

DLR student presentations

Introduction to project WeCare

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NASA/DLR Virtual Institute Event

August 21, 2014



Knowledge for Tomorrow



Project WeCare

Overview



Key objectives

- Study of relationships between non-CO2 effects and meteorological processes to realize an eco-efficient air traffic system.
- Understanding the differences between strategic and tactical climate mitigation options

Research

1. Quantification of the climate impact reduction potential of weather-based climate-optimized aircraft operations (-> trajectory optimization)
2. Cost-benefit-analysis of different operational and technological strategies to reduce the climate impact of the global air traffic, today and in the future (-> scenario-based air traffic simulation)
3. Development of new strategies for measuring the influence of air traffic on the atmosphere and experimental proof of certain effects (-> atmospheric physics)

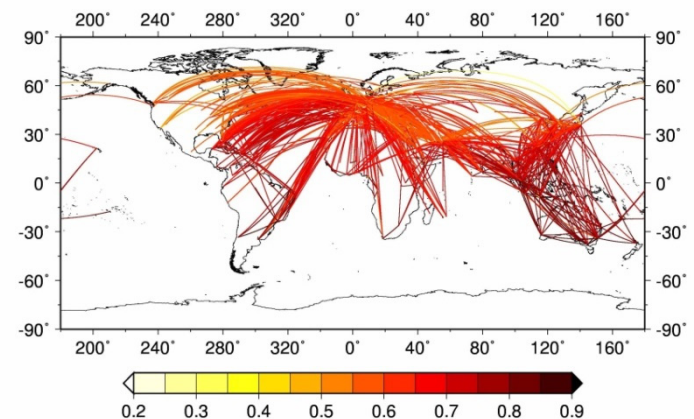
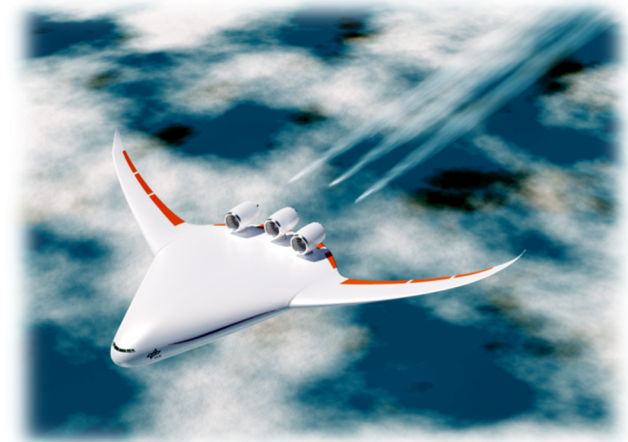


Project WeCare

Future Air Transportation System

Cost-benefit-analysis of different operational and technological strategies to reduce the climate impact of the global air traffic, today and in the future, therefore

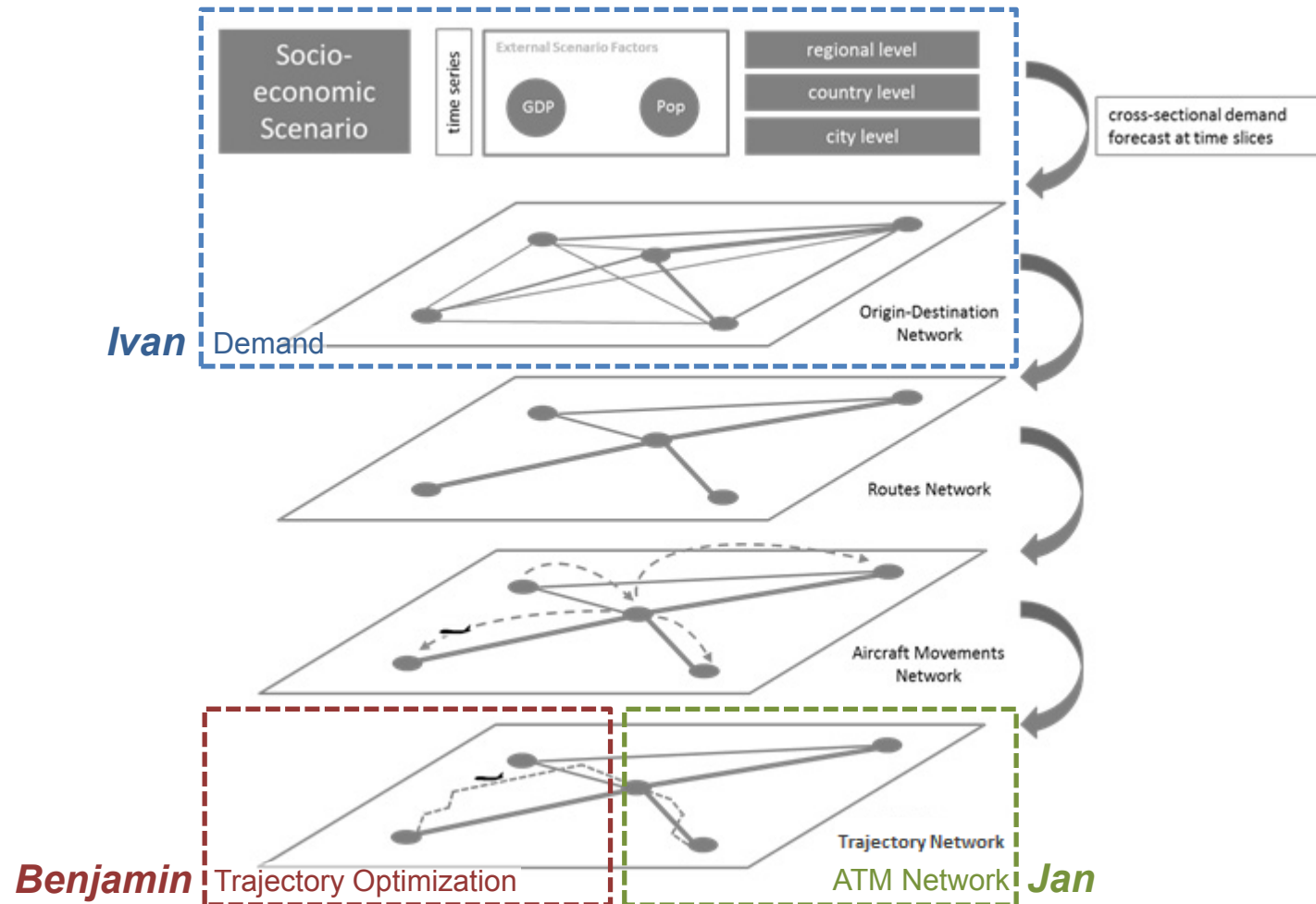
- Scenario based description of the global future air transportation system
- Modelling of worldwide demand, flight routes, fleet mix and aircraft operations
- Integration of unconventional aircraft configurations
- Definition of innovative operational strategies
- Performing air traffic simulations considering adapted operations and/or new aircraft types



Dahlmann, 2012

Project WeCare

4 layers approach



Ghosh et al, 2014

Forecast of origin-destination air passenger demand between global city pairs using future socio-economic development scenarios

