early 1960s, a basic jet-
83 stream turbulence model was developed in United Airlines by their meteorology department.
84 “Some advances in our knowledge of the global distribution of CAT along heavily traveled
85 airways have been derived from programs to collect PIREPs, like one conducted by ICAO in
86 the mid-1960s. A comprehensive global ‘climatology’ of large-scale and upper-level
87 conditions favourable for CAT was created using a numerical model to determine the
88 distribution of a globally averaged CAT index equal to the product of horizontal deformation
89 and VWS” (Ellrod et al. 2003). In the United Kingdom in the late 1970s, “turbulence data
90 from 4500 aircraft reports were compared with 11 co-located numerical parameters derived
91 from a coarse-resolution prediction model. It revealed that the best correlation was between
92 CAT and vertical and horizontal wind shears. Similar studies were completed in the United
93 States in the 1980s using higher-resolution numerical model data that showed CAT to be
94 highly correlated with horizontal deformation and scalar wind speed” (Ellrod et al. 2015).
95 “Two models were introduced for CAT prediction, i.e., eddy and wave motions models”
96 (Lester 1993). An ensemble model can be used to create a Global probabilistic turbulence
97 forecast. Most of the indicators used now are derived from the deterministic model. “Using
98 Met Office Global and Regional Ensemble Prediction System (MOGREPS), an ensemble

99 forecast can predict turbulence by the probabilistic indicator of wind shear” (Gill and
100 Buchanan 2013). Today, no warning system can ascertain CAT at archetypal flight altitudes.
101 CAT is a severe security issue for aircraft as there are no land or onboard detection gadgets.
102 Even the onboard weather radars are unsighted to CAT. “Doppler LIDAR has been under
103 research and development for many years to measure wind velocities and CAT detection.
104 Research indicates that moderate CAT could be detected at ranges of 5-8 km and up to 100 s
105 ahead of an aircraft” (ICAO 2005). “WRF Model is also used widely for turbulence
106 prediction and studies related to CAT” (Passner 2008).

107 It is essential to have an objective and meaningful approach to verify and improve the
108 forecasting of CAT. The aim of this study is mainly to understand the dynamic processes that
109 can trigger CAT events; to simulate a turbulence event of 19 April 2018 using the WRF
110 V4.0.3 Model; to correlate CAT parameters with the WRF Model derived parameters for the
111 selected CAT simulation; and to understand the reliability of turbulence indices, TI1 and TI2,
112 for predicting CAT. Section 2 describes the data, methodology, and model study.
113 Experimental design has been described in section 2.4, whereas results and discussion are
114 explained and analyzed in section 3.

115 **2 Methodology and Modelling Framework**

116 **2.1 Study Area**

117 There are two domains in the ratio 3:1 km resolution, as shown in Fig. 1. Domain 1 (D01)
118 with 6 km horizontal resolution covers the flight route from Amritsar to Delhi that
119 encountered the CAT event. Nested domain 2 (D02) with 2 km horizontal resolution is
120 focused on Amritsar's vicinity where the flight encountered the CAT event. On 19 April 2018,
121 Air India flight AI462 took off at 0923 UTC from Amritsar. While passing from FL80 to
122 FL210, light to moderate turbulence unexpectedly grew severe. The aircraft encountered
123 severe turbulence unexpectedly while climbing out from Amritsar between FL160 and
124 FL190. Once in severe turbulence, while passing through FL180, auto-pilot disengaged at
125 0929 UTC, and auto-thrust was also disengaged at 0930 UTC. Momentarily, the aircraft
126 climbed almost 600 feet above the cleared level (DGCA Report 2019).

