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# Development of Probabilistic Convective Weather Forecast Threshold Parameter for Flight Routing Decisions

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Abstract

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This paper presents a method for determining a threshold value of probabilistic convective weather forecast data. The threshold represents the bounding severe weather forecast probability value that most aircraft are observed to avoid. Given a probabilistic prediction of weather, this value can be used by dispatchers for flight planning; and by air traffic managers to reroute streams of aircraft around convective cells. Both, the intensity and ceiling of the forecasted weather were synchronized with air traffic data in a simulation to derive the probability threshold. The main contribution of this paper is to provide a method to compute probability threshold parameters using an experimental 6-hour probabilistic convective weather forecast product. Air traffic and weather data for a four-month period during the summer of 2007 were used to compute the parameters for the continental United States. Threshold values for each of the 20 Air Route Traffic Control Centers were also computed. Additional details are presented for seven high-altitude Sectors in the Ft. Worth Center. The results are shown for different altitudes, times of day, aircraft types and airspace users.

Airline dispatchers and traffic managers are involved in flight operations to avoid specific areas of convective activity. A secondary contribution of this paper is to describe a simple approach to utilize the threshold parameter for flight routing decisions in Ft. Worth Center during convective weather events. This approach is similar to national severe weather routes currently used by the FAA, except in a Center-based, local event. The paper presents results for contrasting the nationwide reroutes and the local strategy. The results suggest that such an analysis capability could save fuel and reduce air traffic delays.

## 60 **1. Introduction**

61           Based on air traffic delay results from the Federal Aviation Administration's (FAA)  
62 Operations Network (OPSNET) data, more than 70% of the National Airspace System (NAS)  
63 reportable delays are attributed to convective weather. Furthermore, a study by Sridhar, et al.  
64 (2007) indicated that a small number of Air Route Traffic Control Centers (ARTCCs or Centers)  
65 experience a majority of the weather impact. The current deterministic weather prediction in  
66 tactical (next 2 hours) timeframe is used for air traffic operations. Current operational strategies  
67 to route air traffic around convective weather systems, such as the FAA's Severe Weather  
68 Avoidance Plan (or Playbook Routes), use pre-established routes to ensure predictable and safe  
69 circumvention of convective weather zones. These national strategies may force aircraft to take  
70 large deviations, even if the aircraft were unlikely to encounter convective weather, and usually  
71 impose additional workload on Centers not directly affected by the convective weather system. It  
72 is widely accepted that the state of weather forecasting in strategic (2-8 hours) time frame needs  
73 improvement for longer-term flight planning (Clifford 2003, Fahey, et al. 2006). Also, some  
74 work has been done to tactically reroute individual aircraft around weather cells (Grabbe, et al.  
75 2008 and Sridhar, et. al. 2002), little work has been done to strategically route flights for the  
76 Center level weather events.

77           In the absence of improved forecast, considerable research is being conducted to develop  
78 improved weather products and to determine how to make better use of probabilistic data for  
79 improved flight planning. The NCAR has recently released a National Weather Forecast Product  
80 (NCWF) that provides 1, 2, 3, 4, 5, and 6 hour forecasted probabilistic weather contours. An  
81 overview of the weather data needs and benefits to various participants in Air Traffic  
82 Management (ATM) along with available products can be found in Fahey, et al. (2006). Past

83 research has focused on the concept of operations for strategic traffic flow management,  
84 including how weather data can be integrated for efficient traffic management initiatives  
85 (Hoffman, et al. 2004, Nilim and El Ghaoui, 2004, and Song, et al. (2008)). A model of  
86 predicting likelihood of flight deviation and pilot behavior around a convective storm is available  
87 in DeLaura and Evans (2006). Chan, et al. (2007). More recently Matthew and DeLaura (2010),  
88 validate that pilots deviate when the avoidance model predicts 80% or higher likelihood of  
89 deviation for the analyzed altitudes. Weygandt and Benjamin (2004), and Megenhardt, et al.  
90 (2004) present probabilistic weather forecast generation method, and identify a need for the  
91 relevance of these forecasts for aviation use. Steiner, et al. (2008) looked at probabilistic air  
92 traffic management decisions by considering ensemble (and consequently, probabilistic)  
93 forecasts for developing traffic flow evolution scenarios. Another study by Wanke and  
94 Greenbaum (2007) presents the probabilistic decision making for en route traffic management  
95 through Monte Carlo simulations. Other literature suggests the need for a probabilistic  
96 description of weather for strategic Traffic Flow Management applications (Mitchell, et al.,  
97 2006, Sheth, et al. 2007) and a better temporal resolution for ATM strategic planning (Chan  
98 2006). The studies conducted so far have used probabilistic weather as a model, because  
99 operational products, other than the broad-coverage CCFP, are not available. When strategic and  
100 operational probabilistic forecast products are available (e.g., Localized Aviation MOS (Model  
101 Output Statistics) Product (LAMP), Ghirardelli and Glahn (2010)), research will need to be  
102 conducted to assess the corresponding reduction in airspace capacity. A recent research by Klein  
103 (2008) has shown that a probability threshold value is required for the airspace capacity  
104 estimation. Then, the threshold value can be used for efficient flight routing, especially, for the  
105 Center level weather events.

106           The current study presents a method for using probabilistic convective weather forecasts  
107 for strategic air traffic flow management decisions. First, the experimental 6-hour National  
108 Convective Weather Forecast (NCWF) probabilistic data product was chosen and integrated in  
109 an ATM simulation environment. Flight tracks were superimposed on NCWF probability of  
110 level 3 or higher Vertically Integrated Liquid (VIL) contours for each of the forecasts. The  
111 aircraft deviations around the actual severe weather and the forecasted probabilities were noted.  
112 Based on the *maximum* value of probability each flight skirted around, the Probability Threshold  
113 Parameter (PTP) was derived. Once the PTP values were available, a method is suggested for  
114 using it in flight routing decisions. To accomplish this, alternate route strategies were analyzed  
115 for a scenario when severe weather closes the Bonham (BYP) arrival fix for the Dallas/Ft. Worth  
116 International (DFW) airport. Thus, first this paper presents an approach for generating the  
117 threshold parameter, and then a method to apply it for flight routing decisions in a local weather  
118 scenario.

119 The simulation environment used to synchronize convective weather forecasts and air traffic data  
120 is described in Section 2. The method to obtain the PTP for the entire National Airspace System,  
121 the 20 Centers in the Continental US, and values for seven high-altitude sectors of interest in  
122 ZFW is presented in Section 3. The need for using local rerouting is described and a suitable  
123 implementation is detailed in Section 4. Results for a specific weather scenario in ZFW are also  
124 displayed in that Section. The paper ends with concluding remarks in Section 5.

## 125 **2. Integration of weather and air traffic data**

126           In order to study the impact of convective weather on air traffic, a simulation with  
127 integrated traffic and weather information is needed. The Future ATM Concepts Evaluation Tool  
128 (FACET) (Bilimoria, et al. 2001) provides that capability. FACET is a simulation and modeling

129 environment developed to explore advanced ATM concepts. It handles traffic information at  
130 various levels in the NAS, from Centers and the sub-regions of Sectors, to the capability of  
131 modeling and assessing individual aircraft trajectories. FACET can be run in playback mode to  
132 understand how the air traffic evolved on a particular day by replaying recorded data. FACET  
133 processes the FAA's Enhanced Traffic Management System (ETMS) air traffic data and various  
134 convective weather products, such as the Corridor Integrated Weather System (CIWS),  
135 Collaborative Convective Forecast Product (CCFP), National Convective Weather Forecast  
136 (NCWF), and Next Generation Radar (NEXRAD). Integration of newer convective weather  
137 products like Collaborative Storm Prediction for Aviation (CoSPA) (Wolfson, et al. 2008) and  
138 Localized Aviation MOS (Model Output Statistics) Program (LAMP) (Charba and Samplansky,  
139 2009) are also available within FACET. The integrated information can be used for visualizing  
140 the effects of weather in real time, as well as for planning of flights around forecasted weather.