Ivan Terekhov

NASA/DLR Virtual Institute Event

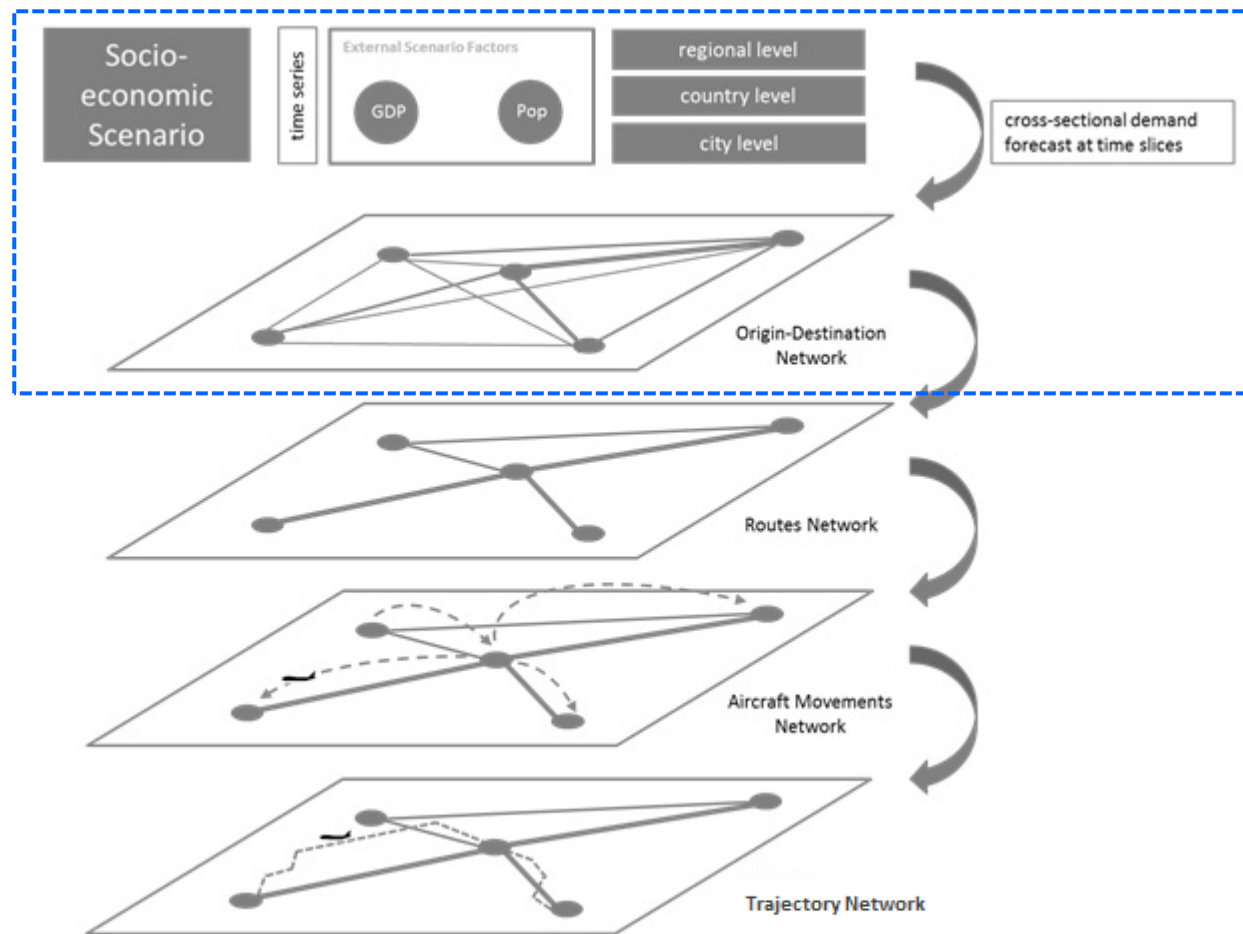
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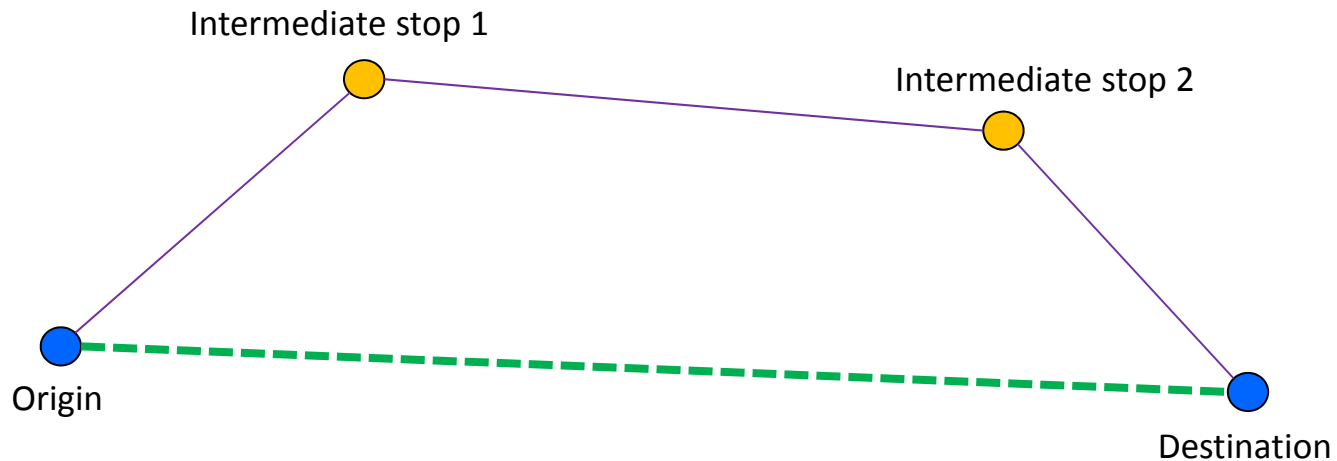
WeCare. 4 layers approach



R.Ghosh et al, 2014

Air passenger demand

- Air passenger demand is a number of passengers which are traveling between two settlements by air transport regardless of intermediate stops within a year.



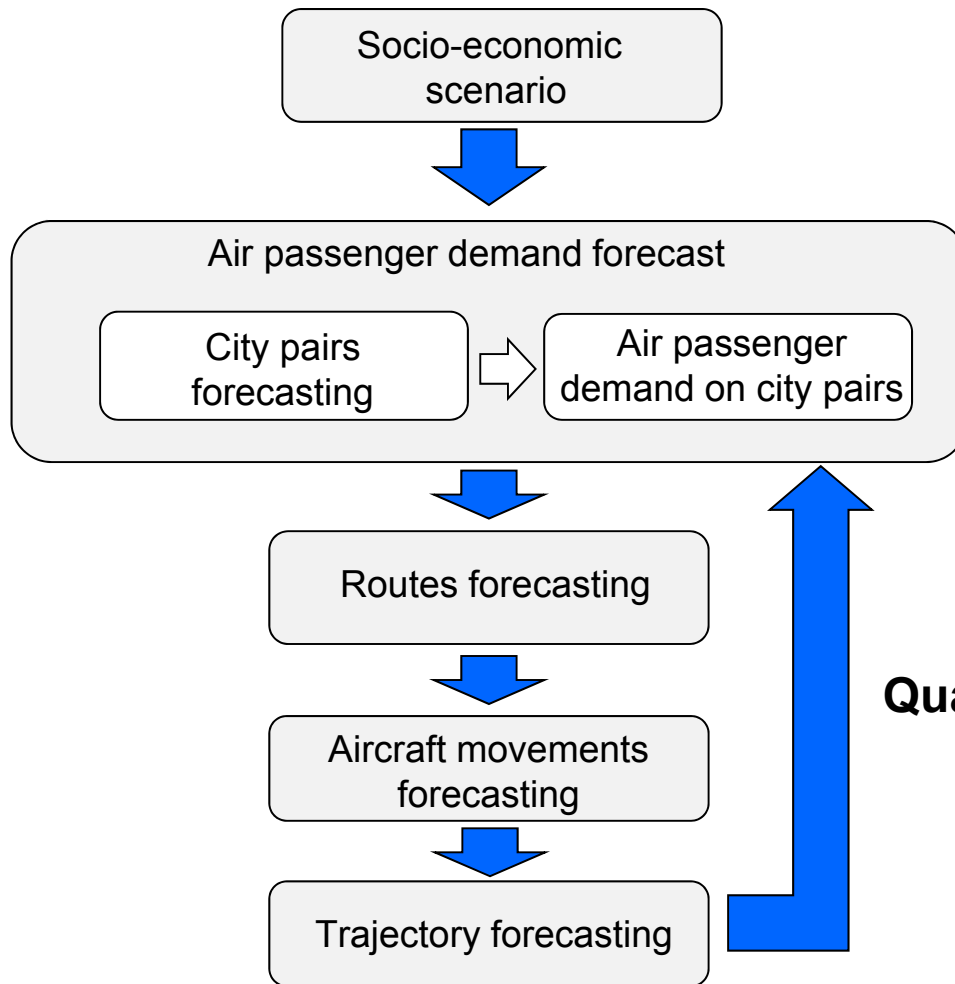
- Real routes between origin and destination
- - - Air passenger demand between origin and destination

Air passenger demand forecast

Aggregation level	<i>city level</i>
Number of elements	<i>as many as possible</i>
Choosing criteria for elements	<i>at least one airport</i>
What is forecasted	<i>number of passengers, changes in city pairs connections</i>
Forecasting method:	<i>Quantitative analogies (passengers), Gravity model (connections)</i>



