

# How robust are robust schedules in reality?

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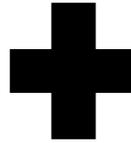
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Funded by :

SNSF - Project 200021-118547

# Introduction



# Some numbers



## □ Huge economical impact

- Air France-KLM 35 Mio € / day
- Lufthansa 48 Mio € / day
- IATA: \$200 Mio / day to air sector

## □ Spill out due to disrupted / blocked passengers

# Why robustness appeals for airline scheduling

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## □ Airlines have low profitability

- < 2% profit margin (US, 2007)

## □ High delays and implied delay costs

- 4.3 Billion hours delay (US, 2008)
- \$41 Billion delay costs (US, 2008)

# Worse is still to come

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## □ Growth:

- 2.5% more flights annually
- Every 1% additional flights incur an additional 5% delays (Schaefer et al., 2005)
- => Yearly increase of delays of 12.5%

## □ Europe: 50% of flights in 2030 depart or land at congested airports

## □ Airlines must react – we try to help

- Improve operations in a congested network

# Outline

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## □ Optimization under uncertainty

- In general
- In airline scheduling

## □ Robust Maintenance Routing Problem

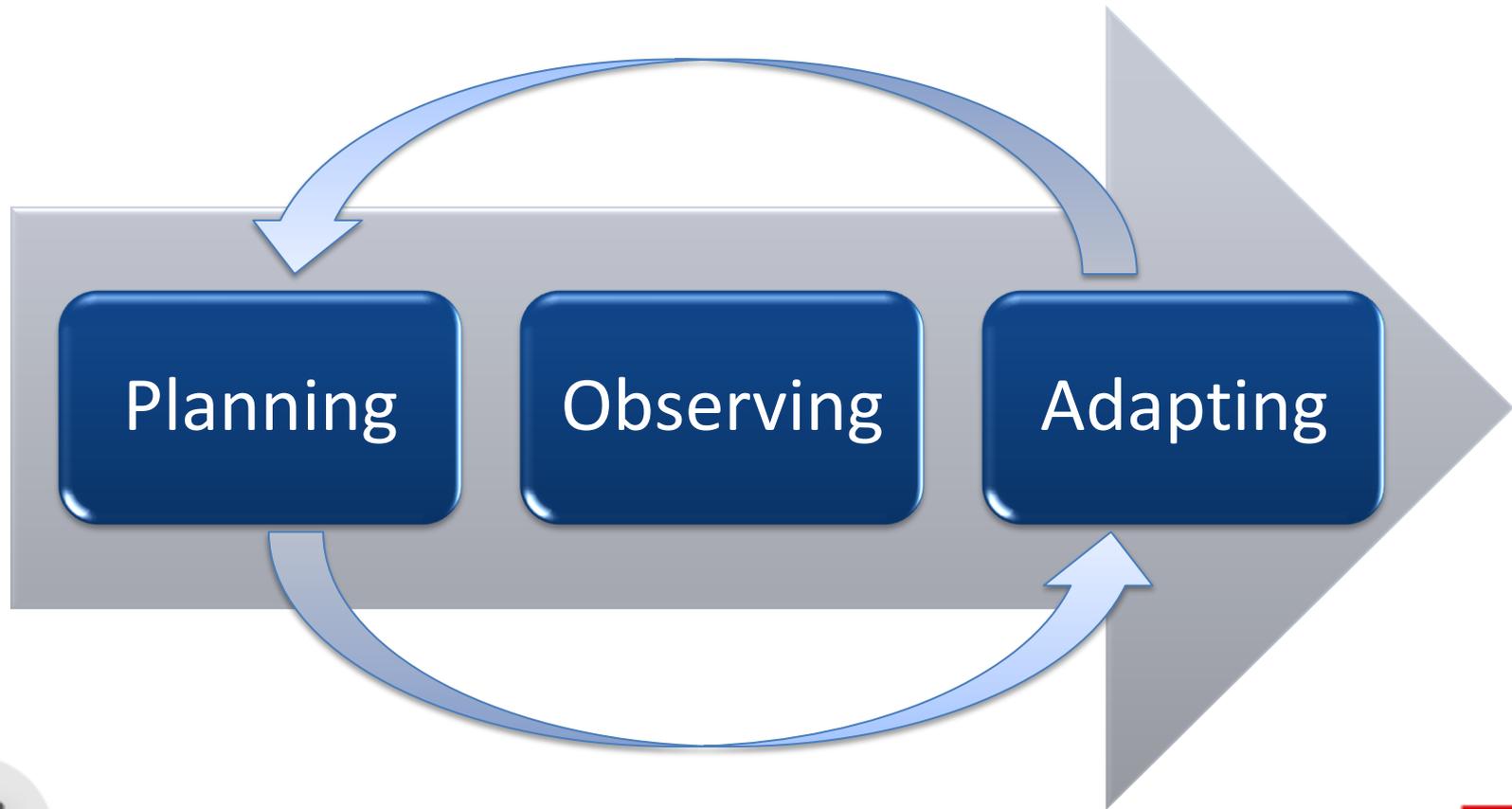
- Definitions
- “Robust” and “Recoverable” models