141         The NCWF-6 data provide one-, two-, three-, four-, five- and six-hour weather forecasts  
142 of VIL level 3 or higher with a continuous probability distribution of severe weather every 15  
143 minutes. Figure 1 shows a snapshot of the synchronized air traffic and convective weather data  
144 displayed in FACET for 5 pm Central Daylight Time (CDT) or 22:00 Coordinated Universal  
145 Time (UTC) on July 10, 2007. The Ft. Worth (ZFW) ARTCC (closed polygon in the center, in  
146 white) is shown with a number of important fixes (circled) in the region. These are: Tulsa (TUL)  
147 and Will Rogers (IRW) in the north, Monroe (MLU) in the east, Waco (ACT) in the south, Wink  
148 (INK) in the west, etc.) along with four DFW arrival fixes of Cedar Creek (CQY) (southeast),  
149 Glen Rose (JEN) (southwest), Bowie (UKW) (northwest) and Bonham (BYP), hidden under  
150 weather, in the northeast. The one-hour NCWF-6 forecast data published at 4 pm CDT are  
151 shown as filled polygons. The color for weather forecast data is continuously varying, and the

152 probability of convective weather varies from 25% (cyan) on the periphery to 100% (dark red) at  
153 the center. Convective weather observations from NEXRAD are shown as unfilled contours of  
154 VIL level 3, 4, 5, and 6 in yellow, orange, red and dark brown, respectively. The aircraft arriving  
155 at and departing from DFW are shown as pink and cyan dots, respectively, along with their 20-  
156 minute track histories. It is observed from this figure that the track histories indicate flight  
157 deviation around weather, as seen just northeast of DFW airport.

### 158 **3. Probability Threshold Parameter**

159 For each aircraft track, the location and height of aircraft were used to find the  
160 corresponding grid cell in the forecast data. If an aircraft was flying below the forecasted  
161 probability ceiling, and its location was contained within a 10% or higher probability contour,  
162 then the aircraft was considered traversing through the probability field of the forecast data. For  
163 each aircraft's flight from origin to destination, the maximum probability value of VIL level 3 or  
164 higher was recorded. These data are recorded only if the probability forecast was valid at the  
165 time of aircraft track and only if at that location a storm ceiling (echo top) value was available  
166 (see Dupree, et al., 2006).

167 In Fig. 2a, a simulated flight on its FAA-filed flight plan on May 16, 2007 is shown with  
168 a yellow triangle for ACID1. As seen in the data block, the aircraft is flying from Norfolk, VA  
169 (ORF) to Indianapolis, IN (IND) at 32,000 ft (or Flight Level FL 320). It's traversing through a  
170 one-hour forecasted convective weather polygon. As can be seen from Fig. 2a, the ACID1 path  
171 traverses the predicted weather probability field between the 03:12 and 03:20 UTC (10:12 and  
172 10:20 pm CDT) times shown with white arrows. While crossing the weather contours, it  
173 traverses the NCWF-6 continuous probability distribution from 0% at 03:00 UTC (10:00 pm  
174 CDT) to about 70% at 03:15 UTC (10:15 pm CDT), as shown in Fig. 2b. Figure 2c shows how

175 the actual flown aircraft tracks completely avoid the weather, and so the probability traversal  
176 curve would have all zero values. It should be noted that actual severe weather on that day  
177 closely represented the forecasted weather, as shown in Fig. 2d. The probability traversal profile,  
178 like the one in Fig. 2b for the simulated flight, is created for all actual flights to study flight  
179 deviation behavior.

180 The *maximum* probability value crossed by each flight is recorded and binned in a reverse  
181 cumulative histogram ranging from 100% to 10% in 1% decrements. The 80<sup>th</sup> percentile number  
182 of this histogram, similar to the one proposed by Chan, et al., 2007 and DeLaura et al. (2009), is  
183 then used to determine the Probability Threshold Parameter (PTP) value. Since the PTP value is  
184 clearly avoided by a large number of aircraft, it is used as the weather probability value to avoid  
185 for flight routing decisions.

186 Figure 3a shows the scatter plot of the probabilities with the aircraft altitudes using  
187 ETMS data for one instant of time (08:12 am CDT) on May 16, 2007 and a one-hour forecast. In  
188 the scatter plot with a total of 48 flights plotted, there are 8 aircraft with values above 35%  
189 probability. It was important to analyze if the aircraft were really traversing through 40% to 65%  
190 probability values, because they could encounter significant convective activity. Analyzing their  
191 tracks, it was found that six of these eight aircraft were either transitioning (climbing or  
192 descending) aircraft or intruding a higher probability contour for one time instant. This may also  
193 be the situation when aircraft venture into the severe weather region or could be airline-  
194 designated pathfinder missions. It should be noted that the current analysis might show aircraft in  
195 higher probability regions due to forecast location error, intensity inaccuracies and flight track  
196 data errors. Figures 3b, c, d, and e show the altitude versus maximum probability data  
197 accumulated for one- through four-hour forecast valid-time instances for May 16 through 22,

198 2007. In this analysis, inconsequential low probability values (below 10%) were ignored, hence,  
199 the blank region in Fig. 3a through 3e, to the left of 10%. As can be seen from Figs. 3b through  
200 3e, the maximum observed probabilities for level 3 or higher convection decrease with time  
201 forecast horizon. A vertical line shows this at 99%, 83%, 58% and 43% in the one-, two-, three-,  
202 and four-hour data sets. This reduction in maximum probabilities is a result of the blending  
203 process used in the generation of these forecasts, described in Germann and Zawadzki (2000),  
204 Weygandt, et al. (2004), Megenhardt, et al. (2004), and Pinto, et al. (2008). The maximum  
205 observed probabilities for the five- and six-hour forecasts were below 30%, and were discarded.  
206 Even the three- and four-hour values were lower fidelity. Therefore, for the rest of this paper,  
207 only one- and two-hour results are presented for the threshold value computation.

#### 208 *a. Probability Threshold Parameter for the NAS*

209 The reverse cumulative histograms of number of aircraft at different altitudes traversing  
210 through the weather probability field are shown in Fig. 4 for (a) one- and (b) two-hour forecasts.  
211 For each of the curves going from ground level up to FL 400, it was observed that for a one-hour  
212 forecast, the 80<sup>th</sup> percentile value resides at about 33% (Fig. 4a). The colored vertical lines  
213 corresponding to various 10,000 ft blocks of altitude demonstrate this. The corresponding value  
214 for two-hour forecasts was about 23% (See Fig. 4b). For the purpose of this research, the 80<sup>th</sup>  
215 percentile value was chosen as the Probability Threshold Parameter (PTP). Aircraft are generally  
216 observed to go around probability values higher than the PTP. The flow management and flight  
217 planning decision-makers can use this value of PTP to generally avoid regions of forecasted  
218 severe weather.

219 Further analysis of the data provided weather traversal characteristics as a function of  
220 airlines and aircraft types. These results are presented in Fig. 4c and 4d. It should be noted that

221 these probabilistic weather data are only used in this post-processing analysis and were not  
222 available to operators. From one-hour data presented in Fig. 4a, the top four aircraft operator  
223 occurrences are presented in Fig. 4c. From the same data set, the top four aircraft type  
224 occurrences are shown in Fig. 4d. The four most frequently found aircraft types are the Boeing  
225 B73x, the Airbus A31x, the McDonnell-Douglas MD8x, and the Canadair Regional Jet CRJx.  
226 All considered aircraft types are observed to avoid flying beyond about 35% probability (the 80<sup>th</sup>  
227 percentile value). Similarly, as seen from Fig. 4c, major airlines appear to deviate beyond the  
228 ~35% probability value. Therefore, the NAS PTP value was concluded to be 35% for one-hour  
229 forecasts and 25% for two-hour forecasts.

#### 230 *b. Center-based Probability Threshold Parameter*

231 In this study, the PTP value was derived for each of the 20 NAS Centers as well. The  
232 purpose of evaluating the PTP value for each Center was to identify if there was a difference  
233 based on Centers. Figures 5a and 5b show the behavior of aircraft traversal for each of the 20  
234 Centers for the one- and two-hour forecasts. These data were recorded for all aircraft flying  
235 between 10,000 and 40,000 ft. It is seen from the one-hour plot on left that Ft. Worth (ZFW),  
236 Houston (ZHU), Atlanta (ZTL), Jacksonville (ZJX), and Miami (ZMA) Centers (all five  
237 neighbors in the southeastern part of the US) show large number of aircraft traversing through  
238 higher probability values. It is also seen from Fig. 5a that there are three bands within which the  
239 data can be classified. The first one consists of those five southeast Centers, ZFW, ZHU, ZTL,  
240 ZJX and ZMA, with larger than 40,000 aircraft crossing the 10% intensity contours, above the  
241 upper brown bar shown on the y-axis. The third consists of less than 10,000 aircraft crossing the  
242 10% intensity contours, below the lower brown bar. These are the 4 western Centers, Los  
243 Angeles (ZLA), Oakland (ZOA), Seattle (ZSE) and Salt Lake (ZLC), where there's less

244 convective activity generally. The middle band between the two brown bars consists of the 11  
245 remaining Centers showing between 10,000 and 40,000 aircraft. From Fig. 5b for the two-hour  
246 forecasts, similar banded behavior is observed, with the same Center members, but the middle  
247 band has between 20,000 and 70,000 aircraft. As noted earlier, the probability threshold values  
248 decrease (due to increased uncertainty) with increase in forecast time, which explains the curves  
249 steepening to the left. The computed PTP values for the one-hour forecasts were as follows: the  
250 minimum value was 18% (from the lower band Centers), the maximum value was 33% (from the  
251 upper band Centers), the median was 33% and the average was 29% for all Centers. For the two-  
252 hour forecasts, the values were 13%, 23%, 23% and 20%, respectively.