Air passenger demand forecast. The method



Quality Travel Index (QTI)

$$QTI = (QTI_f, QTI_{tt}, QTI_s)$$

QTI_f – frequency (frequency \uparrow , $QTI_f \uparrow$)

QTI_{tt} – travel time (tt \downarrow , $QTI_{tt} \uparrow$)

QTI_s – segments (segments \downarrow , $QTI_s \uparrow$)

Quality Travel Index (QTI)

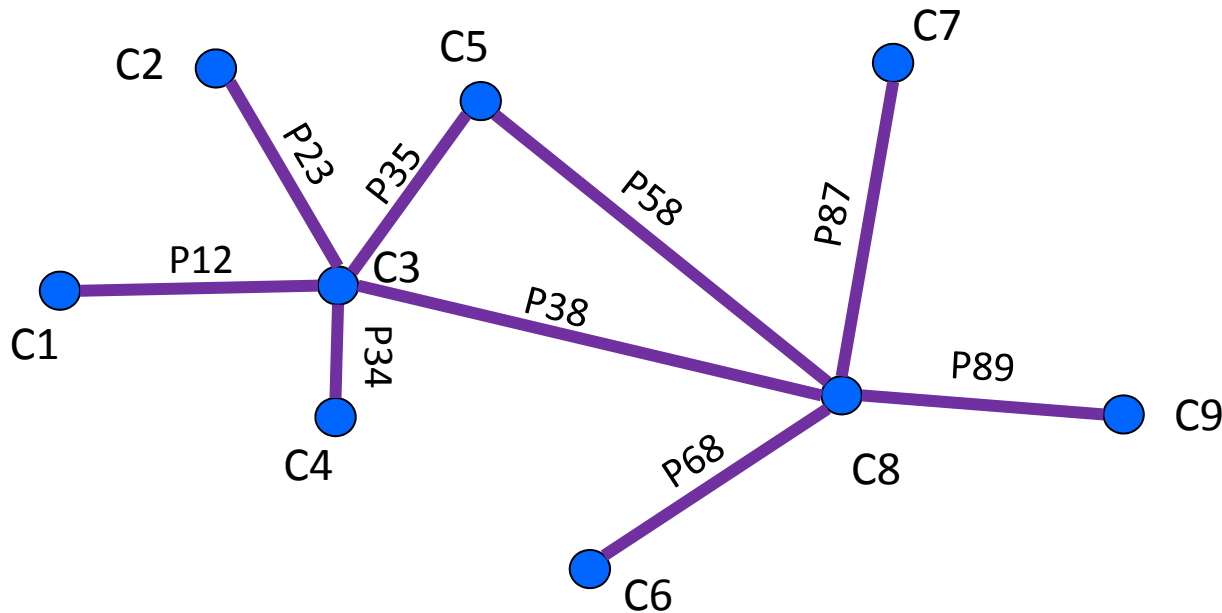
Air passenger demand forecast. The method

- **City pairs definition**

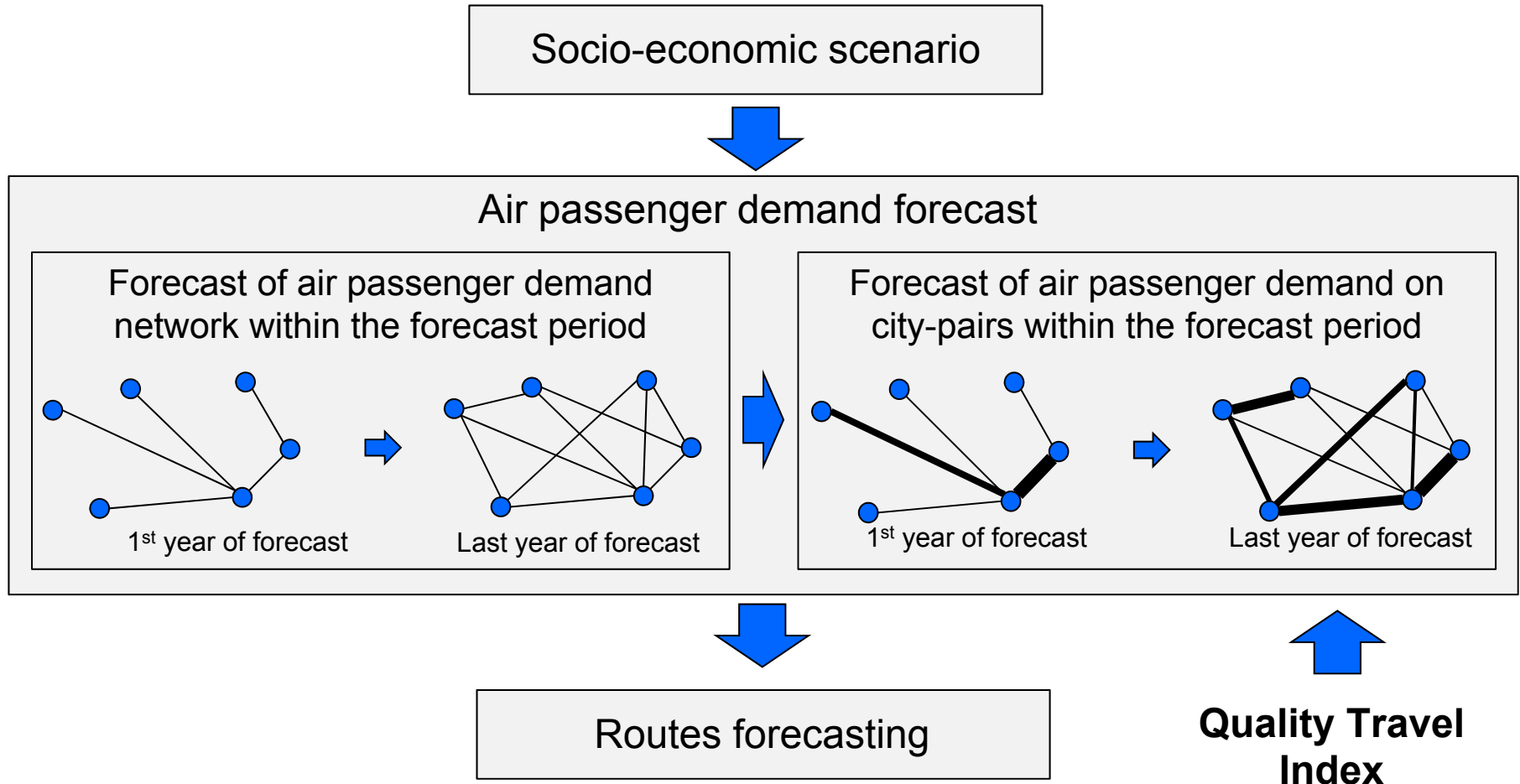
Gravity model based on cities' socio-economic indicators (GDP, population, airfare)

- **Air passenger demand on city pairs definition**

Quantitative analogies between forecasted year and the base year



Expected results



Current results

What has been done:

- For 2012, data has been collected on the demand and airfare of airport pairs from ADI database;
- 2012 data has been rearranged from the airport to city level;
- Geographical coordinates, city GDP and city population have been found for cities in the database ;
- Clusterization of the cities in 2012 database has been made;
- Gravities between cities for the 2012 base year have been calculated in order to calculate reference gravities for each cluster pair;
- Based on forecast, gravities between cities for years 2015-2050 (5 years step) have been calculated;



Current results. Base year 2012

Origin ADI data (airport level)