127 **2.2 Data**

128 NCEP FNL Operational Global Analysis Data (NCEP/NWS/NOAA 2015) at $0.25^\circ \times 0.25^\circ$

129 resolution with 6-hourly intervals was used as initial and boundary conditions for the model
130 integration for 48 hours from 0000 UTC of 18 April 2018 to 0000 UTC of 20 April 2018.
131 Model is configured as 6 km for Parent Domain (D01) and 2 km for Child Domain (D02).
132 ECMWF ERA-Interim (ECMWF 2011), IMDAA (Rani et al. 2021), and NGFS (Prasad et al.
133 2016) data sets were also used in this study to analyze the CAT event. The specifications of
134 all the data utilized in this study are stated in Table 1.

135 **2.3 Methodology**

136 The WRF Model Version 4.0.3 is used to simulate the Turbulence or CAT for the given date.
137 The Advanced Research WRF (ARW) dynamic core in the model is used for this project as it
138 is suitable for use in various applications across scales ranging from meters to thousands of
139 kilometers. In this work, GrADS is used as a Post-Processing & Visualization tool. The
140 atmospheric reanalysis datasets can be used to identify the areas in which the onset of
141 hydrodynamics instability in the atmospheric flow occurred and is maintained, resulting in
142 CAT. Several indices exist to predict CAT, such as Ellrod's Index, Brown index, and Dutton
143 index. Ellrod's Indices (TI1 and TI2 given in Eq. (6) and (8), respectively) are calculated to
144 analyze the CAT event in this study as they are popular due to their performance,
145 computational speed, and easy implementation. "Studies indicate that Ellrod's Indices,
146 proposed by Ellrod and Knapp (1992), tend to perform better than others investigated"
147 (McCann 1993; Brown et al. 2000). "The TI1 and TI2 are in operational use at aviation
148 forecasting offices in several countries" (Ellrod and Knox 2010). "For example, TI1 is used at
149 Dutch KNMI (Royal Netherlands Meteorological Institute)" (Overeem 2002), "TI2 is used at
150 Swedish SMHI (Swedish Meteorological and Hydrological Institute)" (Bergman 2001), and
151 "Met Office in U. K. has implemented the indices as well" (Turp and Gill 2008).

152 **2.3.1 Deformation**

153 CAT occurrence chance will increase when deformation in the upper-level frontal zone due to
154 horizontal temperature gradient increases as deformation affects the horizontal temperature
155 gradient. Deformation includes D_{st} , deformation by stretching (s^{-1}), i.e., downwind,

$$D_{st} = \frac{\partial u}{\partial x} - \frac{\partial v}{\partial y} \quad (1)$$

156 and D_{sh} , deformation by horizontal shearing (s^{-1}), i.e., crosswind,

$$D_{sh} = \frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} \quad (2)$$

157 The Total Deformation (s^{-1}), $DEF = (D_{st}^2 + D_{sh}^2)^{1/2}$ (3)

158 where u and v are the wind components (m/s). “The relatively low values of the second
159 Ellrod’s index (TI2) are due to the lower contribution of deformation in areas of horizontally
160 homogeneous and relatively low curvature flow” (Spensberger and Spengler 2014).
161 Turbulence will be more substantial in the more sharply defined deformation zones.

162 2.3.2 Convergence

163 The CVG, convergence term (s^{-1}) included in TI2 is shown below (Eq. (4)).

$$CVG = -\left(\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y}\right) \quad (4)$$

164 The two indices, i.e., TI1 and TI2, show fewer differences in the geographical location of the
165 highest frequency during the summer season. This could be explained by the fact that the jet
166 stream is weaker during summer, although the frequencies of the higher values of TI2 are
167 significantly higher than the same for TI1. The added convergence term in TI2 can explain
168 this. “CVG term is typically much smaller than DEF, but in some cases, it can still contribute
169 significantly to CAT potential” (Kao and Sizoo 1966; Ellrod 1985).