253         In order to understand the traversal trend around forecasted weather probabilities, the  
254 numbers of grid cells with 10% or higher forecast probability value were computed for the entire  
255 four-month one- and two-hour NCWF-6 forecast data set. The NCWF-6 has a 2 nmi grid  
256 resolution, which implies that over the continental United States, there are over 1 million grid  
257 cells. The numbers for one-hour weather forecasts are presented at left, and the two-hour results  
258 are presented at right in Fig. 5c and 5d. With the exclusion of Atlanta Center and inclusion of  
259 Minneapolis Center, each of the five upper band Centers has the most number of >10%  
260 probability value cells. This suggests that those five Centers experienced most convective  
261 weather (at least for the data under consideration.) It should also be noted that for PTP  
262 computation to be relevant, existence of large number of weather cells (over 100,000 for 10%  
263 value), as well as high air traffic is necessary.

#### 264 *c. Fort Worth Center (ZFW) Threshold Parameter*

265         For this study, Ft. Worth Center was selected for further evaluation due to relatively high  
266 convective weather presence, its central location in the NAS and observed probability traversal

267 data. Figures 6a through 6d show the results for ZFW for different parameters for a one-hour  
268 forecast, four-month data set. Figure 6a shows the number of aircraft at various altitudes starting  
269 from ground level up to flight level (FL) 400 in 10,000 ft increments. It is observed that more  
270 aircraft in the ZFW region traverse the probabilities in the lowest 10,000 ft (closer to the  
271 Terminal Radar Approach Control or TRACON), and between flight levels 300 and 400. In the  
272 FL 100-200 range, aircraft fly visual flight rules. In the FL 200 to 300 range, mostly regional jets  
273 are present. The overflights largely fly through the Center between FL 300-400. In the FL 100-  
274 200 and FL 200-300 ranges, 28% PTP was observed (shown by vertical lines in the figures)  
275 while in FL 0-100 and 300-400 altitude bands, PTP values of 30-32% were observed. Fig. 6b  
276 shows results for the time of day statistics. The convective weather usually appears in the  
277 afternoon through evening hours. The 7 am through 1 pm CDT (12-18 UTC) and 1 pm through 7  
278 pm CDT (18-24 UTC) times see intermediate probability traversal activity (PTP=30%). The 7  
279 pm through 1 am CDT (00-06 UTC) sees lower PTP of 28%, as there is lesser traffic and lower  
280 convective activity in the atmosphere. It is seen from the purple curve that the hours of 1 am  
281 through 7 am CDT (06-12 UTC) show PTP of 31% when there is minimal traffic activity.

282         Additionally, the behavior of different airlines and aircraft types was also studied. Figure  
283 6c shows the behavior of four dominant airlines in the Ft. Worth Center. All 4 Airlines were  
284 avoiding between 28 and 30% probability values. Airlines 1 and 3 have DFW as the hub while  
285 the other two do not. Rhoda, et al. (2002) showed that pilots tend to venture into convective  
286 activity more, when they are closer to destination. On the other hand, Fig. 6d shows the number  
287 of aircraft crossing probability values for the four aircraft types in the center. The main  
288 observation was that the MD8x aircraft (green) appear to avoid the 28% contour value, but the  
289 three other aircraft types were avoiding the 32% intensity contours.

290 The two-hour forecast data were processed as well and all the graphs showed similar  
291 behavior to the one-hour cases. For altitudes between FL 100-200 and FL 200-300, 18% PTP  
292 was observed, while all other altitude bands showed a PTP of 23%. For the 11 am to 11 pm CDT  
293 (18-24 and 0-6 UTC) a 23% PTP value was observed while the remaining times of 11 pm to 11  
294 am CDT (6-12 and 12-18 UTC), it was 18%. The airline behavior was similar with the top two  
295 DFW users showing 18% while the other two users had 23% PTP value. Following a similar  
296 trend to one-hour forecasts, MD8x showed 18% PTP while the others were avoiding 23%  
297 intensity contours.

#### 298 *d. Probability Threshold Parameter for Sectors in Fort Worth Center*

299 In order to study the impact of weather in the Ft. Worth Center, PTP value in various  
300 Sectors were computed. Figure 7 presents all the high-altitude sectors (all at and above FL 240)  
301 in the Ft. Worth Center. The 7 sectors for which data are presented in Table 1 are highlighted in  
302 cyan in Fig. 7. These seven sectors contain the four main arrival fixes (shown in yellow) and  
303 have more complex traffic patterns (e.g., transitioning and merging) in the Center. Other ZFW  
304 sectors have lower traffic complexity. Table 1 shows the one- and two-hour forecast (comma-  
305 separated) PTP values. PTP values from FL 240-400 are shown in row 1. Data in other rows are  
306 for times of day, airlines and aircraft types. It is worth noting that sector ZFW86 has a complex  
307 traffic pattern due to arrivals from the east, departures from the south (Houston Center airports)  
308 and multi-directional overflights. It can be observed that mostly ZFW86 has a PTP value, which  
309 on average is at or above other sectors for the altitude range shown. The highest one-hour PTP  
310 value noted is for aircraft-type 4 with 36% in ZFW42, while the lowest one-hour PTP value is  
311 14% in ZFW46 between 6-12 UTC (1 to 7 am CDT) when there's almost no convective weather  
312 and low arrival or overflight traffic. For all ZFW sectors, one-hour values lie between 27 and

313 32% with a 30% average, while the two-hour values lie between 17 and 21%, with a 20%  
314 average. Overall, the average 30% (one-hour) and 20% (two-hour) values for this large case are  
315 valid across all airlines, aircraft types, altitudes and times of day. Since the Sector level values  
316 are close to the Center PTP values, additional Sector level analysis was not deemed necessary to  
317 study aggregate behavior of aircraft streams. A similar analysis can be conducted for three-  
318 through six-hour forecasts but was not done due to widespread low forecast probability values  
319 (see Fig. 3d and 3e).

## 320 **4. Flight routing decisions**

321 When severe weather is forecasted, various options for filing flight plans are available to  
322 airspace users (e.g., Airline Operations Center flight dispatcher) for routing their flights around  
323 or away from regions of severe weather. These include the use of FAA's Severe Weather  
324 Avoidance Plan (SWAP or Playbook) Routes, Coded Departure Routes (CDRs), historic flight  
325 plan databases, individual airline's Preferred Routes, etc. A dispatcher often has to determine if  
326 their flight is going to be moved due to weather or congestion (Sridhar, et al. 2002, Sridhar, et al.  
327 2005). On the other hand, a Traffic Management Coordinator's perspective is to maintain a safe  
328 and efficient flow of traffic through their Center with minimal delays and congestion. Aspects of  
329 flight routing decision processes are considered in this Section. Results presented in previous  
330 sections help in better decision-making during severe weather events.

### 331 *a. Local Reroutes*

332 During the times when severe weather is predicted to occur, it obviously benefits the  
333 operators and users to assess the impact on air traffic. For both the parties, it is useful to have a  
334 capability to evaluate possible rerouting options. Such a system should have Center-level routing  
335 strategies available for a local weather event.