Connections: **759 816**
Passengers: **2 602 804 691**
IATA codes: **6435**

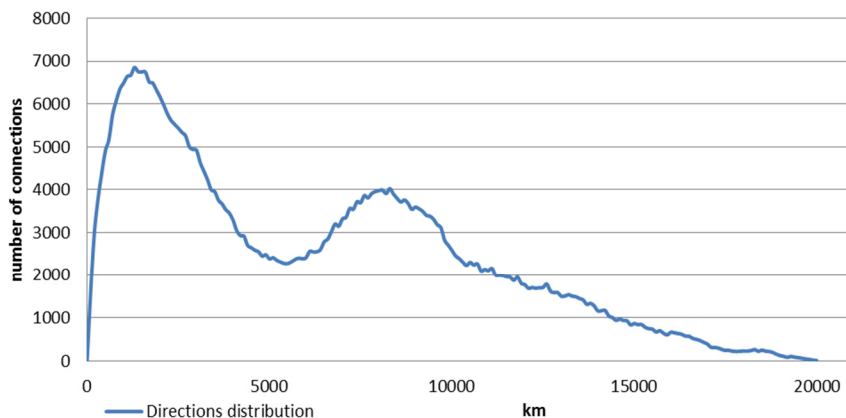
Rearrangement

Connections: **-31,77%**
Passengers: **-2,29%**
IATA codes: **-2,88%**

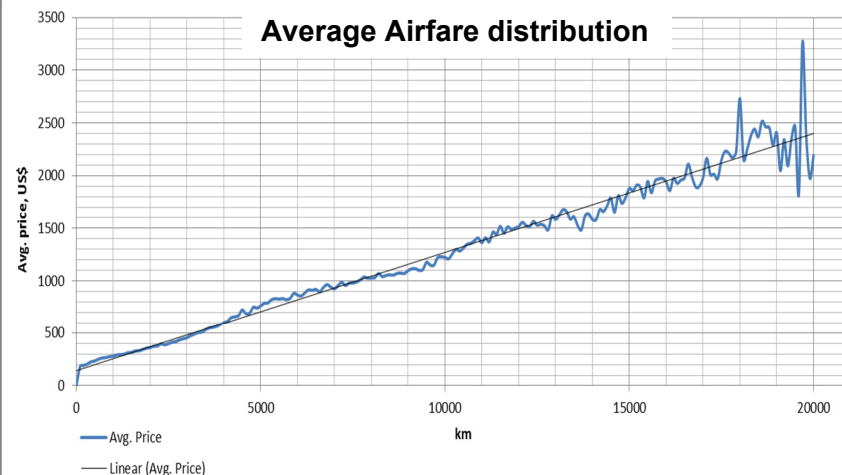
Rearranged ADI data (city level)

Connections: **518 396**
Passengers: **2 543 211 982**
IATA codes: **4319**

Directions distribution



Average Airfare distribution





Thank you!



Flight Trajectory Optimization

Considering wind and penalty areas

Benjamin Lührs

NASA/DLR Virtual Institute Event

August 21, 2014

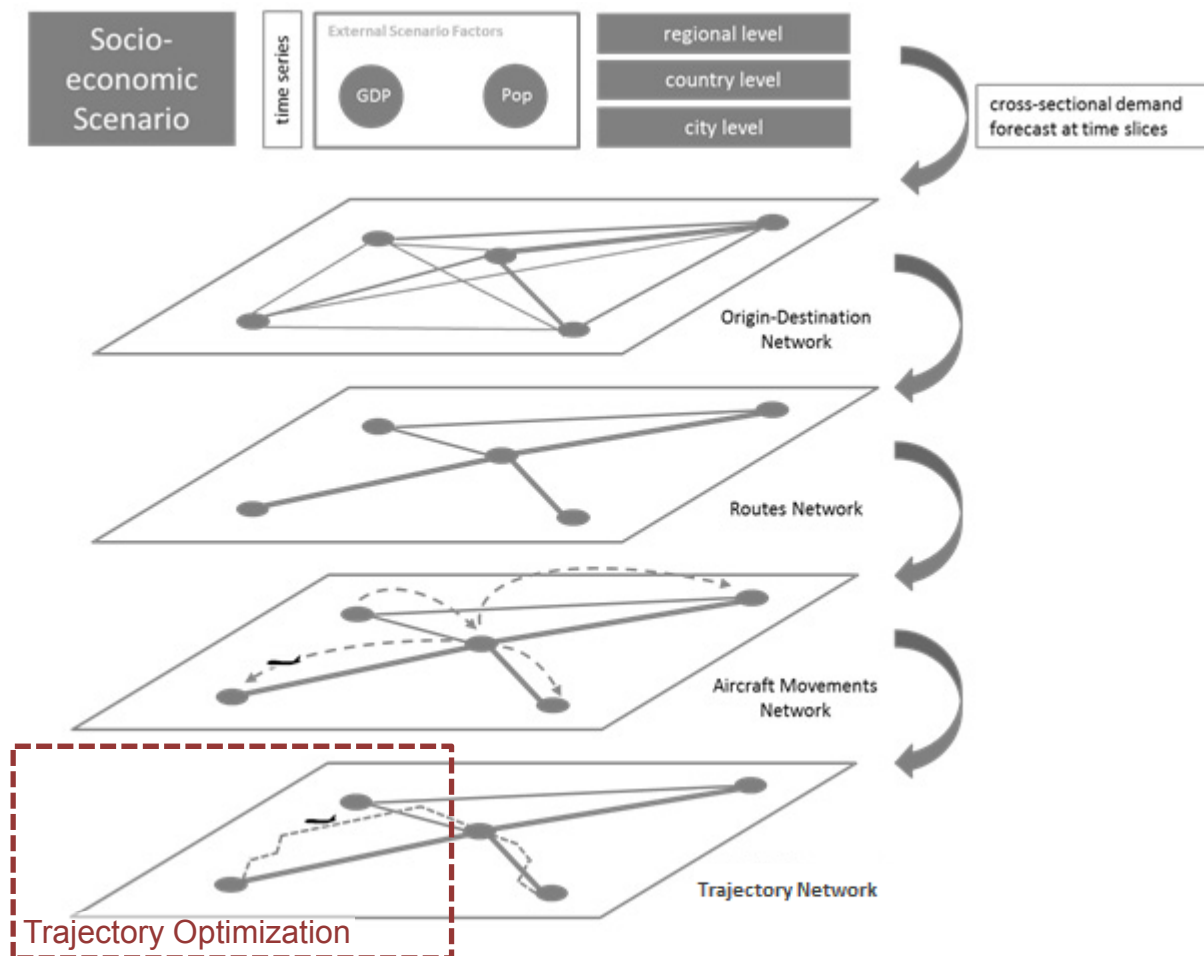


Knowledge for Tomorrow



Project WeCare

4 layers approach



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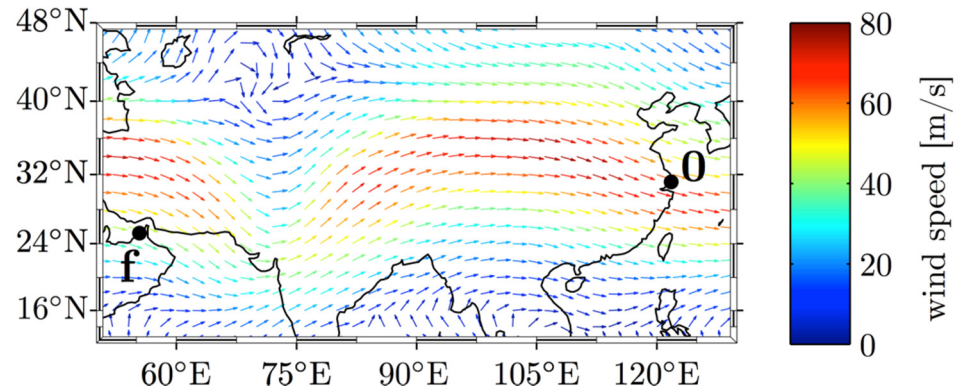
1. **Algorithm**
2. **Climate optimized trajectories**
3. **Current work**



Definition of the optimization problem

Assumptions

- massless point
- $v_{TAS} = \text{const}$
- $H = \text{const}$ or $H_p = \text{const}$
- $H \ll R_E$
- flight path angle $\gamma \approx 0$
- stationary windfield
- flight direction controlled by heading angle χ_H



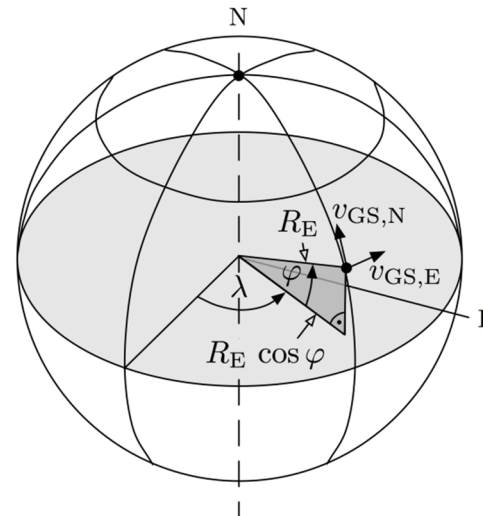
Equations of motion

$$v_{GS,E} = v_{TAS} \sin \chi_H + u_W(\lambda, \varphi)$$

$$v_{GS,N} = v_{TAS} \cos \chi_H + v_W(\lambda, \varphi)$$

$$\dot{\lambda} = \frac{v_{GS,E}}{R_E \cos \varphi} = \frac{v_{TAS} \sin \chi_H + u_W(\lambda, \varphi)}{R_E \cos \varphi}$$

$$\dot{\varphi} = \frac{v_{GS,N}}{R_E} = \frac{v_{TAS} \cos \chi_H + v_W(\lambda, \varphi)}{R_E}$$