170 2.3.3 Vertical Wind Shear (VWS)

171 The measurement of the VWS is often understood as the best indicator for the presence of
172 CAT, as it can occur in areas with strong vertical and horizontal wind shear. “The VWS is a
173 triggering mechanism for KHI, the primary mechanism for CAT formation” (Ellrod and
174 Knapp 1992). The phenomenon occurs when a sufficiently large VWS within a stable layer
175 produces breaking waves that lead to CAT. “Operational meteorologists consider Shear
176 values of at least 6 kt / 1000 ft $\left(\left[\frac{3 \text{ ms}^{-1}}{1000 \text{ ft}}\right] \text{ or } 9.7 \times 10^{-3} s^{-1}\right)$ as the threshold for significant
177 (moderate or greater) CAT” (Lee et al. 1984). VWS uses results of the layer difference in u
178 and v wind components from forecast data of the model as calculated in TI.

$$VWS = \sqrt{\left(\frac{\partial u}{\partial z}\right)^2 + \left(\frac{\partial v}{\partial z}\right)^2} \quad (5)$$

179 where ∂z is the thickness between the pressure levels. VWS is calculated for a layer amid two
180 pressure levels, not a discrete pressure level. “Low-level cold advection under a ridge tends to

181 increase the VWS, thereby increasing the probability of moderate to severe CAT” (Hopkins
 182 1977). CAT is classified into various categories, as shown in Table 2.

183 **2.3.4 CAT Index (TI1 and TI2)**

184 It was discovered by Mancuso and Endlich (1966) that “VWS and deformation products gave
 185 the best correlation with CAT generation”. Therefore, the turbulence index TI1 was
 186 simplified and defined by Ellrod and Knapp (1992) as Eq. (6) and (7),

$$187 \quad \text{TI1} = \text{VWS} \times \text{DEF} \quad (6)$$

188 From Eq. (1), (2), (3), and (5),

$$189 \quad \Rightarrow \text{TI1} = \left(\left(\frac{\partial u}{\partial z} \right)^2 + \left(\frac{\partial v}{\partial z} \right)^2 \right)^{1/2} \times \left(\left(\frac{\partial u}{\partial x} - \frac{\partial v}{\partial y} \right)^2 + \left(\frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} \right)^2 \right)^{1/2} \quad (7)$$

190 Ellrod and Knapp (1992) also “defined a second turbulence index (TI2) which included the
 191 convergence term”, as shown in Eq. (8) and (9),

$$192 \quad \text{TI2} = \text{VWS} \times [\text{DEF} + \text{CVG}] \quad (8)$$

193 From Eq. (1), (2), (3), (4) and (5),

$$194 \quad \Rightarrow \text{TI2} = \left(\left(\frac{\partial u}{\partial z} \right)^2 + \left(\frac{\partial v}{\partial z} \right)^2 \right)^{1/2} \times \left[\left(\left(\frac{\partial u}{\partial x} - \frac{\partial v}{\partial y} \right)^2 + \left(\frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} \right)^2 \right)^{1/2} - \left(\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y} \right) \right] \quad (9)$$

195 “TI1 is the most skillful and widely used CAT indicator in operational forecasts”
 196 (Sharman et al. 2006; Kim et al. 2011; Gill 2012). “It was found that the two indices (TI1 and
 197 TI2) give the best performance compared to other indices” (Overeem 2002). Though they do
 198 miscalculate CAT areas, the indices are still helpful. WAFC Washington uses TI1 for
 199 forecasting shear-induced turbulence, and the AFGWC in Nebraska uses TI2. “For the lower
 200 threshold, both TI1 and TI2 perform well on detecting CAT. Both indices overestimate CAT
 201 at a lower threshold but cover most turbulence reports. Based on the hits at the lower
 202 threshold, TI1 score more hits than TI2. However, based on false alarm rates, TI2 performs
 203 slightly better than TI1. For the higher threshold, the indices underestimate CAT occurrences.
 204 TI2 values are generally larger than TI1, which is probably the explanation for the higher
 205 false alarm rate.” (Williams 2017). Onset thresholds for each CAT intensity category are
 206 mentioned in Table 3. Values listed in Table 3 may differ from those computed in other
 207 studies as thresholds depend on the atmospheric model's grid resolution.