336           Once the probability threshold values have been computed as described earlier, various  
337 route options can be analyzed to assess the balance of demand and capacity. For example, if a fix  
338 for arrival traffic (e.g., Bonham, BYP, see Fig. 8) for DFW airport or overflight traffic  
339 transitioning through the ZFW Center is forecasted to be under convective weather in the next  
340 one- through six-hours, which reroutes can be employed? Which route options can be utilized to  
341 maintain the stream of aircraft flowing without major schedule disruption and minimal additional  
342 workload for controllers, while providing sufficient predictability? In general, Playbook routes at  
343 the national level will impact a large number of aircraft, with associated potential loss of  
344 schedule integrity. For local weather scenarios of a Center-level scope, it is desired that the  
345 impact on other Centers be minimized. The proposal is to reduce the burden on other Centers  
346 while the impacted Center works cooperatively with the Air Traffic Control System Command  
347 Center (ATCSCC). Depending on the situation, traffic managers could employ a local method, a  
348 national strategy, or a hierarchical approach.

349           A local Center-based rerouting what-if analysis capability is presented, in which the  
350 affected Centers can employ local and predefined routes for assessing the impact of various  
351 strategies, in coordination with the ATCSCC. While implementing the National Playbook,  
352 generally the aircraft's flight plan is often modified from origin to destination, resulting in larger  
353 deviation from nominal operations for better system predictability. The concept of Center Routes  
354 proposed here, keeps the flight plan unchanged until the point of entry into the weather impacted  
355 Center. The flight plan is changed only after the last fix before entering the affected Center, with  
356 the planned local reroute up to the destination (for arrivals) or exit from Center (for overflights).  
357 This provides a level of predictability (assuming a satisfactory level of forecast accuracy) to the  
358 dispatcher as well as the traffic manager. It also eliminates the need to route each aircraft

359 individually and maintains the traffic stream. Since the probabilistic convective weather data are  
360 available up to six hours in advance, such strategies could constantly be evaluated for air traffic  
361 management planning decisions in the long term.

362 *b. ZFW Scenario*

363 Traffic enters Ft. Worth Center (ZFW) from four neighbors. Figure 8 (a) shows that the  
364 traffic from Albuquerque Center (at left) mainly enters ZFW through Texico (TXO) and  
365 Panhandle (PNH); from Kansas City (ZKC) Center (above) through Tulsa (TUL); from Memphis  
366 (ZME) Center (at right) through Little Rock (LIT), Ft. Smith (FSM) and Munroe (MLU); and  
367 from Houston (ZHU) Center (below) through Alexandria (AEX) and GIFFA fixes. In this study,  
368 local routes were designed for the scenario where one of the arrival fixes (e.g., Bonham, BYP)  
369 was closed, as in the events of July 10, 2007. Consider a flight plan for an aircraft arriving from  
370 Chicago O'Hare International Airport (ORD), routinely filed with the FAA as  
371 ORD..RBS..SGF..BYP.BYP5 .DFW. In this implementation, the route would be modified, for  
372 example, as ORD..RBS..SGF..TUL..IRW..UKW.UKW9.DFW, using a potential route option  
373 incorporating alternate fixes and a non-impacted arrival fix Bowie (UKW). Once these routes  
374 were designed for arrivals into DFW, what-if analyses were conducted to study the impact on  
375 flights. Metrics of delay, congestion, additional fuel, and distance were then computed.

376 *c. Results of Local Rerouting*

377 Figure 8 presents a scenario when BYP (the northeast arrival fix for DFW) is closed, as  
378 was the case on July 10, 2007 with significant delays for DFW arrivals. The PTP values  
379 computed earlier were used to look at the area covered by one-hour forecast 30% probability  
380 values over the BYP arrival fix. The traffic originally planned to arrive through BYP from  
381 various northeastern origin airports is rerouted along TUL, IRW, SPS and UKW to arrive into

382 DFW. Figure 8 shows the situation before (Fig. 8a) and after (Fig. 8b, routing through IRW)  
383 implementation of the local reroutes through ZFW. Cyan lines show flights that were to arrive at  
384 DFW through BYP, magenta lines show arrivals through UKW and green lines are arrivals  
385 through CQY. The reroutes for this BYP closure scenario were implemented using three  
386 different strategies, which would depend on the location and spread of predicted weather. First  
387 strategy rerouted aircraft to ADM and UKW to arrive into DFW (not shown in Fig. 8 to avoid  
388 clutter). The second strategy rerouted through IRW and UKW (Fig. 8b); while the last strategy  
389 rerouted aircraft even further to go from IRW, SPS and UKW to arrive into DFW (again, not  
390 shown in Fig. 8 to avoid clutter). In each of the three strategies, aircraft coming from Ft. Smith  
391 (FSM) and north of it (upper cyan arrival stream in Fig. 8a) were diverted to the ADM arrival  
392 stream. The aircraft coming from Little Rock (LIT) and southeast of it (lower cyan arrival stream  
393 in Fig. 8a) were routed through Belcher (EIC) and Cedar Creek (CQY) into DFW. These can be  
394 observed by contrasting Figs. 8a and 8b. The lower arrival stream (EIC..CQY..DFW) flight  
395 reroutes were held constant in each of the three strategies.

396         In order to understand how effective these routes are and what the impact on traffic is,  
397 results are presented for each of the three strategies in Table 2. It shows the effect of each  
398 strategy as applicable to a different weather impact and coverage scenario. The reroutes were  
399 implemented in FACET for a four-hour period from 3 to 7 pm CDT (20 to 24 UTC) using traffic  
400 data from July 24, 2007. The data from July 10, 2007 (a Tuesday) would be corrupted with  
401 controller input of rerouting the aircraft due to presence of convective weather over BYP.  
402 Therefore, traffic data from July 24, 2007 (another clear weather Tuesday) and convective  
403 weather data from July 10, 2007 were used for simulating reroutes. In each of the three cases, the  
404 number of impacted flights was 155. Table 2 provides the metrics for each of the three strategies.

405 The aircraft incurred an average of 12, 15 and 18 minutes of delay; 794, 1,012 and 1,235 pounds  
406 of additional fuel; and 42, 54 and 66 nmi additional distance, per aircraft for the three strategies,  
407 respectively. It is worth noting that in each of the three cases, there was no congestion (number  
408 of aircraft above Monitor Alert Parameter) observed in the northwestern sector ZFW47 (where  
409 UKW lies) or in the southeastern sector ZFW89 (where CQY lies). This behavior is observed  
410 mainly due to a smaller number of aircraft present during the evaluation interval. However, this  
411 suggests that rerouting flights to the same region of airspace may not necessarily overload the  
412 airspace but may provide a reasonable alternative to dealing with the weather problem.

413 The last column in Table 2 corresponds to the implementation of the FAA-published  
414 Playbook route, DFW\_BYP1, for arrivals into DFW airport during a BYP closure event. The  
415 result indicates that 218 DFW arrivals are affected. A leading cause for a larger number of flights  
416 impacted is that the current description of DFW\_BYP1 modifies flights not only flying over  
417 BYP, but also over other arrival fixes, CQY and JEN. The use of DFW\_BYP1 does not include  
418 other flights (e.g., overflights or arrivals at other airports) in the Center and separate Playbook  
419 routes need to be implemented to account for those flights. In the local rerouting concept  
420 proposed and implemented here, flights flying over BYP, either arriving at DFW, DAL, Houston  
421 (Intercontinental, IAH and Hobby, HOU), or other nearby airports like San Antonio (SAT), etc.  
422 were all accounted for with less than 10 minutes of flying time change. The DFW\_BYP1 plan  
423 could start modifying flight routes up to two hours (or more) in advance. Figure 8c shows the  
424 scope of the DFW\_BYP1 plan. The green lines show the flight plan amendment that would be  
425 used for aircraft arriving at DFW airport from origins across the northern and eastern part of the  
426 United States. The weather pattern shown is the same as in Fig. 8b. It is clear from Figs. 8b and  
427 8c that the scope of local rerouting is smaller and less impact is felt by air traffic compared to the

428 larger DFW\_BYPI or similar plan, especially for a convective weather problem of a local scope.  
429 It can be observed from Fig. 1 that on July 10, 2007, the aircraft from the north and east were  
430 arriving at DFW through IRW and SPS, which is closest to Strategy 3 implemented for this  
431 research.

432 It is acknowledged that for larger, multi-Center convective weather scenarios, the  
433 National Playbook provides appropriate rerouting and predictability. The capability of local  
434 reroutes proposed here address local weather events. The selection made by traffic managers of  
435 the strategy to implement depends largely on the involved traffic densities and timing of reroutes  
436 to be imposed along with other traffic management initiatives under consideration.

