FORESEE: Future proofing strategies FOr RESilient transport networks against Extreme Events H2020-MG-7-1-2017: Resilience to extreme (natural and man-made events)

-vergleich-in-bildern-362263069

An algorithm to determine optimal risk reducing intervention programs on networks

> Prof. Dr. Bryan T. Adey, Head of the Infrastructure Management Group, ETHZ



https://www.bernerzeitung.ch/janrhunderthochwasser-2005

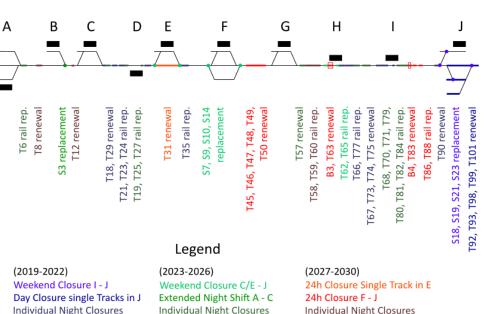
## Results

- I. Burkhalter, M., Moghtadernejad, S., Hackl, J., Toribio-Dia, C., Antonio Moli Diaz, A, Martani, C., Adey, B.T., Jimenez Redondo, N., Pardi, L., Di Gennaro, F., Robles Urquijo, I., Beltran-Hernando, I., (2020), Final versions of the algorithms to determine optimal restoration and risk reduction intervention programs for transportation networks, Deliverable 4.7, EU Grant number 769373, pages 171, DOI: https://doi.org/10.3929/ethz-b-000503443.
- 2. Burkhalter, M., Adey, B.T., (accepted in 2022), Digitalising the determination of railway infrastructure intervention programs: A network optimisation model, Infrastructure Systems.
- 3. Burkhalter, M., Adey, B.T., (published on-line 2021), Required accuracy of information when determining optimal railway intervention programs, Infrastructure Asset Management, DOI: https://doi.org/10.1680/jinam.20.00032
- 4. Burkhalter, M., Adey, B.T., (2021), Quantifying net-benefits of intervention programs to enable their digitalized generation, Infrastructure Asset Management, 8(3), pp. 141–154, DOI: 10.1680/jinam.20.00020
- 5. Burkhalter, M., Adey, B.T., (2020), Modelling the complex relationship between interventions, interventions costs and the service provided when evaluating intervention programs on railway infrastructure networks, Infrastructures, 5(12) 113, 20 pages, DOI:10.3390/infrastructures5120113.





- Plans risk reducing interventions on networks taking into consideration
  - The interventions required per asset according to their optimal individual life-cycles
  - The possible savings from grouping interventions
  - The restrictions on the possibility of executing multiple interventions at the same time
  - The traffic configurations that would need to be in place during a project with one or more interventions, and
  - The consequences of moving interventions away from their optimal time of execution, in terms of additional life cycle costs or increased risk.



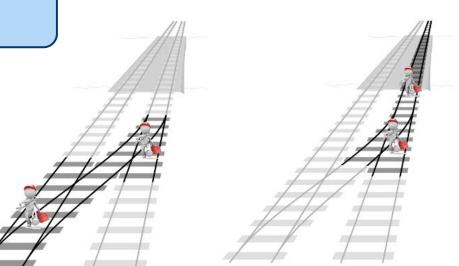


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Individual Night Closures

Optimal intervention program: Max Net Benefit = Benefit - Cost

- Benefit:
  - Reduction in future expected impacts
- Cost:
  - Impacts during the execution of the interventions







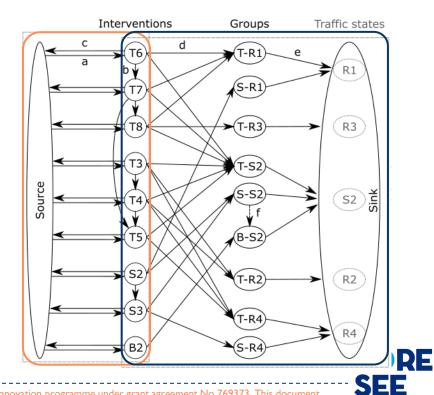
Mixed integer linear program

$$Max \ Z = \sum_{u \in V} \sum_{v \in V} \delta_{u,v} \cdot NB_{u,v} + \sum_{u \in V} \sum_{v \in V} \gamma_{u,v} \cdot NB_{u,v}$$