Definition of the optimization problem

Optimal control problem

minimize

$$\begin{aligned}\mathcal{J} &= \int_{t_0}^{t_f} [c_t + c_\Psi \cdot \Psi(\lambda, \varphi)] dt \\ &= c_t \cdot \underbrace{(t_f - t_0)}_{\text{flight time}} + c_\Psi \cdot \int_{t_0}^{t_f} \underbrace{\Psi(\lambda, \varphi)}_{\text{penalty function}} dt\end{aligned}$$



Minimum Principle

(Pontryagin et al., 1967)

subject to

$$\dot{\lambda} = \frac{v_{GS,E}}{R_E \cos \varphi} = \frac{v_{TAS} \sin \chi_H + u_W(\lambda, \varphi)}{R_E \cos \varphi}$$

$$\dot{\varphi} = \frac{v_{GS,N}}{R_E} = \frac{v_{TAS} \cos \chi_H + v_W(\lambda, \varphi)}{R_E}$$

and

$$\lambda(t_0) = \lambda_0; \quad \varphi(t_0) = \varphi_0$$

$$\lambda(t_f) = \lambda_f; \quad \varphi(t_f) = \varphi_f$$

Two point boundary value problem

system of 3 differential equations

$$\begin{aligned}\dot{\chi}_H &= \frac{\partial u_W}{\partial \varphi} \cdot \frac{\sin^2 \chi_H}{R_E} - \frac{\partial v_W}{\partial \lambda} \cdot \frac{\cos^2 \chi_H}{R_E \cos \varphi} + \dots \\ &+ \left(\frac{\partial v_W}{\partial \varphi} - \frac{\partial u_W}{\partial \lambda} \cdot \frac{1}{\cos \varphi} \right) \cdot \frac{\sin \chi_H \cos \chi_H}{R_E} + \dots \\ (1) \quad &+ \frac{\tan \varphi \sin \chi_H}{R_E} \cdot (v_{TAS} + u_W \sin \chi_H + v_W \cos \chi_H) + \dots \\ &+ c_\Psi \left(\frac{\partial \Psi}{\partial \lambda} \cdot \frac{\cos \chi_H}{\cos \varphi} - \frac{\partial \Psi}{\partial \varphi} \cdot \sin \chi_H \right) \cdot \dots \\ &\cdot \frac{v_{TAS} + u_W \sin \chi_H + v_W \cos \chi_H}{R_E \cdot (c_t + c_\Psi \cdot \Psi)}\end{aligned}$$

$$(2) \quad \dot{\lambda} = \frac{v_{GS,E}}{R_E \cos \varphi} = \frac{v_{TAS} \sin \chi_H + u_W(\lambda, \varphi)}{R_E \cos \varphi}$$

$$(3) \quad \dot{\varphi} = \frac{v_{GS,N}}{R_E} = \frac{v_{TAS} \cos \chi_H + v_W(\lambda, \varphi)}{R_E}$$

boundary values

$$\lambda(t_0) = \lambda_0; \quad \varphi(t_0) = \varphi_0$$

$$\lambda(t_f) = \lambda_f; \quad \varphi(t_f) = \varphi_f$$

Solving the two point boundary value problem

Two point boundary value problem system of 3 differential equations

$$\begin{aligned} \dot{\chi}_H = & \frac{\partial u_W}{\partial \varphi} \cdot \frac{\sin^2 \chi_H}{R_E} - \frac{\partial v_W}{\partial \lambda} \cdot \frac{\cos^2 \chi_H}{R_E \cos \varphi} + \dots \\ & + \left(\frac{\partial v_W}{\partial \varphi} - \frac{\partial u_W}{\partial \lambda} \cdot \frac{1}{\cos \varphi} \right) \cdot \frac{\sin \chi_H \cos \chi_H}{R_E} + \dots \\ (1) \quad & + \frac{\tan \varphi \sin \chi_H}{R_E} \cdot (v_{TAS} + u_W \sin \chi_H + v_W \cos \chi_H) + \dots \\ & + c_\Psi \left(\frac{\partial \Psi}{\partial \lambda} \cdot \frac{\cos \chi_H}{\cos \varphi} - \frac{\partial \Psi}{\partial \varphi} \cdot \sin \chi_H \right) \cdot \dots \\ & \cdot \frac{v_{TAS} + u_W \sin \chi_H + v_W \cos \chi_H}{R_E \cdot (c_t + c_\Psi \cdot \Psi)} \end{aligned}$$

$$(2) \quad \dot{\lambda} = \frac{v_{GS,E}}{R_E \cos \varphi} = \frac{v_{TAS} \sin \chi_H + u_W(\lambda, \varphi)}{R_E \cos \varphi}$$

$$(3) \quad \dot{\varphi} = \frac{v_{GS,N}}{R_E} = \frac{v_{TAS} \cos \chi_H + v_W(\lambda, \varphi)}{R_E}$$

boundary values

$$\lambda(t_0) = \lambda_0; \quad \varphi(t_0) = \varphi_0$$

$$\lambda(t_f) = \lambda_f; \quad \varphi(t_f) = \varphi_f$$

Approach: solve a sequence of initial value problems

- Numerical integration of the system of differential equations (Dormand-Prince method)
- Choose initial values for λ and φ according to boundary values at starting point 0
 - Free initial value χ_{H0} must be chosen such that boundary conditions at ending point f are satisfied
 - Initial heading χ_{H0} is obtained iteratively using the shooting method

Contents

1. Algorithm
- 2. Climate optimized trajectories**
3. Current work



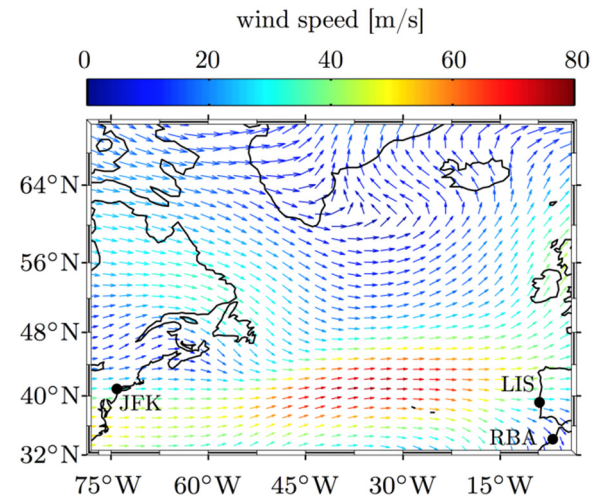
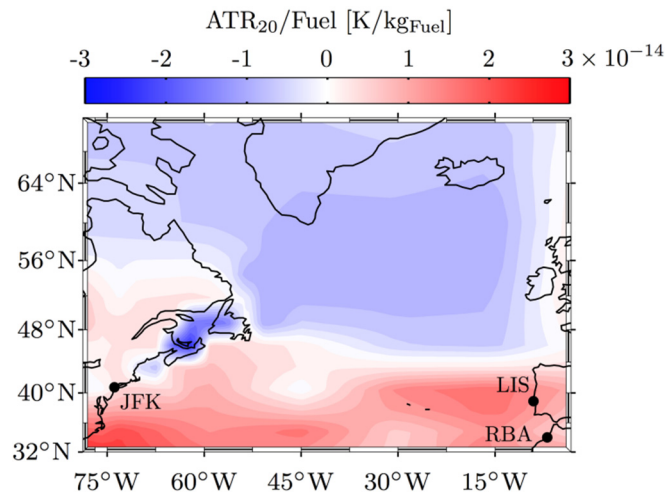
Climate optimized trajectories

Factors that contribute to aviation induced climate change

- Volume of emissions
 - Location of emissions
 - Local weather
- } technological/operational improvements
- } operational strategies (i.e. avoiding climate sensitive regions)

EU-Project REACT4C (Reducing Emissions from Aviation by Changing Trajectories for the benefit of Climate)