## 437 **5. Conclusions**

438 A method is presented for using probabilistic convective weather forecasts for air traffic  
439 management. Current air traffic and forecasted weather data are synchronized to obtain statistics  
440 of aircraft deviating around weather. A Probability Threshold Parameter (PTP) is derived, which  
441 represents the limiting value of probability that is largely avoided by aircraft. This quantitative  
442 metric is used to assess the probability contour that aircraft are observed to traverse in the  
443 vicinity of forecasted convective weather. The study provided threshold values for the National  
444 Airspace System (NAS) and all the 20 Centers in the Continental United States. The nominal  
445 PTP values for the NAS were computed as 35% and 25% for one- and two-hour forecasts,  
446 respectively. The corresponding values for the 20 Centers were 30% and 20% on average. It was  
447 observed that the 20 Centers are divided into three bands of small, medium, and large number of  
448 aircraft traversing around the forecasted probabilities. The Atlanta and New York Centers  
449 demonstrated higher number of aircraft flying through probability field with proportionately  
450 lower forecasted weather activity, while Minneapolis Center had higher weather occurrence but

451 lower number of aircraft traversal through the probability field. The aircraft behavior in the Ft.  
452 Worth Center was further investigated in detail. The PTP values for different altitudes, times of  
453 day, airlines and aircraft types for Ft. Worth Center and seven high-altitude sectors therein are  
454 also presented. Most of the PTP values observed were in the vicinity of 30% and 20% for one-  
455 and two-hour forecasts, respectively.

456         Using the computed PTP values, a concept of Center-level rerouting is presented. Local  
457 reroutes were implemented in the FACET simulation environment for a rapid what-if analysis  
458 and estimation of impact on arrival and over-flights in a Center. Results for a specific scenario of  
459 the Dallas/Ft. Worth's Bonham (BYP) arrival fix closure are also presented. The metrics include  
460 arrival delay, additional fuel and distance, and congestion in the airspace due to rerouting. It was  
461 observed that the total impact on affected flights was smaller compared to larger scope National  
462 Playbook plan. In the suggested concept, the fewer flights were impacted and handled by locally  
463 impacted Center with no additional congestion.

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## List of Figures

FIG. 1. Integrated display of air traffic and convective weather over Ft. Worth Center, shown with NCWF-6 probabilistic weather (filled) contours and NEXRAD weather contours. The pink and cyan dots represent arrivals to and departures from Dallas/Ft. Worth (DFW) airport with their 20-minute histories.

FIG. 2. (a) A simulated flight ACID1 going west to IND from ORF, traversing one-hour forecasted weather probability contours and corresponding probability contour traversal curve (b). Actual tracks for flight ACID1 with NCWF-6 probabilities (c) and NEXRAD weather (d).

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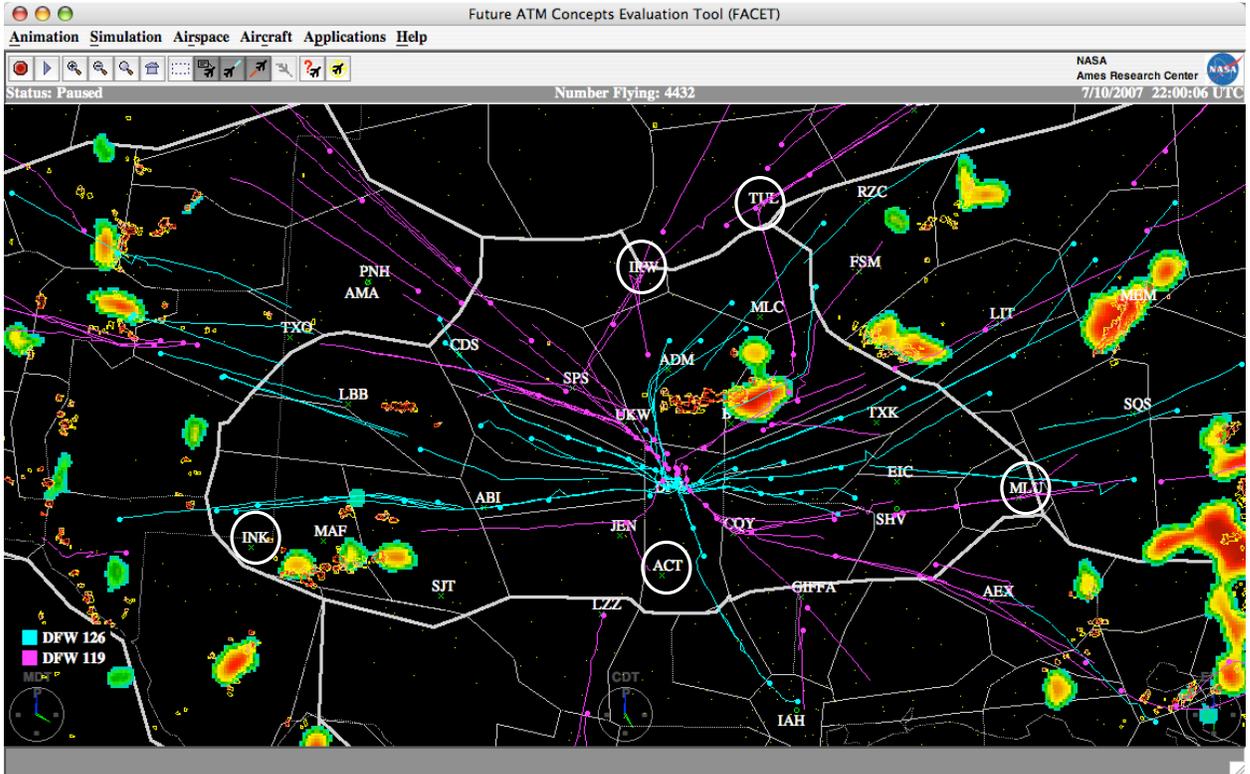
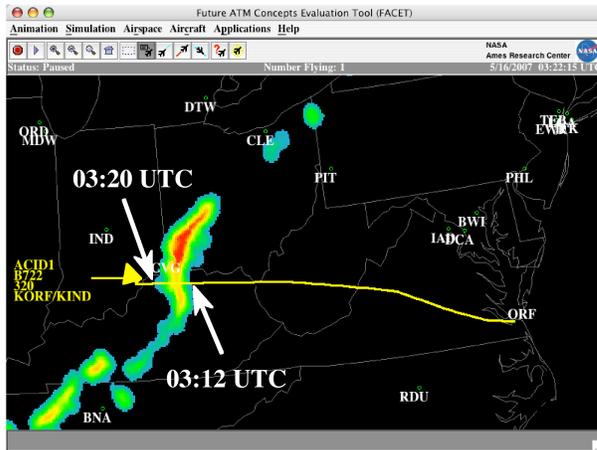
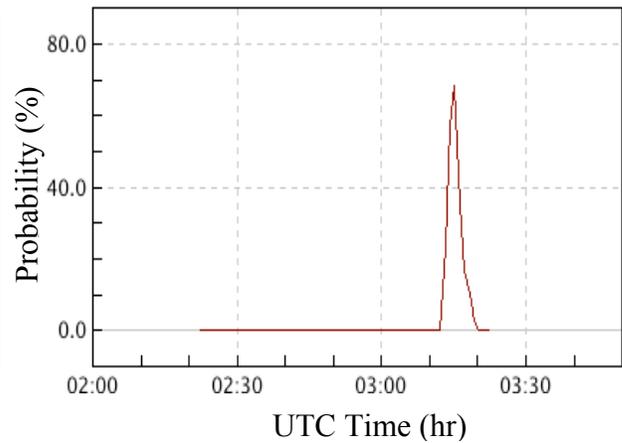


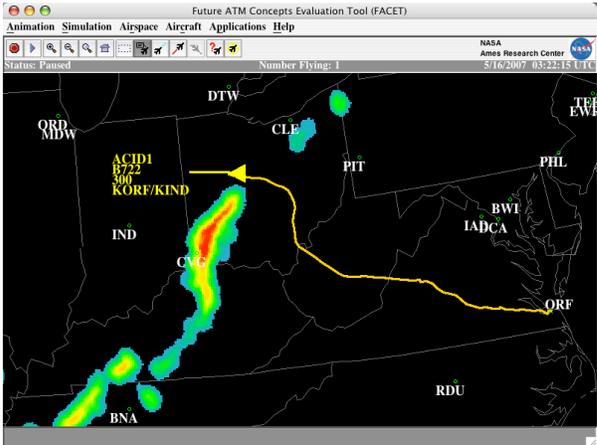
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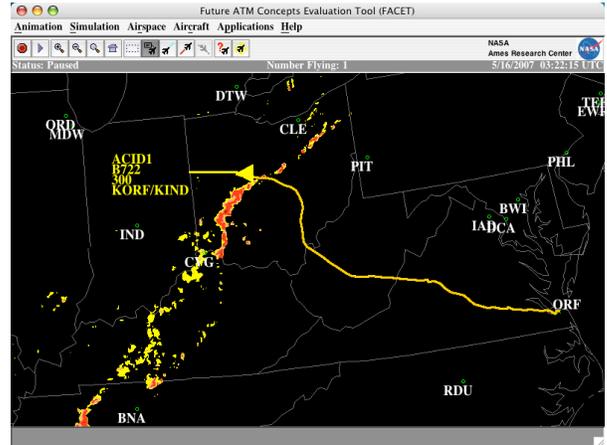
(a)