## □ Simulation – preliminary results

- Methodology to evaluate and compare robust solutions
- Preliminary a priori and a posteriori results

# General Optimization Problems

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# Robustness: plan for stability and reliability

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## □ Optimized solutions have

- Highest “*expected*” revenue/yield/profit
- Known to be sensitive to noise

## □ Robust solutions have

- Lower expected revenue/yield/profit
- Higher reliability

# Definition of robustness

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## □ Unclear in literature

- For more “***stable***” solutions (that remain feasible)
- For more “***flexible***” solutions
- For solutions with lower “***operational costs***”

## □ How to determine what “more robust” means?

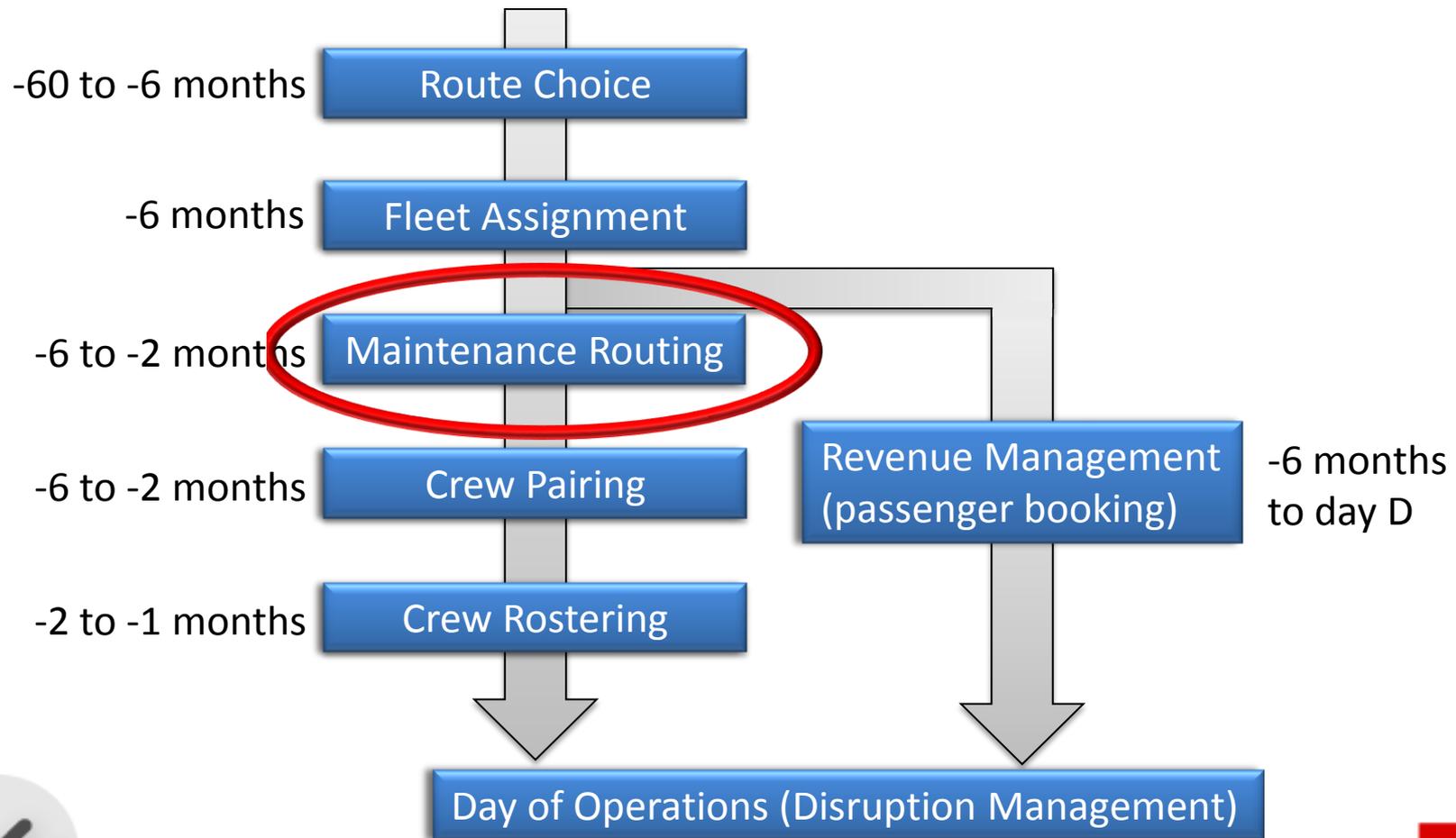
- What metric to use?
- Should it be a priori or a posteriori?