208 **2.3.5 Bulk Richardson Number Shear (BRNSHR)**

209 “Bulk Richardson Number Shear (BRNSHR) is used to quantify the Vertical Wind
 210 Shear (VWS)” (Moncrieff and Green 1972), such that,

$$BRNSHR(m^2/s^2) = 0.5(\underline{u}^2 + \underline{v}^2) \quad (10)$$

211 where \underline{u} and \underline{v} are “zonal and meridional wind components of the difference between the
 212 density-weighted mean winds over the lowest 6000 m (\approx 500 hPa) and the lowest 500 m
 213 (\approx 950 hPa) above ground level” (Droegemeier et al. 1993). “Mesomodel output BRNSHR
 214 values between 40-100 m^2/s^2 are indicative of a greater likelihood of tornadic supercell
 215 thunderstorms” (Stensrud et al. 1997). The threshold value of BRNSHR for supercell
 216 thunderstorms development is 40 m^2/s^2 .

217 **2.4 Experimental Design**

218 This study can be classified as a climatological analysis and a regional study of a CAT event
 219 in India. WRF Model V 4.0.3 is used for all simulations and configured with 6 km and 2 km
 220 grid space domains. WRF model was integrated for two days, from 0000 UTC of 18 April
 221 2018 to 0000 UTC of 20 April 2018. The model was set to a time step of 36 seconds with
 222 1 hour (60 minutes) history intervals for the 1st domain (D01) and 15 minutes history
 223 intervals for the 2nd domain (D02). The 1st domain (D01) had 110 x 110 grid points, and the
 224 2nd domain (D02) had 109 x 109 grid points in the west-east and north-south directions. Both
 225 the domains were set with 33 vertical levels. Various literature on WRF modelling and
 226 simulation of CAT were studied at the beginning of this project (Wasson 2021). Thus,
 227 considering results and suggestions from those works of literature, the physics
 228 parameterization schemes were given weightage, and accordingly, each scheme used in this
 229 study was selected. Model physics parameterization schemes used in this study to simulate
 230 turbulence are given in Table 4. The most important scheme for the simulation of CAT is the
 231 Planetary Boundary Layer (PBL) Scheme. When grid size is more than or equal to 1 km, we
 232 rely on PBL schemes to handle the upper air turbulence or CAT to its vertical diffusion. So,
 233 different PBL schemes may give different results because they handle vertical diffusion in
 234 different ways. In this study, Yonsei University Scheme is used as a PBL scheme.

235 **3 Results and Discussion**

236 This section discusses the results obtained from the simulations and their comparison with the
 237 corresponding observations. The weather incident occurred on 19 April 2018 during 0900
 238 UTC to 1200 UTC, and the model has been integrated for 48 hours, from 0000 UTC of 18
 239 April 2018 to 0000 UTC of 20 April 2018. The model-designed parameters are Vertical

240 Velocity, Horizontal Velocity, Geopotential Thickness, Vertical Wind Shear, Clear Air
241 Turbulence (CAT) Index, and Bulk Richardson Number Shear (BRNSHR).

242 **3.1 Vertical Velocity**

243 Vertical Velocity is an essential parameter for the occurrence of turbulence. Fig. 2 and 3
244 illustrate the time series of Vertical Velocity (m/s). Longitude varies from 73.4° E to 75.6° E
245 as the turbulence was encountered for a small region. Vertical Velocity is plotted for latitude
246 31.3149° N at 450 hPa level in Fig. 2 and different pressure levels (from 400 hPa to 550 hPa
247 with intervals of 50 hPa) in Fig. 3 at 1000 UTC on 19 April 2018. The blue and red color
248 region indicates downward (-2 m/s) and upward (6-8 m/s) velocity, respectively, at 450 hPa
249 level (Fig. 2 and 3(c)). So, this is a severe to extreme turbulence case (refer to Table 2). In
250 reality, turbulence was encountered between 0900 UTC and 1000 UTC. The model simulated
251 strong Vertical Velocity at 1000 UTC at 74.97° E (Fig. 2 and 3), closer to reality.