Asset level

Network level

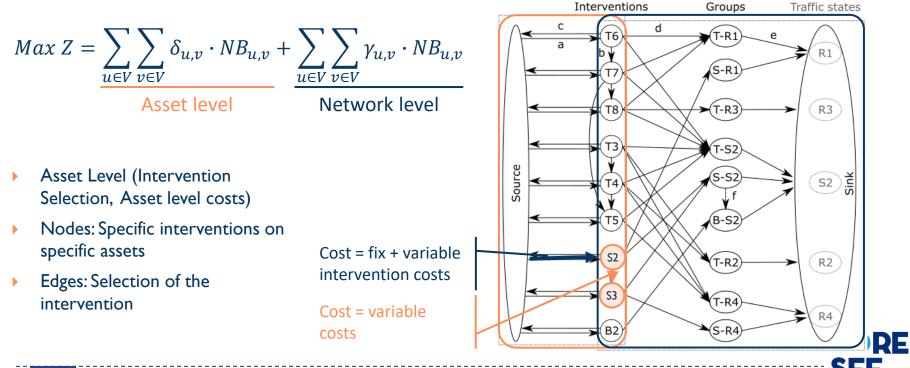
- Subject to:
  - Flow constraints
  - Capacity constraints
  - Budget constraint
  - Structural constraints



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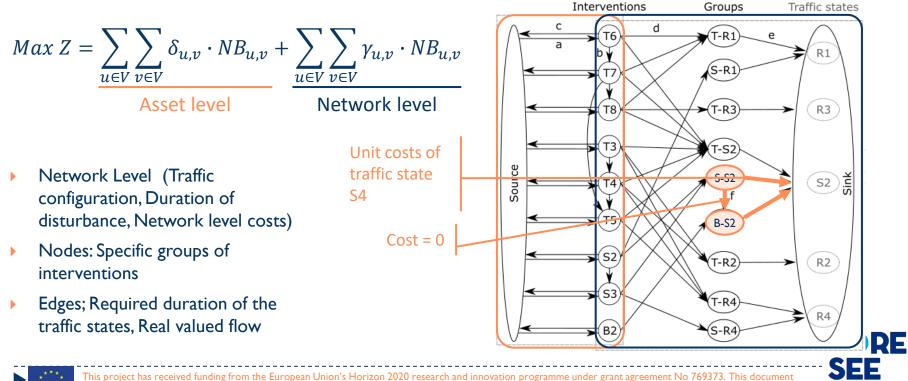
Mixed integer linear program



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Mixed integer linear program



Interventions

Groups

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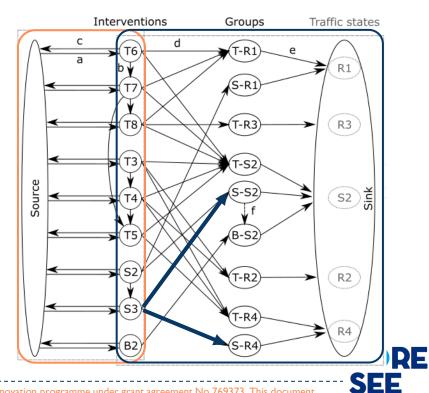
Mixed integer linear program

$$Max \ Z = \sum_{u \in V} \sum_{v \in V} \delta_{u,v} \cdot NB_{u,v} + \sum_{u \in V} \sum_{v \in V} \gamma_{u,v} \cdot NB_{u,v}$$
Asset level
Network level

Level interaction

 $\rightarrow$  Duration of selected interventions

$$\sum_{\mathbf{v}} \delta_{\mathbf{v}, u} \cdot d_u = \sum_{\mathbf{v}} \gamma_{u, \mathbf{v}}$$



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## Input data

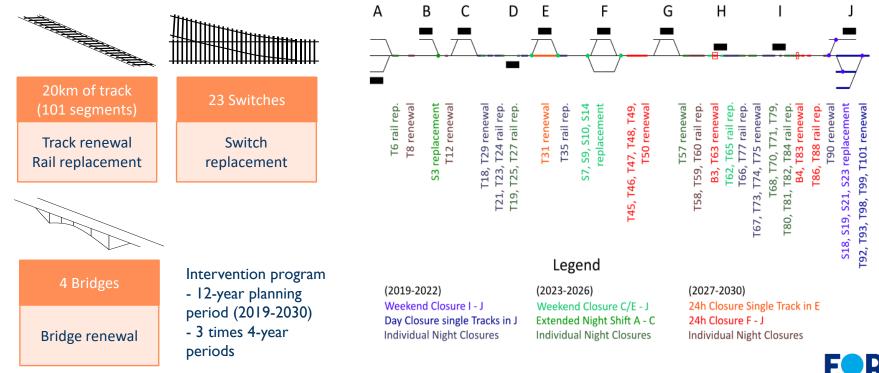
- Infrastructure assets, possible risk reducing interventions, variable unit costs and percentages of fixed costs per intervention,
- The type of intervals possible, the costs per traffic configuration, the number and type of vehicles using the lines
- The condition of the assets
- > The risk associated with each asset in each condition state
- The optimal asset level intervention strategies
- The possible type of restoration interventions, variable unit costs and percentages of fixed costs per intervention, amount of down time expected
- The consequences of failure for each type of asset, including approximate material damage, approximate number of injuries and fatalities







### Demonstration





#### Conclusion

- The algorithm can be used to plan risk reducing interventions on networks in a digital environment, taking into consideration all of the issues confronting infrastructure asset managers, including
  - The interventions required per asset according to their optimal individual life-cycles
  - The potential savings of grouping interventions together
  - Restrictions on the possibility of executing multiple interventions at the same time due to potentially large effects on the provided service, e.g. train movements
  - The traffic configurations that would need to be in place during a project with multiple interventions, and
  - The consequences of moving interventions away from their optimal time of execution, in terms of additional life cycle costs or increased risk.