(b)



(c)



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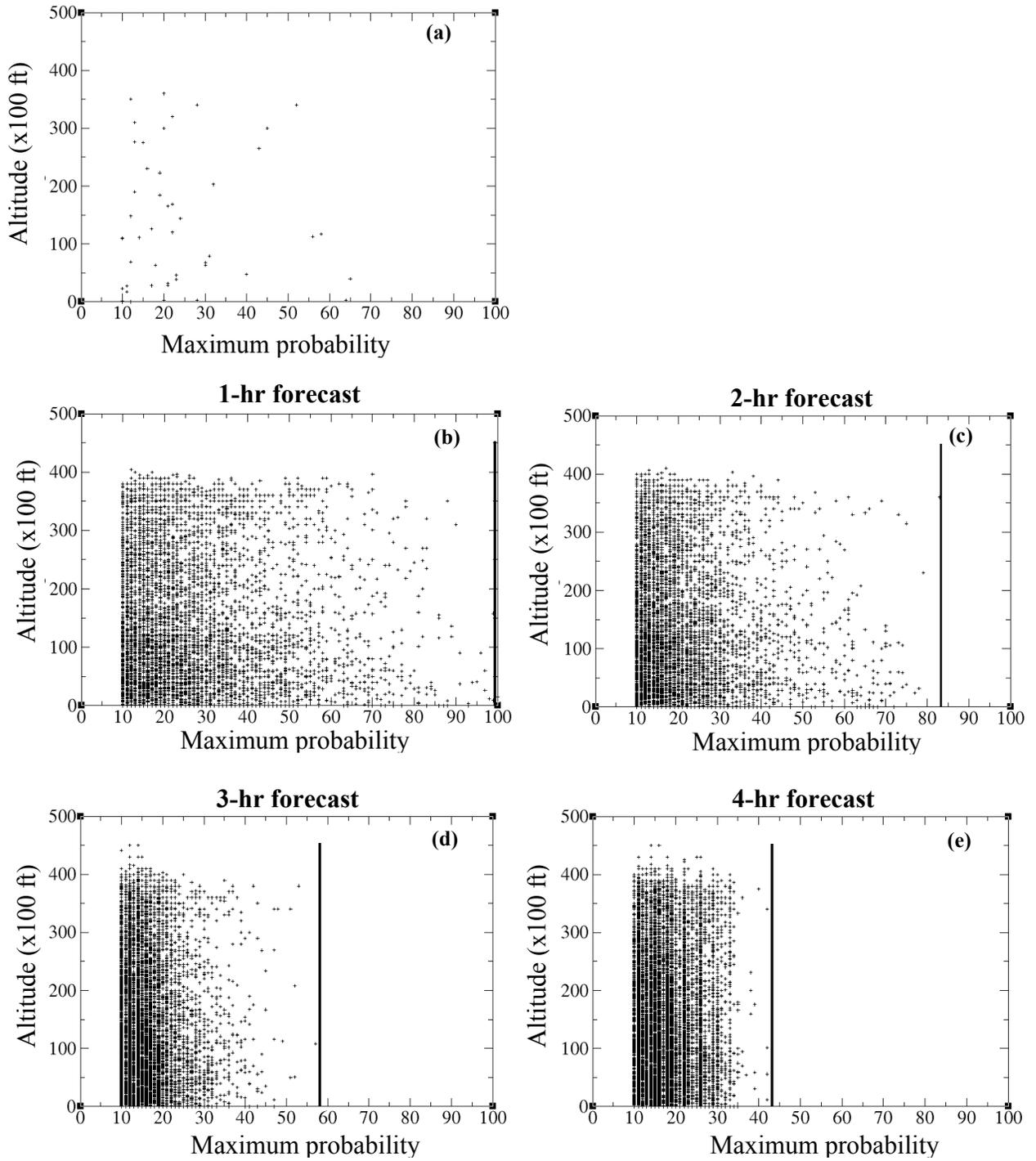
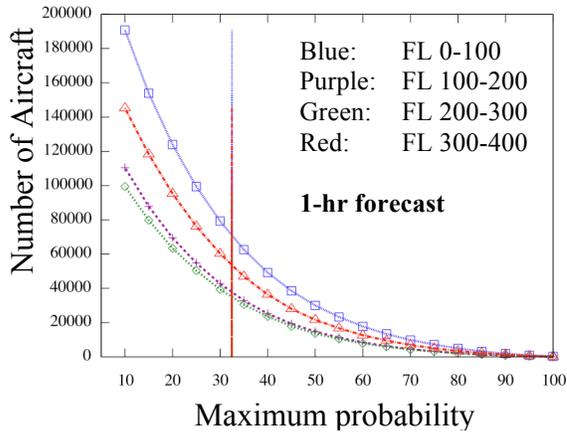
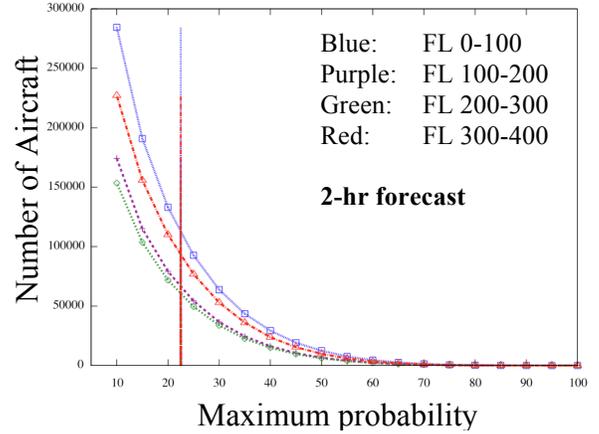


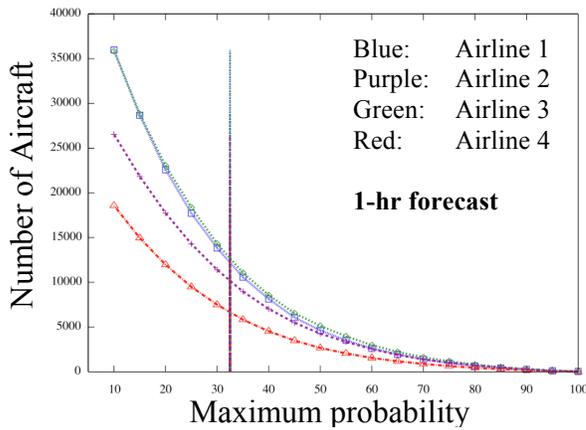
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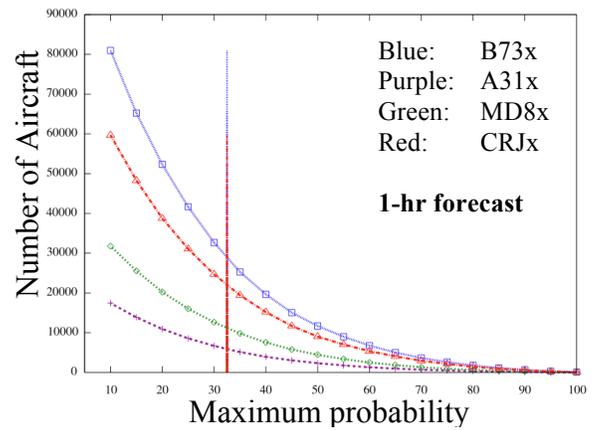
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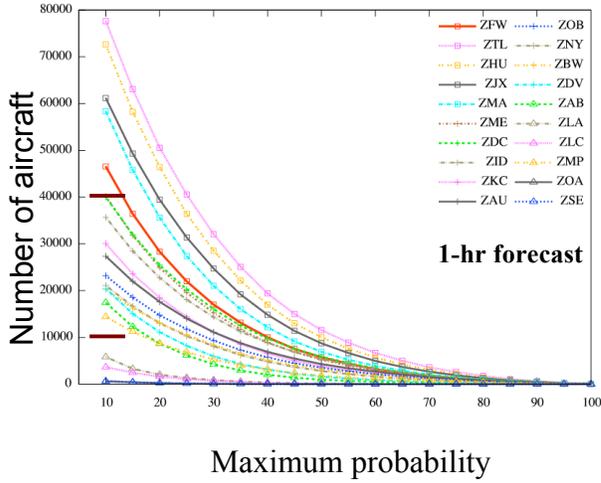