252 Fig. 4 illustrates the spatial distribution of Vertical Velocity where the aircraft
253 encountered turbulence. Plots (A) to (D) in (D01) Parent Domain (6 km) and plots (a) to (d)
254 in (D02) Child Domain (2 km) represent Vertical Velocity at different levels (from 400 hPa
255 to 550 hPa with 50 hPa interval), respectively, at 1000 UTC on 19 April 2018 as simulated by
256 the model. It shows substantial and strong Vertical Velocity near the 'X' mark (i.e., 74.97° E
257 and 31.3149° N). The negative values indicate accelerating updraft, and the positive values
258 indicate descending downdraft. At 450 hPa ((B) and (b)), both upward (6-8 m/s) and
259 downward (-2 m/s) velocities are seen very close to each other. From Fig. 1, it is clear that
260 the location near 'X' is where CAT was encountered. The dashed circle marks this area.

261 Fig. 5 illustrates the vertical profile of Vertical Velocity over the accident location
262 with a longitude of 74.97° E and latitude of 31.3149° N at 1000 UTC on 19 April 2018.
263 Vertical Velocity varies with changes in pressure level (altitude). Dash patches represent the
264 region between 400 hPa and 550 hPa where the model simulated strong Vertical Velocity (6-
265 8 m/s). The left side of the straight line drawn on zero represents the downward velocity, and
266 the right side represents the upward velocity. There is a sudden increase in Vertical Velocity
267 by at least 4-5 m/s, and the speed is rising to 7 m/s or more (from Fig. 3, 4, and 5).

268 **3.2 Horizontal Velocity**

269 Fig. 6 illustrates a comparison between the Horizontal Velocity (m/s) vertical profile over the
270 accident location as simulated by the (a) Model at 1000 UTC, (b) ERA-Interim data at

271 1200 UTC, (c) IMDAA data at 0900 UTC, and (d) NGFS data at 1200 UTC. Horizontal
272 Velocity varies with changes in pressure levels (altitude). The dash patches represent the
273 region between 100 hPa and 350 hPa where the model and other data sets simulated strong
274 horizontal velocity (60-65 m/s). There is a sudden increase in Horizontal Velocity by at least
275 40-50 m/s, and speed is rising to 62 m/s or more. It could lead to the formation of CAT.

276 **3.3 Geopotential Thickness**

277 Fig. 7 illustrates a comparison between the Geopotential Thickness (m) on 19 April 2018
278 between 400 hPa and 650 hPa, as simulated by the ((A) and (a)) Model at 1000 UTC and ((B)
279 and (b)) ERA-Interim data at 1200 UTC. Plots (A) and (B) represent Geopotential Thickness
280 in (D01) Parent Domain (6 km), whereas plots (a) and (b) represent Geopotential Thickness
281 in (D02) Child Domain (2 km). Here, 'X' indicates the position where the model predicted
282 strong Vertical Velocity near the accident location (Fig. 4). The area enclosed by the dashed
283 circle is where the model and ERA-Interim data sets simulated a strong Geopotential
284 Thickness (more than 36.5 km). This could lead to CAT. Patches in the model simulation are
285 not visible in ERA-Interim data simulation, as CAT is generally a microscale phenomenon.
286 Thus, the Geopotential Thickness gradient is visible in the plot (a) of (D02) Child Domain.

287 **3.4 Vertical Wind Shear (VWS)**

288 Fig. 8 illustrates a comparison between the Vertical Wind Shear (m/s per 1000 ft) between
289 200 hPa and 850 hPa for both the domains (D01 and D02) on 19 April 2018 as simulated by
290 the ((A) and (a)) Model at 1000 UTC, ((B) and (b)) ERA-Interim data at 1200 UTC, ((C) and
291 (c)) IMDAA data at 0900 UTC and ((D) and (d)) NGFS data at 1200 UTC. Plots (A), (B),
292 (C), and (D) represent VWS between 200 hPa and 850 hPa in (D01) Parent Domain (6 km),
293 whereas plots (a), (b), (c), and (d) are representing VWS between 200 hPa and 850 hPa in
294 (D02) Child Domain (2 km).

