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3 **A Wider Perspective on Reducing Fatigue Risk in Aviation**

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7 **Problem**

8 The traditional method of containing fatigue risk when planning airline pilots  
9 has been to apply a collection of strict cut-off rules, for example for the duration  
10 of duty time. Today, when having bio-mathematical models (BMMs) available  
11 which more accurately predict fatigue risk, many organisations are still limiting  
12 their risk reduction approach to only focus on the flights beyond a cut-off limit  
13 from a BMM. This leaves a significant amount of untapped improvement  
14 potential, as risk is also present below the cut-off limit. This paper proposes  
15 metrics for quantification of overall fatigue risk and a method for improved risk  
16 reduction in a crew planning process.

17 **Method**

18 Fatigue risk in aviation is, in simple terms, equal to the risk of a crew member  
19 performing a lapse, slip, mistake or violation with a potentially negative impact  
20 on flight safety, as an effect of low levels of alertness (high sleepiness or fatigue).  
21 Bio-mathematical fatigue models (BMM's), predict, on various scales, the  
22 performance degradation of crew on future rosters. These outputs are  
23 continuous and with a significant variance around the output value that typically  
24 is a prediction for the median of the population of crew operating the sequence  
25 of activities. The variance stems from the models being imperfect, the models  
26 being under-informed (e.g. individual commute times and actual sleep not being  
27 available) and from inter-individual and individual differences in crew. This  
28 leaves the BMM, in individual cases, with a rather low accuracy - particularly for  
29 predictions that are high in alertness.

30

31 For these reasons, it becomes important to look not only at the "few worst  
32 flights" but to attempt to address the overall risk which is reflected by the sum  
33 over all flights. In this paper we have a) proposed definitions for this aggregated  
34 fatigue risk, and b) quantified the difference in overall fatigue risk level obtained  
35 between a few crew planning problems solved (i) using the traditional cut-off  
36 approach, and (ii) using a BMM that, via a metric on overall fatigue risk, in real-  
37 time, guides the solution away from risk. Measurements were made using crew  
38 planning tools and optimisers common for many of the larger airlines. Both crew  
39 efficiency aspects as well as the overall fatigue risk were quantified in a  
40 sensitivity analysis.

## 41 Results

42 Building on the correlation between sleepiness levels and the fatigue risk of an  
43 individual, and the known variance of BMM's, this paper proposes a definition of  
44 two metrics for overall fatigue risk: Absolute Fatigue Risk (AFR) reflecting the  
45 overall risk for a fatigue related incident or accident taking both severity and  
46 frequency into account, and Normalised Fatigue Risk (NFR) reflecting the risk  
47 profile; average risk per flight. Using these two metrics for quantification of  
48 overall risk, the results clearly show that fatigue risk, in the studied planning  
49 problems, can be reduced 4-9 times more effectively with a holistic proactive  
50 approach and without a drop in crew efficiency, , compared to a more traditional  
51 approach using only a strict cut-off.

52  
53 It is further shown, when keeping all constraints intact, that there is a point after  
54 which the return diminishes and costs are increasing much faster than fatigue  
55 risk is being reduced. We also demonstrate that an even higher improvement, at  
56 a lower cost, can be achieved if a few rules are allowed to be re-aligned with  
57 human physiology - something a BMM can be used for in a series of iterative  
58 what-if simulations.

59  
60 This study is done using just a few network structures for two pilot operations  
61 and it should be noted that results are highly dependent on the flight schedule,  
62 the crew base establishment and additional constraints present.

63  
64 [Picture over the results when solving a planning problem without (left), and  
65 with (to the right), the assistance of a BMM. Crew efficiency here represented by  
66 the number of duty days.]

## 67 Discussion

68 There is no doubt that flight and duty time limits, using hours and minutes of  
69 work/rest in various combinations, will remain the main constraints governing  
70 the design of crew pairing and rosters. It is also clear that only complementing  
71 them with an additional rule, stating that crew should not be planned beyond a  
72 certain threshold on an output from a BMM, is not an effective approach  
73 compared to the alternative investigated here. This alternative is to quantify  
74 overall fatigue risk and allow such a metric guide the basic design of crew  
75 schedules, while naturally still respecting all of the traditional constraints. Even  
76 if we are unlikely, in the industry, to arrive at defining rules for overall risk, it is  
77 clear that it is these metrics we should place a higher focus on in the future as  
78 they much closer reflect the actual risk for a fatigue related incident or accident.  
79 Verifying that a definition of overall fatigue risk is the best possible one, should  
80 not be taken as an excuse for not using one, as we know the overall risk is what  
81 will produce the undesirable outcomes.

82  
83 BMM's will hopefully be used increasingly in a proactive way rather than only  
84 being applied after the planning results are ready, changing only a few flights.  
85 BMM's may also be used in this way to improve the rules through iterative what-

86 if's, alternatively relaxing them to achieve efficiency, and alternatively making  
87 them stricter to better capture problematic flight combinations.

88 **Summary**

89 This study clearly illustrates how fatigue risk can be more effectively reduced in  
90 crew planning by guiding the construction of crew pairings and rosters with bio-  
91 mathematical modelling, ensuring risk contribution is taken into account from a  
92 much wider set of flights than typically done today. It would be highly valuable if  
93 the airline industry could agree on a definition for the aggregation of overall risk,  
94 and shift the focus away somewhat from only the absolute value of an individual  
95 flight - using a wider perspective for fatigue risk reduction.  
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