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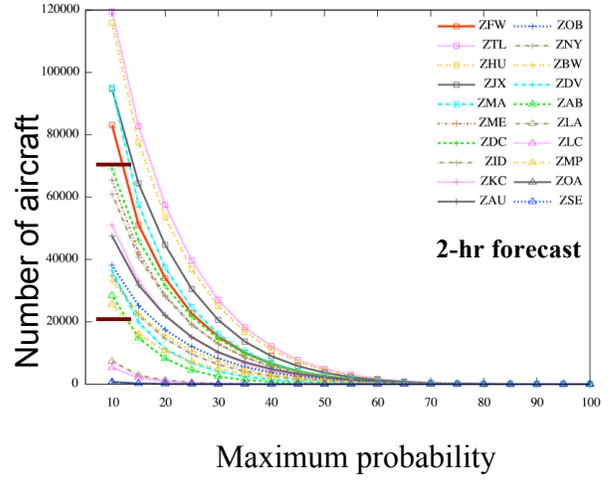


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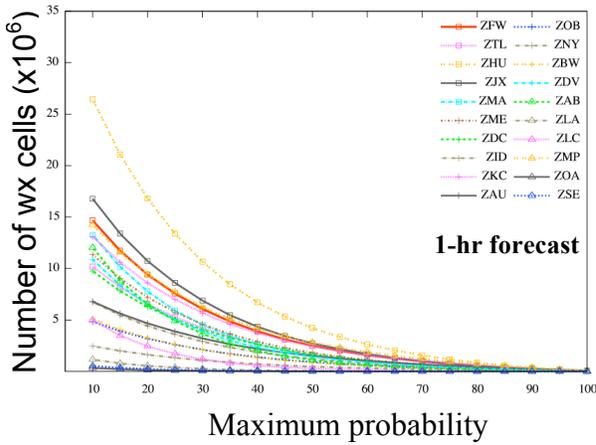
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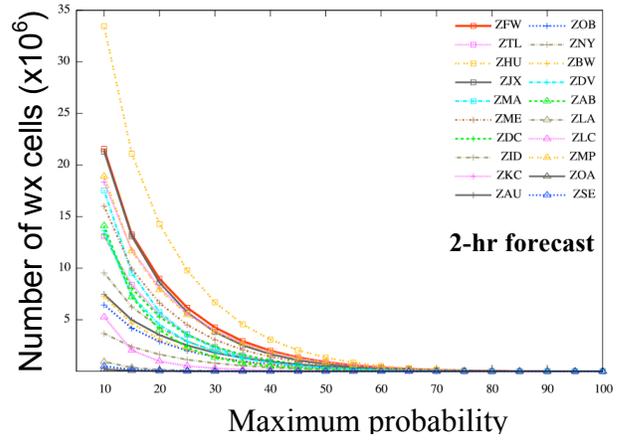
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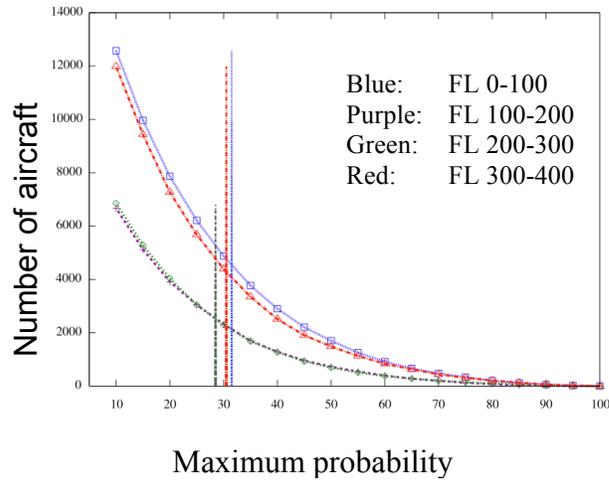


(c)

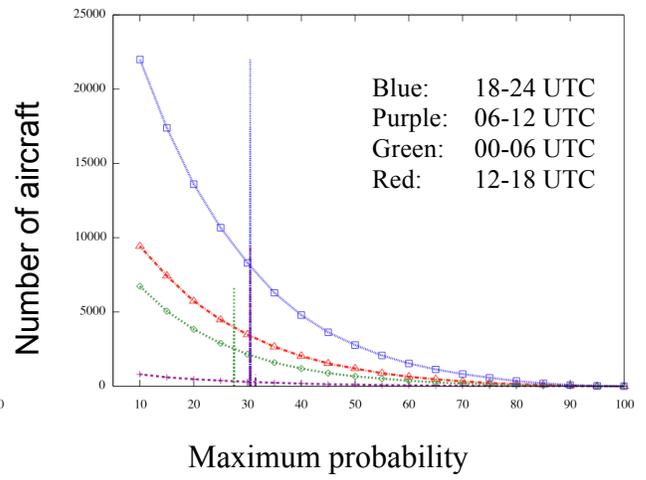


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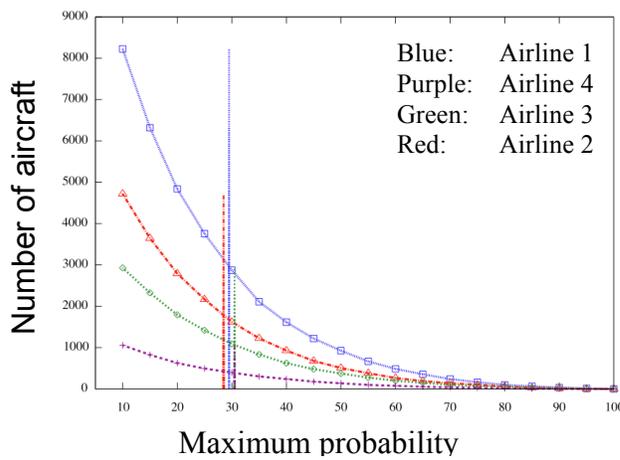
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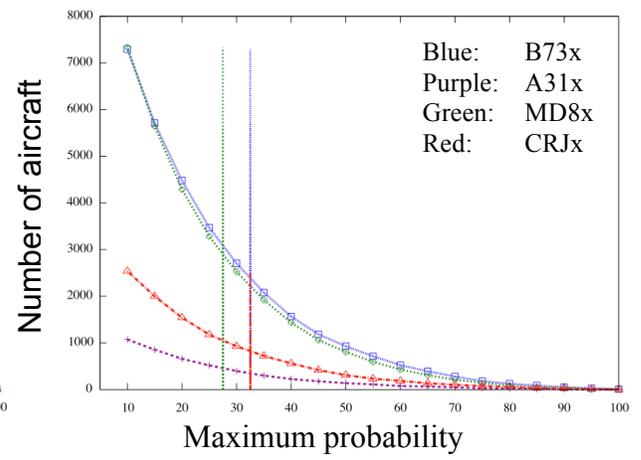
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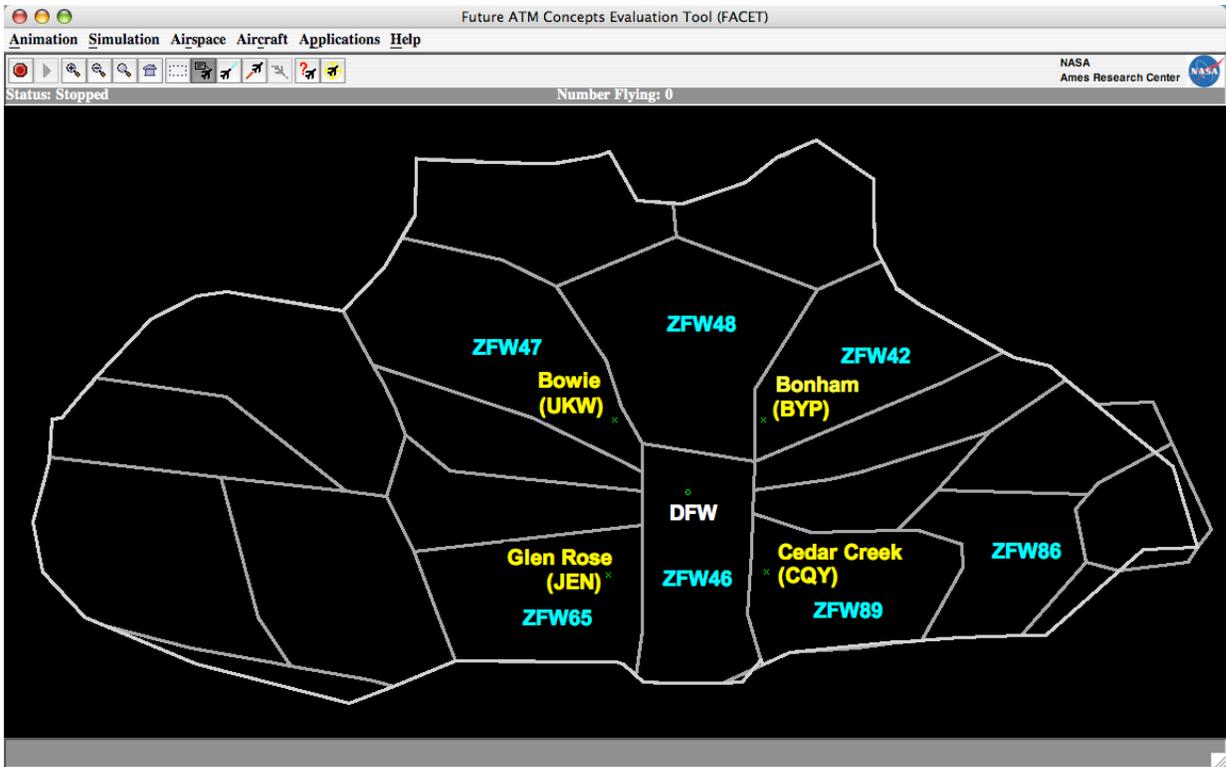