- Identification of typical weather situations (Irvine et al., 2013)
- Determination of climate cost functions for CO_2 , O_3 , CH_4 , H_2O , and aviation induced cirrus cloudiness (Grewe et al., 2014)
- Overall climate cost function (summation of all effects)



Climate optimized trajectories

Choice of the costfunction

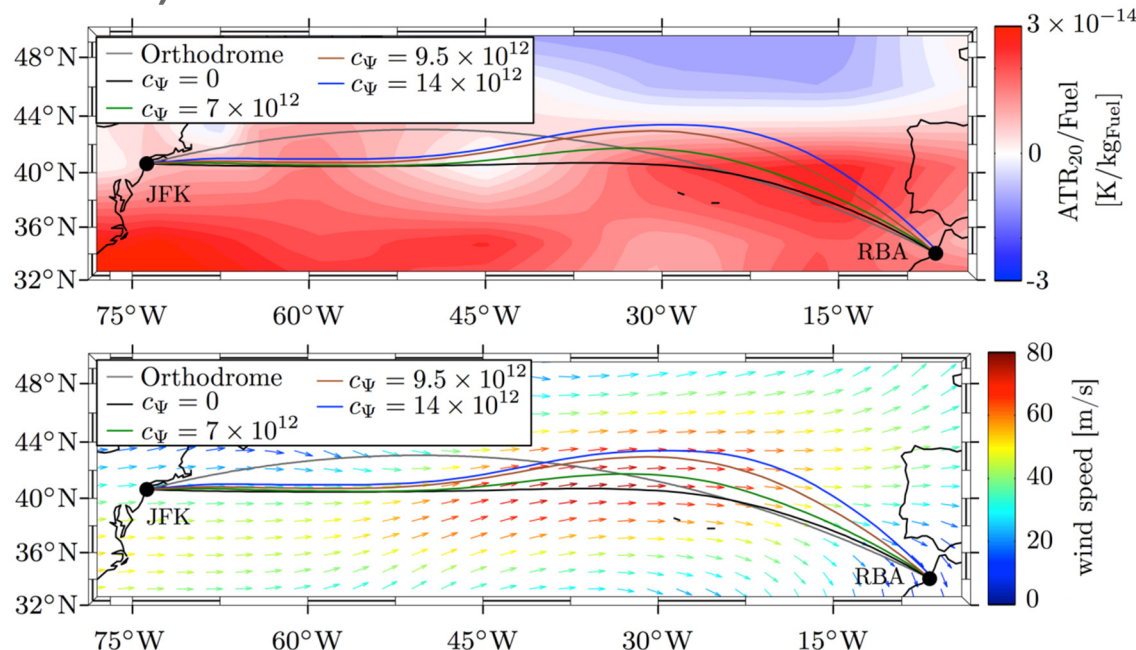
$$\mathcal{J} = \int_{t_0}^{t_f} [c_t + c_\Psi \cdot \overbrace{\text{FF} \cdot \text{CCF}_{\text{tot}}(\lambda, \varphi)}^{\text{ATR}_{20} \text{ per time}}] dt$$

$$= \underbrace{c_t \cdot (t_f - t_0)}_{\text{flight time}} + \underbrace{c_\Psi \cdot \int_{t_0}^{t_f} \text{FF} \cdot \text{CCF}_{\text{tot}}(\lambda, \varphi) dt}_{\text{ATR}_{20} \text{ of flight}}$$

Further assumptions

- $v_{\text{TAS}} = 247.85 \text{ m/s}$
- $H_p = 11278 \text{ m}$
- $\text{FF} = 1.51 \text{ kg/s}$ (A330-200, based on BADA 3.9)
- $c_t = 1$
- $c_\Psi = 0 \dots 1.4 \cdot 10^{13}$

Results (qualitative)



Climate optimized trajectories

Choice of the costfunction

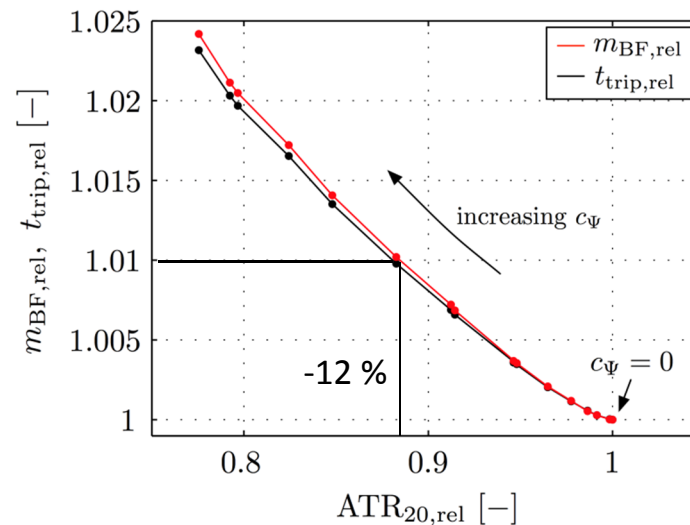
$$\mathcal{J} = \int_{t_0}^{t_f} [c_t + c_\Psi \cdot \overbrace{\text{FF} \cdot \text{CCF}_{\text{tot}}(\lambda, \varphi)}^{\text{ATR}_{20} \text{ per time}}] dt$$

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Results (quantitative)



New York - Rabat

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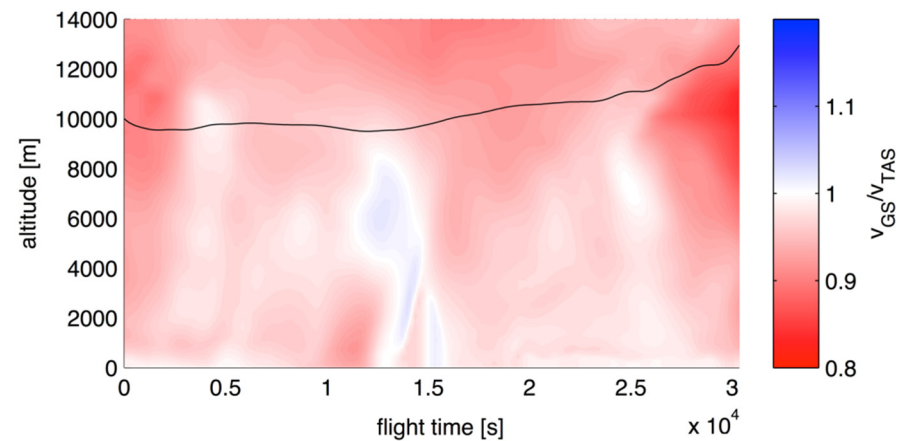
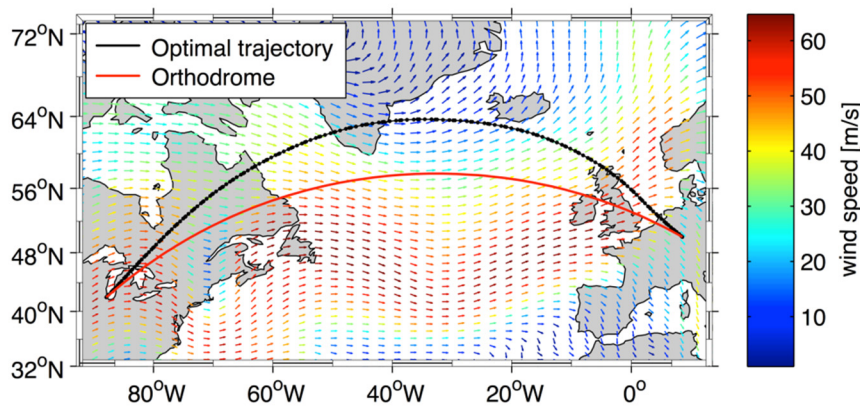
1. Algorithm
2. Climate optimized trajectories
3. Current work



Current work

Improvements

- Integration of performance calculations within the optimization loop
- Modeling of engine emissions
- Control variables: heading and thrust
- Variable altitude (3D-optimization)



Frankfurt – Chicago, minimum fuel trajectory, weather: January 1st, 2012

Thank you very much for your attention



Knowledge for Tomorrow



Application of the Column Generation Method for Large-Scale Air Traffic Flow Management scenarios