295 Here, 'X' indicates the position where the model predicted strong vertical velocity
296 (Fig. 4) and strong geopotential thickness gradient (Fig. 7) near the accident location. The
297 area enclosed by the dashed circle is where the model and other data sets simulated a strong
298 Wind Shear gradient (more than 5.5 m/s per 1000 ft). So, this is a severe turbulence case
299 (refer to Table 2). The patches in the model simulation are not visible in other data
300 simulations as CAT is generally a microscale phenomenon that is harder to predict and
301 requires high-resolution data sets. Thus, the Wind Shear gradient is more clearly visible in

302 plot (a) of (D02) Child Domain (2 km). It could lead to CAT.

303 **3.5 Clear Air Turbulence (CAT) Index**

304 The CAT Index indicates the most probable region of CAT. Fig. 9 illustrates a comparison
305 between CAT Index TI1 (s^{-2}) and CAT Index TI2 (s^{-2}) at different pressure levels (from
306 300 hPa to 550 hPa with 50 hPa interval) in (D02) Child Domain (2 km) at 1000 UTC on
307 19 April 2018, as simulated by the model. Plots (A) to (F) represent CAT Index TI1 (s^{-2})
308 whereas plots (a) to (f) represent CAT Index TI2 (s^{-2}). Here, 'X' indicates the position
309 where the model predicted strong vertical velocity (Fig. 4), strong geopotential thickness
310 gradient (Fig. 7), and strong wind shear gradient (Fig. 8) near the accident location. The area
311 enclosed by the dashed circle is where the model simulated strong CAT Index intensities at
312 1000 UTC for TI1 ($480.404 \times 10^{-9} s^{-2}$) and TI2 ($503.783 \times 10^{-9} s^{-2}$) at 450 hPa level
313 ((D) and (d), respectively). Model simulated variations in CAT Index with change in
314 different pressure levels. The dissipation of the CAT is visible as we go higher in the
315 atmosphere above 550 hPa. In Fig. 9, the white-colored region represents the region with No
316 CAT intensity, the blue-colored region represents the region with Light CAT intensity, the
317 green-colored region represents the region with Light to Moderate CAT intensity, the
318 yellow-colored region represents the region with Moderate CAT intensity, the orange-
319 colored region represents the region with Moderate to Severe CAT intensity, and the red-
320 colored region represents the region with Severe CAT intensity (refer to Table 3).

321 Fig. 10 illustrates the comparison of Time-Series between CAT Index TI1 (s^{-2}) and
322 TI2 (s^{-2}) for 24 hrs, i.e., from 0000 UTC on 19 April 2018 to 0000 UTC on 20 April 2018,
323 at 450 hPa near the accident location as simulated by the model. Here, the red-colored plot
324 with 'o' markings represents the time-series of CAT Index TI2 (s^{-2}) and the blue-colored
325 plot with 'x' markings represents the time-series of CAT Index TI1 (s^{-2}). Model simulated
326 variations in CAT Index with change in time. Model simulations show that CAT evolved or
327 generated at approximately 0900 UTC near 'X' and reached the peak intensity at 1000 UTC;
328 after that, it started dissipating at around 1200 UTC. The time-series of the CAT Index
329 shows that at 1000 UTC, the CAT index is high, implying more chances of encountering
330 CAT. In Fig. 10, it is visible that the model simulated strong CAT Index intensities at
331 1000 UTC for both TI1 ($480.404 \times 10^{-9} s^{-2}$) and TI2 ($503.783 \times 10^{-9} s^{-2}$) at 450 hPa
332 level. So, this is a severe CAT case (refer to Table 3) that can lead to dangerous conditions.