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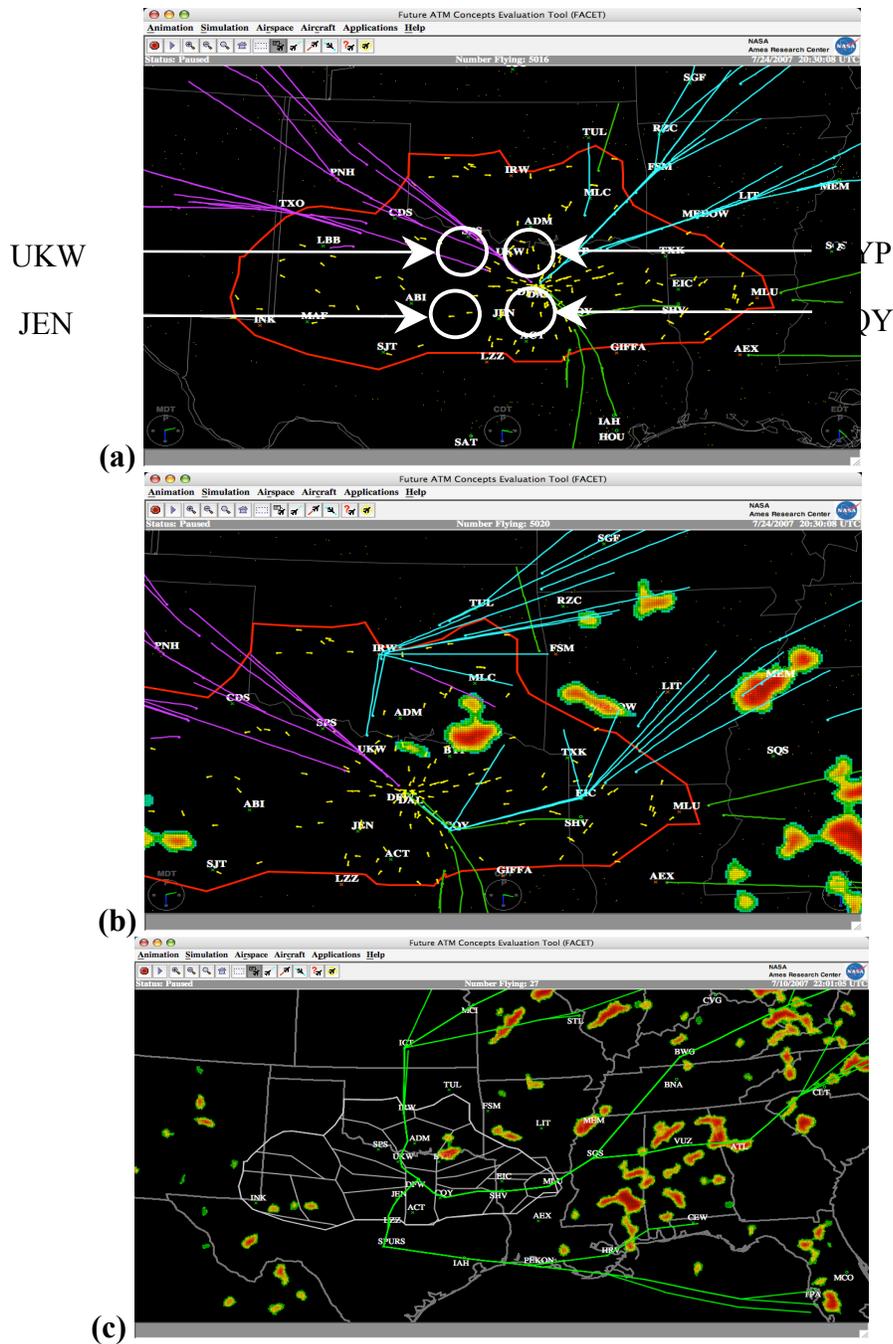


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TABLE 1. The one- and two-hour forecast PTP values for seven sectors (shown in Fig. 6 above, counter clockwise) in ZFW.

Sector	ZFW47 (1hr,2hr)	ZFW48 (1hr,2hr)	ZFW42 (1hr,2hr)	ZFW86 (1hr,2hr)	ZFW89 (1hr,2hr)	ZFW46 (1hr,2hr)	ZFW65 (1hr,2hr)
FL: 240-400	28, 18	29, 19	30, 19	30, 21	29, 19	29, 19	28, 17
Time: 0-6	26, 18	30, 20	34, 21	30, 20	32, 20	30, 19	30, 18
Time: 6-12	32, 17	32, 18	33, 14	25, 23	27, 22	14, 15	32, 18
Time: 12-18	30, 19	30, 17	29, 18	28, 18	30, 16	24, 14	29, 19
Time: 18-24	28, 18	29, 19	28, 19	31, 21	27, 19	29, 19	27, 16
Airline 1	27, 17	28, 18	28, 19	29, 19	27, 19	31, 18	27, 18
Airline 2	25, 18	33, 20	27, 18	27, 18	27, 19	29, 19	25, 15
Airline 3	27, 19	30, 20	31, 20	29, 21	30, 19	28, 19	22, 19
Airline 4	28, 18	27, 20	31, 21	35, 20	32, 19	33, 19	27, 20
Aircraft type 1	28, 18	30, 19	29, 19	32, 21	29, 19	27, 19	24, 19
Aircraft type 2	28, 18	27, 18	29, 18	29, 19	27, 19	29, 18	27, 18
Aircraft type 3	29, 20	29, 19	29, 20	34, 21	30, 19	30, 17	28, 14
Aircraft type 4	32, 18	32, 19	36, 18	27, 21	24, 19	32, 20	28, 18

TABLE 2. The total delay, extra fuel and extra distance metrics for Bonham arrival fix closure, for the three rerouting strategies as well as the National Playbook plan simulation.

	Strategy1 (TUL.ADM. UKW)	Strategy2 (TUL.IRW. UKW)	Strategy3 (TUL.IRW.SPS. UKW)	DFW_BY_P1 (Playbook route)
Number impacted flights	155	155	155	218
Total delay (min)	1,789	2,296	2,823	2,821
Total extra fuel (lbs)	123,108	156,992	191,388	185,833
Total extra distance (nmi)	6,520	8,343	10,287	13,186