Jan Berling

NASA/DLR Virtual Institute Event

August 21, 2014

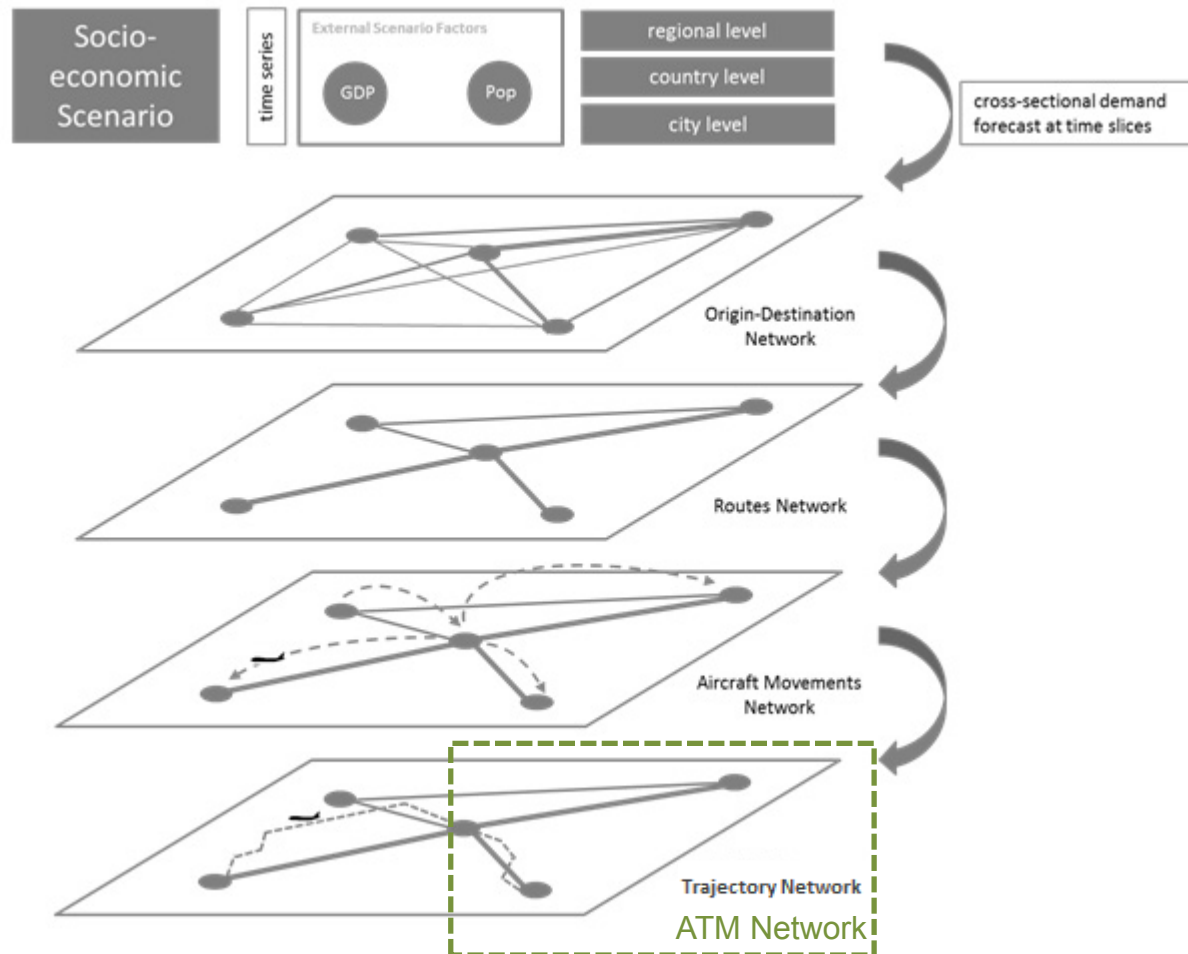


Knowledge for Tomorrow



Project WeCare

4 layers approach



Agenda

1. **NFE Network Flow Environment**
2. **Column Generation Approach**
3. **Exemplary Results**
4. **Outlook**



NFE – Network Flow Environment

What is NFE?

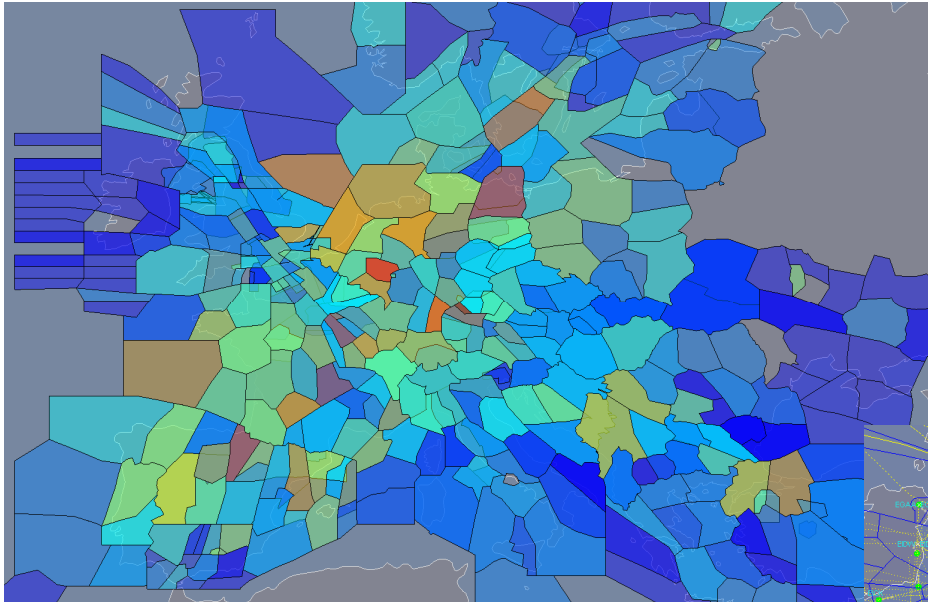
- DLR's **tactical Network Management** model developed since 2011
 - 1 Dissertation
 - 4 Diploma theses
 - 2 Publications
- Cooperation with TU Dresden (Traffic Flow Science)
- Model for the allocation of **ATFM departure slots** and **pre-flight re-routings**
- Binary Integer Programming (BIP) optimization algorithm

How is it actually applied?

- **Air Traffic Flow Management (ATFM) delay quantification** due to adverse capacity impact and initiated regulations
- Identification of **weather forecast benefits** within a network model
- Network performance quantification



NFE – Network Flow Environment

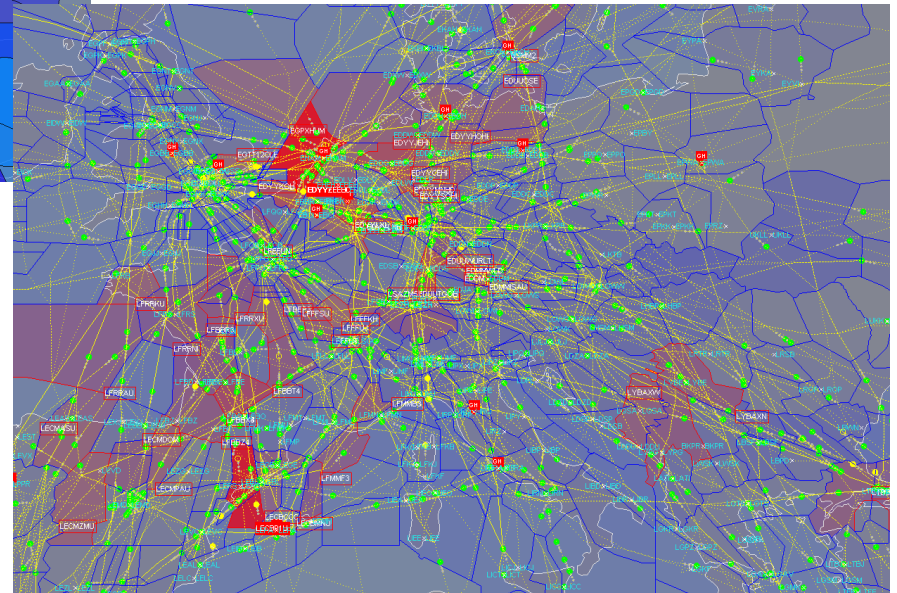


Air Traffic Management network

- lower and upper airspace
- sector network of ~640 Air Traffic Control (ATC) sectors
- around 20.000 airports
- underlying AIRAC cycle

Demand-Capacity-Balancing

- flight plan data from Network Manager
- network impact data, e.g. weather data
- qualitative mapping of network airport- and sector loads, network performance