333 It is observed that TI1 tends to overpredict, but the model simulation for TI2 is very
334 similar to TI1. TI1 and TI2 perform similarly, but TI1 performs slightly better than TI2.
335 Thus, (from Fig. 9, 10, and Table 3) for the lower threshold, both TI1 and TI2 perform
336 satisfactorily for CAT detection. It is observed that both TI1 and TI2 sometimes overestimate
337 CAT at a lower threshold, due to which they might have a high rate of false alarm; however,
338 this might cover the maximum region of turbulence. For the higher threshold, it is observed
339 that both TI1 and TI2 might underestimate CAT occurrences. The TI2 value is more than the
340 TI1 value, explaining the high rate of false alarms.

341 **3.6 Bulk Richardson Number Shear (BRNSHR)**

342 “Bulk Richardson number shear (BRNSHR) is used to quantify the VWS” (Moncrieff and
343 Green 1972). Fig. 11 illustrates BRNSHR (m^2/s^2) on 19 April 2018, showing variations
344 with time (from 0900 UTC to 1200 UTC with 1 hour time interval), as simulated by the
345 model. Plots (A) to (D) and plots (a) to (d) represent BRNSHR variations with time in (D01)
346 Parent Domain (6 km) and (D02) Child Domain (2 km), respectively. Here, ‘X’ indicates the
347 position where the model predicted strong vertical velocity (Fig. 4), strong geopotential
348 thickness gradient (Fig. 7), strong wind shear gradient (Fig. 8), and strong CAT Index
349 intensity (Fig. 9) near the accident location. The area enclosed by the dashed circle is the
350 area where the WRF model simulated strong BRNSHR ($235.047 \text{ m}^2/\text{s}^2$). High values of
351 BRNSHR ($235.047 \text{ m}^2/\text{s}^2$) are simulated for the accident day around 1000 UTC at
352 450 hPa level ((B) and (b)), indicating sheared environment. So, this could show the
353 probability of encountering turbulence that can lead to dangerous conditions.

354 Fig. 12 illustrates the Time-Series of BRNSHR (m^2/s^2) for 12 hours, i.e., from 0600
355 UTC to 1800 UTC on 19 April 2018, near the accident location as simulated by the model.
356 Model simulated variations in BRNSHR with change in time. The model simulations show
357 that BRNSHR evolved or generated at approximately 0900 UTC near ‘X’ and reached the
358 peak intensity at 1000 UTC; after that, it started dissipating at around 1200 UTC. The time
359 series of BRNSHR shows that at 1000 UTC, BRNSHR is high, which implies more chances
360 of encountering turbulence. So, this could be a severe turbulence case that can lead to
361 dangerous conditions. Most of the BRNSHR values are more than the threshold value (40
362 m^2/s^2) for the development of supercell thunderstorms. The highest value of the BRNSHR
363 simulated by the model is approximately $235.047 \text{ m}^2/\text{s}^2$ which is usually large enough to
364 generate spinning storms.