# Parallel to Stochastic Programming

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- What is the equivalent to robustness
  - Stochastic optimization
  - Stochastic optimization with recourse
  - Risk management / chance constraint programming?
  
- Or are these robust methods themselves?

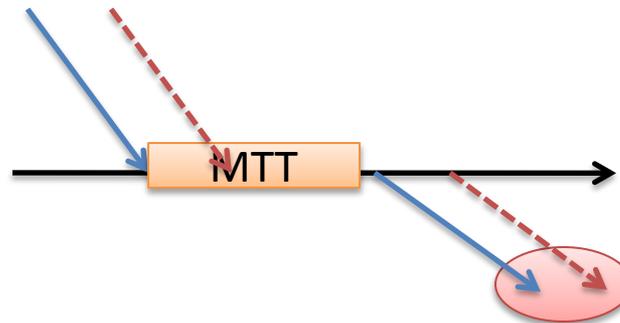
# Airline Scheduling: An iterative Process



# Robustness in airline scheduling

## □ Robust airline schedules are

- Operationally more efficient
- Less sensitive to delay
  - i.e. with reduced delay propagation



# Delay Propagation

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- 2 types of delays for each flight
  - **Independent** delay: generated during a flight
    - At any stage (taxi, runway, landing,...)
  - **Propagated** delay
    - Delay due to previously delayed flight
    - Propagation is downstream (possibly to several flights)

□  $Del(f) = ID(f) + PD(f)$

□ Robustness proxy = expected PD

- To be minimized

# Other meanings of robustness

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- Robustness is also used as a “*flexibility*” measure
  - Facilitates recovery
  - Reduces recovery costs

- We differentiate
  - **ROBUSTNESS** vs **RECOVERABILITY**

# Robust Maintenance Routing Problem (MRP)

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## □ Deterministically known

- Original schedule (1 maintenance route/aircraft)

## □ To determine

- New routes for each aircraft
- And/or new departure times for each flight

## □ Constraints

- Maintenance routes are feasible for each aircraft
- All flights are covered exactly once
- Each flight is retimed by at most  $\pm 15$
- Total retiming of all flights of at most  $C$  minutes (500 or 1000)

## □ Objective

- Optimize robustness metric

# Used Uncertainty Feature Optimization (UFO) Models

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## □ Use different UFs:

- IT: maximize total idle time
- MIT: maximize sum of minimal idle time of each route
- CROSS: maximize nbr plane crossings
- PCON: maximize passenger idle connection time
- MinPCON: maximize minimal PCON

## □ Solved with CG algorithm (COIN-OR – BCP package)

# Benchmark

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## □ Models from literature

- EPD: minimize expected propagated delay (Lan et al., 2006)
  - No retiming
  - Allow only plane swaps
- EPD2: minimize expected propagated delay (AhmadBeygi et al., 2008)
  - No plane swaps
  - Allow for retiming by  $\pm 15$  minutes
  - Total retiming bounded (500 or 1000 minutes)

## □ Solved with same CG algorithm (COIN-OR – BCP package)

# Measuring Recoverability: Methodology

- ❑ Solve Robust MRP using different models
- ❑ Apply some disruption scenarios
  - Differentiate *independent* and *propagated* delay
  - Update propagated delay according to schedule
- ❑ Solve the recovery problem
  - Using same recovery algorithm
- ❑ Evaluation with external recovery cost evaluator
  - Data and cost-evaluator provided by the *ROADEF Challenge 2009*

Planning

Observing

Adapting

# Scenario Generation

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- EPD and EPD2 require expected delay for each flight
  - Generate two distributions using historical data from similar airline (scenarios 1 and 2)
  - Generate several scenarios drawing from each scenario
    - No variability (perfect information)
    - Low variability (  $\sigma = 0.1 \hat{\mu}$  )
    - High variability (  $\sigma = 0.5 \hat{\mu}$  )
- Evaluate solutions on all scenarios and apply recovery algorithm