#### Future proofing strategies FOr RESilient transport networks against Extreme Events

H2020-MG-7-1-2017: Resilience to extreme (natural and man-made events)

An algorithm to determine optimal restoration programs for transportation networks

Dr. Saviz Moghtadernejad, ETH Zurich



#### The <u>objective function</u> to be minimized is the costs:

- I. Direct costs of executing the physical interventions
- 2. Indirect costs due to the inadequate service by the network.

$$Min Z^{R} = \sum_{t \in T} \left[ \sum_{n \in N^{S}} \sum_{i \in I(n|s)} \delta_{n,i,t} \cdot C_{n,i} + \gamma \sum_{p \in P_{od}^{s \setminus g}, e \in P} \Pi(t|x_{e,t}) - \Pi^{0}(t|x_{e,0}) + \sum_{p \in P_{od}^{g}} \Lambda(t) \right]$$

Direct costs

Indirect costs

This is a bi-level optimization where the indirect costs depend on the link traffic flow:

$$x_{e,t} \in \min Z^T = \sum_{e \in \varepsilon^s} \int_0^{x_{e,t}} C_e^T(\omega) d_\omega$$

#### Subject to:

- Budget constraint
- Resource constraints
- Traffic flow constraints
- The number and type of interventions in a time period

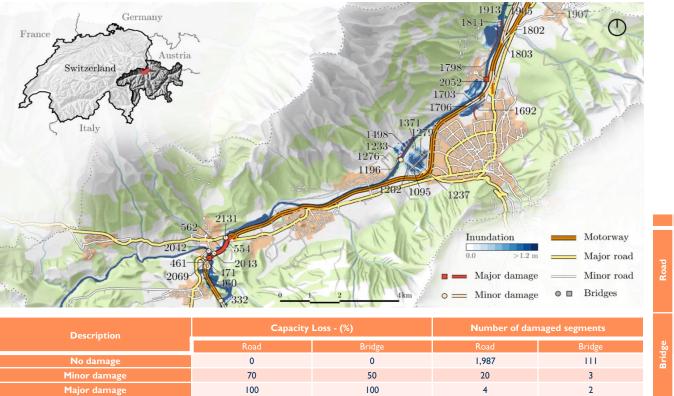


## Input data

- Infrastructure assets, and the GIS map of the network
- Origin-Destination matrix of the network
- > The condition of the assets, and their related capacity losses after the hazard
- The possible type of restorative interventions, variable unit costs and fixed costs per intervention, amount of down time expected
- The indirect costs per traffic configuration, the number and type of vehicles using the lines (mean fuel consumption, operation costs, etc.)
- Available budget and work crews



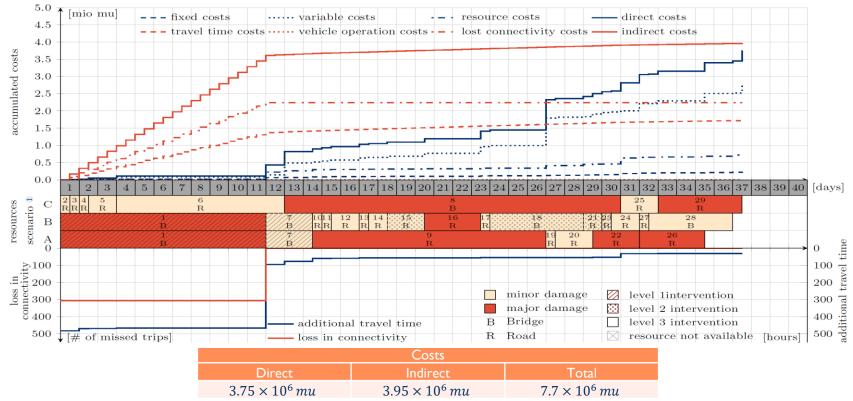
### Demonstration



	State	Priority	Recovery (%)
Road	Minor	High	100
		Normal	100
		Low	30
	Major	High	100
		Normal	100
		Low	10
Bridge	Minor	High	100
		Normal	100
		Low	20
	Major	High	100
		Normal	100
		Low	10



## Demonstration





#### Conclusion

- This algorithm has the flexibility to be used in real-world situations and for a variety of infrastructure types.
- It can provide estimations on the time required to restore the desired level of service following an extreme event.
- It can be used to identify objects whose failure will result in relatively large disruptions to service.
- It can identify when it is advantageous to execute less extensive interventions in the interest of speeding up the restoration of service.