365 4 Summary and Conclusion

366 At high altitudes, aircraft may encounter turbulence unexpectedly without any significant
367 cloudiness. We have already discussed how difficult is the prediction of CAT. The role of
368 CAT for various processes in the atmosphere is still ambiguous. To know the unknown about
369 CAT, modification of tools is essential for simulating turbulence. This study is focused on
370 understanding and investigating favourable conditions for the occurrence of CAT. The results
371 show that the WRF model simulated the CAT incident between Amritsar and Delhi, as
372 reported by the Air India flight AI462 on 19 April 2018. Although limitations exist as a
373 model cannot predict exact atmospheric conditions, it can predict close to the case; this work
374 has proven that WRF Model V 4.0.3 used for this study is sensitive to simulate turbulence.
375 NCEP FNL Operational Global Analysis Data at $0.25^\circ \times 0.25^\circ$ resolution with 6-hourly
376 intervals was used as initial and boundary conditions for the model integration for 48 hours
377 from 0000 UTC of 18 April 2018 to 0000 UTC of 20 April 2018. This study also shows that
378 the Rapid Radiative Transfer Model (RRTM) Scheme, Yonsei University Scheme, WSM 3-
379 Class Simple Ice Scheme, Kain-Fritsch (New Eta) Scheme, and Revised MM5 Monin-
380 Obukhov Scheme can be used for simulating the atmospheric conditions during a CAT event
381 and predicting the turbulence over a region. This study shows that Vertical Velocity,
382 Geopotential Thickness gradient, Wind Shear gradient, and BRNSHR are correlated with
383 CAT. The TI1 and TI2 are based on deformation and might not consider additional
384 mechanisms that might produce CAT, e.g., mountain waves. Some more studies must be
385 done to capture the climatology of other mechanisms. The main conclusions of this study are:

- 386 • This is a severe CAT case, and the results show that the WRF model reasonably predicts
387 turbulence. The most important scheme for the simulation of CAT is the PBL Scheme.
388 In this study, Yonsei University Scheme is used as a PBL scheme.
- 389 • Model simulated both upward (6-8 m/s) and downward (-2 m/s) velocity, very close to
390 each other between 400 hPa and 550 hPa levels at 1000 UTC near the accident location.
391 So, this is a severe to extreme turbulence case (refer to Table 2).
- 392 • Model and other data sets simulated strong horizontal velocity (60-65 m/s) at the same
393 region between 100 hPa and 350 hPa levels near the accident location.
- 394 • Model and other data set simulated strong Geopotential Thickness (more than 36.5 km)
395 and strong Wind Shear gradient (more than 5.5 m/s per 1000 ft) near the 'X' location at
396 1000 UTC. So, this is a severe turbulence case (refer to Table 2).

397 • Model simulated variations in CAT Index with change in time and pressure levels. CAT
398 dissipates as we go higher in the atmosphere above 550 hPa.

399 • Model simulations show that CAT evolved or generated at approximately 0900 UTC
400 near ‘X’ location and reached the peak intensity at 1000 UTC; after that, it started
401 dissipating at around 1200 UTC.

402 • The model simulated strong CAT Index intensities for TI1 ($480.404 \times 10^{-9} \text{ s}^{-2}$) and
403 TI2 ($503.783 \times 10^{-9} \text{ s}^{-2}$) at 450 hPa level at 1000 UTC (from Fig. 9, 10, and Table 3),
404 at the exact location where the model predicted strong vertical velocity (Fig. 4), strong
405 geopotential thickness gradient (Fig. 7), and strong wind shear gradient (Fig. 8) near ‘X’
406 location, closer to reality. So, this is a severe CAT case (refer to Table 3) that can lead to
407 dangerous conditions.

408 • TI1 and TI2 perform similarly, but TI1 performs slightly better than TI2.

409 • The model also simulated variations in BRNSHR with changes in time. Model
410 simulations show that BRNSHR evolved or generated at approximately 0900 UTC near
411 ‘X’ and reached the peak intensity at 1000 UTC; after that, it started dissipating at
412 around 1200 UTC. High values of BRNSHR ($235.047 \text{ m}^2/\text{s}^2$) are simulated at around
413 1000 UTC for the accident day at 450 hPa level (Fig. 11 and 12), indicating sheared
414 environment. Thus, this could show the probability of encountering turbulence as the
415 high intensity of BRNSHR indicates more chances of encountering turbulence.

416 Even though there are differences in reality and model prediction, the model can still predict
417 turbulence. We should also know how significantly other schemes can predict turbulence by
418 finding different parameterization scheme combinations to predict turbulence more
419 accurately. Only one case is discussed in this study, but many other cases are being reported
420 worldwide, and only after studying a more significant number of cases can we tell the exact
421 behaviour of CAT. These all will do in the future as a continuation of this work.

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