# Generated schedules

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- ❑ UFO solutions are the same for both scenarios
  - UFs are non-predictive models
  
- ❑ EPD solutions are different
  - Solution depends on estimated delay distribution
  
- ❑ Use two “*realities*” to simulate erroneous predictive models

# Simulation Overview – UFO solutions

Scenario/Solution	Solutions Sc. 1	Solutions Sc. 2
Scenario 1	NEUTRAL	NEUTRAL
Scenario 2	NEUTRAL	NEUTRAL

# Simulation Overview – EPD and EPD2

Scenario/Solution	Solutions Sc. 1	Solutions Sc. 2
Scenario 1	OK	WRONG DISTRIBUTION
Scenario 2	WRONG DISTRIBUTION	OK

# Comparison Criteria

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- ❑ Compare a priori AND recovery statistics
- ❑ A priori
  - UF values
  - EPD
- ❑ Recovery statistics
  - Recovery costs
  - Aircraft statistics
    - Total aircraft delay
    - Canceled flights
  - Passenger statistics
    - Total passenger delay
    - Rerouted passengers
    - Canceled passengers

# Used Instance

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□ 608 flights

□ 85 aircraft

□ 36010 passengers

□ 1 day

# A priori robustness statistics (max retiming = 500 minutes)

		Original	IT	MIT	PCON	EPD	EPD2
<b>Scenario 1</b>	EPD [min]	8453	8265	8431	8496	8411	<b>7953</b>
	IT [min]	12000	<b>12185</b>	12010	12135	12010	12060
	PCON [min]	10815	10950	10860	<b>11815</b>	10815	10795
<b>Scenario 2</b>	EPD [min]	7282	7185	7221	7221	7251	<b>6732</b>
	IT [min]	12000	<b>12185</b>	12010	12135	12065	12110
	PCON [min]	10815	10950	10860	<b>11815</b>	10815	10855

# Simulation Overview – EPD and EPD2

Scenario	S1	S2
S1	OK	WRONG DISTRIBUTION
S2	WRONG DISTRIBUTION	OK

# Average Results (25 scenarios in each “reality”)

		Original	IT	MIT	PCON	EPD	EPD2
Scenario 1	# canc. Flts	13.2	13.2	12.3	11.8	<b>8.5</b>	11.2
	P.D. [min]	17,738	17,352	17,692	17,843	17,827	<b>16,866</b>
	Rec Cost [€]	872,942	#	#	714,236	<b>676,273</b>	866.298
Scenario 2	# canc. Flts	9.9	9.8	9.4	8.1	<b>6.5</b>	7.7
	P.D. [min]	14,115	13,973	14,029	14,052	13,967	<b>13,310</b>
	Rec Cost [€]	548,194	#	#	<b>422,551</b>	423,997	449,128

# Simulation Overview – EPD and EPD2

Scenario	S1	S2
S1	OK	WRONG DISTRIBUTION
S2	WRONG DISTRIBUTION	OK

# Average Results (25 scenarios in each “reality”)

		EPD_S1	EPD_S2	EPD2_S1	EPD2_S2	PCON
<b>Scenario 1</b>	# canc. Flts	8.5	8.6	11.2	11.6	11.8
	P.D. [min]	17,827	17,697	16,866	17,186	17,843
	Rec Cost [€]	676,273	684,246	866,298	915,433	714,236
<b>Scenario 2</b>	# canc. Flts	6.5	6.5	7.9	7.7	8.1
	P.D. [min]	13,971	13,967	13,624	13,310	14,052
	Rec Cost [€]	428,885	423,997	461,774	449,128	422,551

# Conclusions

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- ❑ No absolute meaning of robustness
  - How to measure?
  - How to evaluate?
  
- ❑ Methodology to compare solutions
  - A priori using pre-defined proxies
  - A posteriori using recovery statistics
  
- ❑ Preliminary results show that
  - Proxies are inter-correlated
  - Using evaluation approach allows better understanding of these inter-correlations and their implications

# Open Research Directions

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- Extend simulations and perform deeper analysis to
  - Better understand relations between proxies
  - Understand correlations between
    - a priori proxies
    - a posteriori proxies (recovery statistics)
    - Structure of the recovery algorithm
  
- Will this analysis allow to define robustness...
  - ... with respect to a given recovery algorithm?
  - ... with respect to a chosen proxy?

# The End



# Thank you for your attention!