Binary integer ATFM optimization problem

Cost Function

$$\min_x Z = \sum_{f \in F} \sum_{d \in D} \omega_{fd} x_{fd} \quad x_{fd} \in [0,1] \quad \text{Minimization of total system delay.}$$

Start Constraint

$$\sum_{d \in D} x_{fd} = 1 \quad \forall f \quad \text{Every flight departs exactly one time.}$$

Capacity Constraint

$$\sum_{f \in F} \sum_{d \in D} cto_{st}(f, d) \leq cs_t \quad \forall s, t \quad \text{Demand shall never exceed capacity.}$$

Agenda

1. NFE Network Flow Environment
2. Column Generation Approach
3. Exemplary Results
4. Outlook



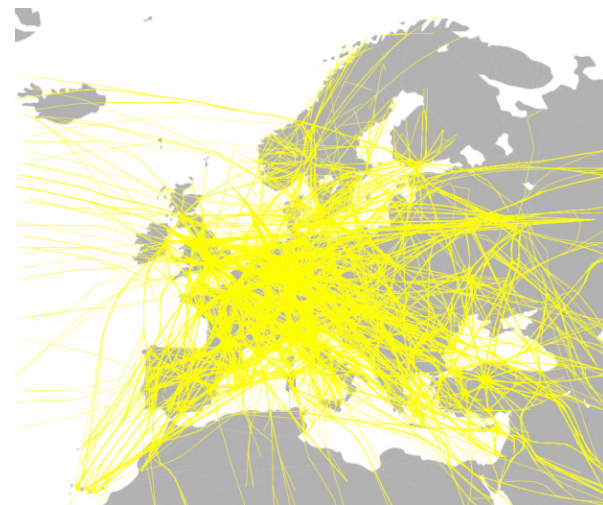
Large-scale ATFM scenario

Motivation for column generation application

- Total 1-day scenario
- Complete European Air Traffic
- Gigabytes of memory usage
- Weather impact data from different sources
- Regulation data for ATC sectors and airports
- First-Planned-First-Served (FPFS) start solution from a heuristic algorithm

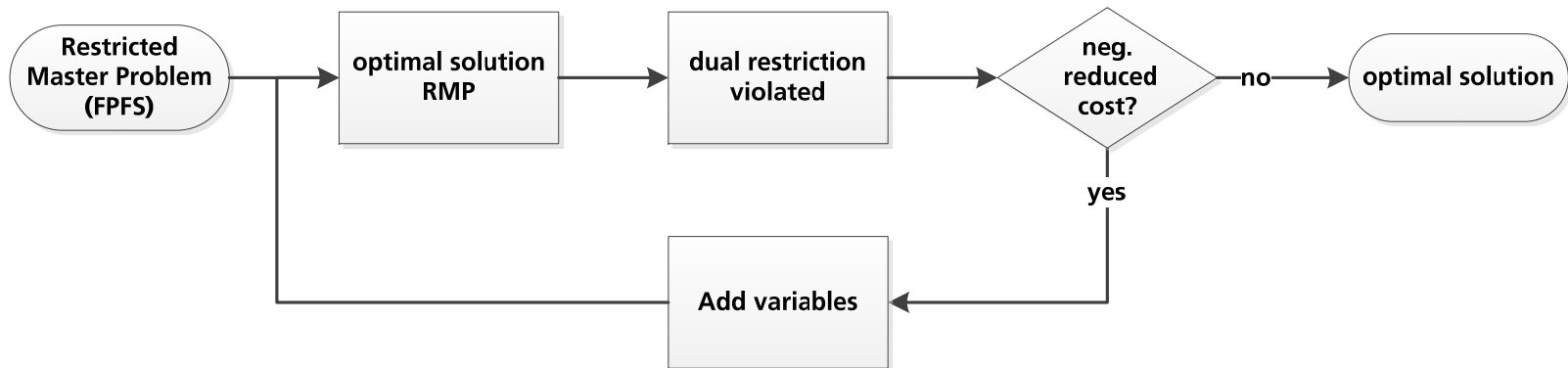
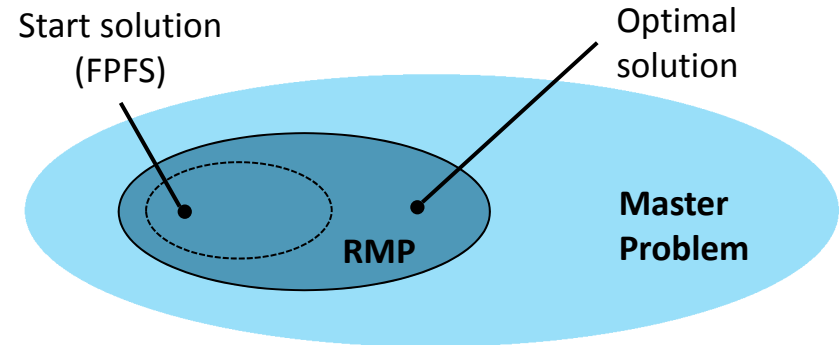
Timeframe	Total [min]	1440
	Granularity (Slot) [min]	15
Movements	Flight Movements [#]	28.784
ATC Sectors	Quantity [#]	638
Airports	Quantity [#]	19.798
Regulations	Sectors	85
	Airports	126

Exemplary 1-Day scenario



Column Generation optimization approach

- Iterative solution process
- Uses a Restricted Master Problem (RMP)
- Starts with a feasible solution (FPFS)
- RMP's are solved by simplex algorithm
- Adds variables with delay saving potential



Agenda

1. NFE Network Flow Environment
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Exemplary Results:

Total delay, calculation time and added variables

Total delay

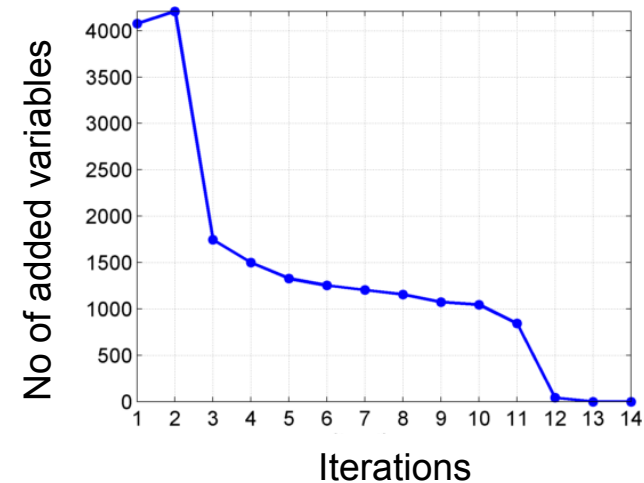
FPFS [min]	170.475
Optimum [min]	15.975
Col Gen relative [%]	9,4

Number of variables

Total [#]	272.624
Column Generation [#]	44.261
Col Gen relative [%]	16,2

Calculation time

Simplex [s]	1.436
Column Generation [s]	64,4
Col Gen relative [%]	4,5



Agenda

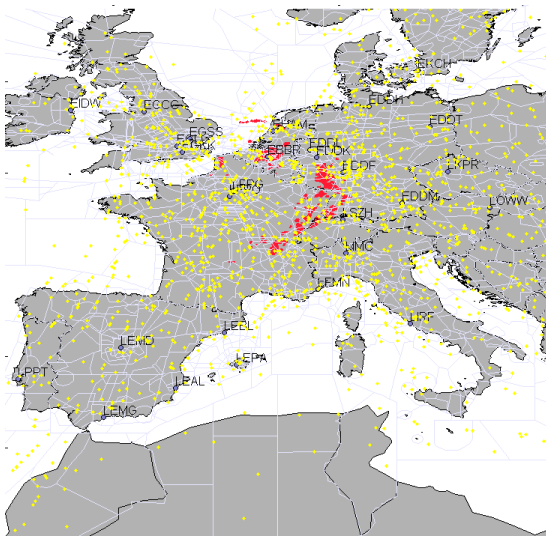
1. NFE Network Flow Environment
2. Column Generation Approach
3. Exemplary Results
4. Outlook



Outlook

NFE model improvements

- Integration of convective radar data within different forecast scenarios
- Application of the model within a *European volcanic ash crisis mitigation concept*
- Application of different routing concepts (free routing concepts)
- Integration of a Dynamic Airspace Management as an ATFM measure



Thank you very much for your attention!



Knowledge for Tomorrow

