Predicting and Mitigating Emergency Situations on the roads: a Data-Driven Approach

Agenda

- The problem is ... 1
- Action Plan and approach 2
- It's all about data 3







What is the problem?

Frequent derailment incidents involving freight cars

High repair costs

Limited information on predictive factors

Impact of accidents on business operations, safety, and the environment

PURP	
Development	it of a solution
for predic	oting freight
car de	crailments
on tra	ock sections

Problem solving plan

Preparing of historical data and a target variable for analysis



Assessment of data quality for modeling

Mathematical model for predicting the probability of freight car falling based on data (and risk profiles)



Results of the project

Dependency detection between the facts of the freight car falling and the input characteristics of the wagons and railway infrastructure

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Risk Profiles assessment of freight car falling and risk reduction recommendations

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Parameters:

Data for the period: Jan 2016 – July 2019

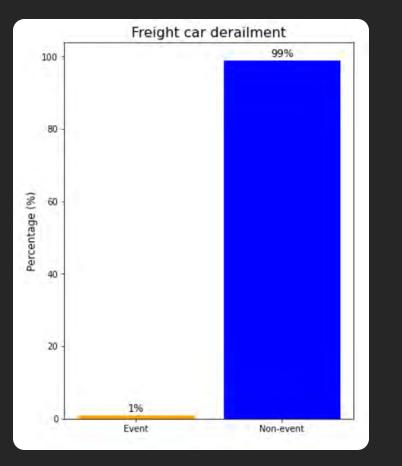
Data Category	Description	Source	
Location Data	Wagon linked to station and time	Location from ASOUP, dispatches from TSO-31	
Wagon Characteristics and Repair Data	Description of wagons and their repairs	ICH VP. Repairs	
Derailment Data	List of derailments, enriched with infrastructure and wagon characteristic data	Derailment register from Excel by TCS	••••••••••••
Location Data	Wagon linked to track	ASOUP-2V	
Track Incident Data	List of track malfunction data, date of resolution, type of malfunction	ASUI	
Repair Schedule Data	List of repair windows	APVO	
Defect Drawing Data	Detailed description of defects	ASUI	
Manifest Data	List of wagons in the train, wagon weight and cargo	ASOUP-2V	
Wagon Passport Data	List of detailed technical characteristics	ABD PV	
Speed Restriction Data	List of imposed restrictions and reasons for their implementation	ASUVOP 2	
Weather Data	Weather conditions	System with data from weather stations	
Station Coordinates Data	Geographical location of the station	GIS Railways	





The challenge №1:

1% event to 99% non-event



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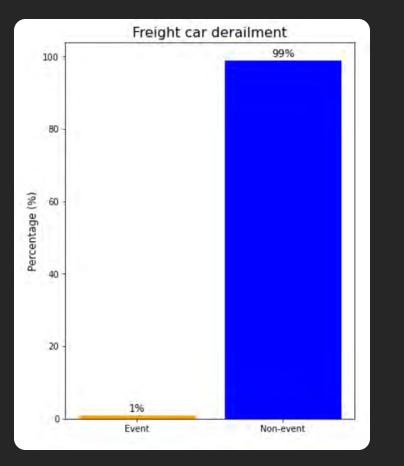
2 approaches — Reduce the number of non-events (we are losing information)

Bad practice

Model

The challenge №1:

1% event to 99% non-event

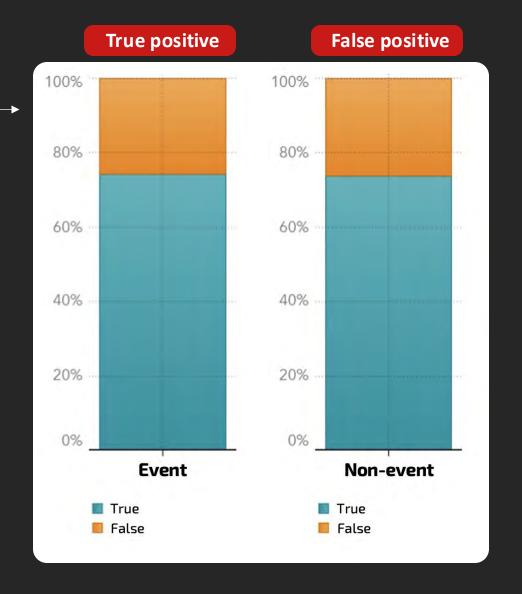


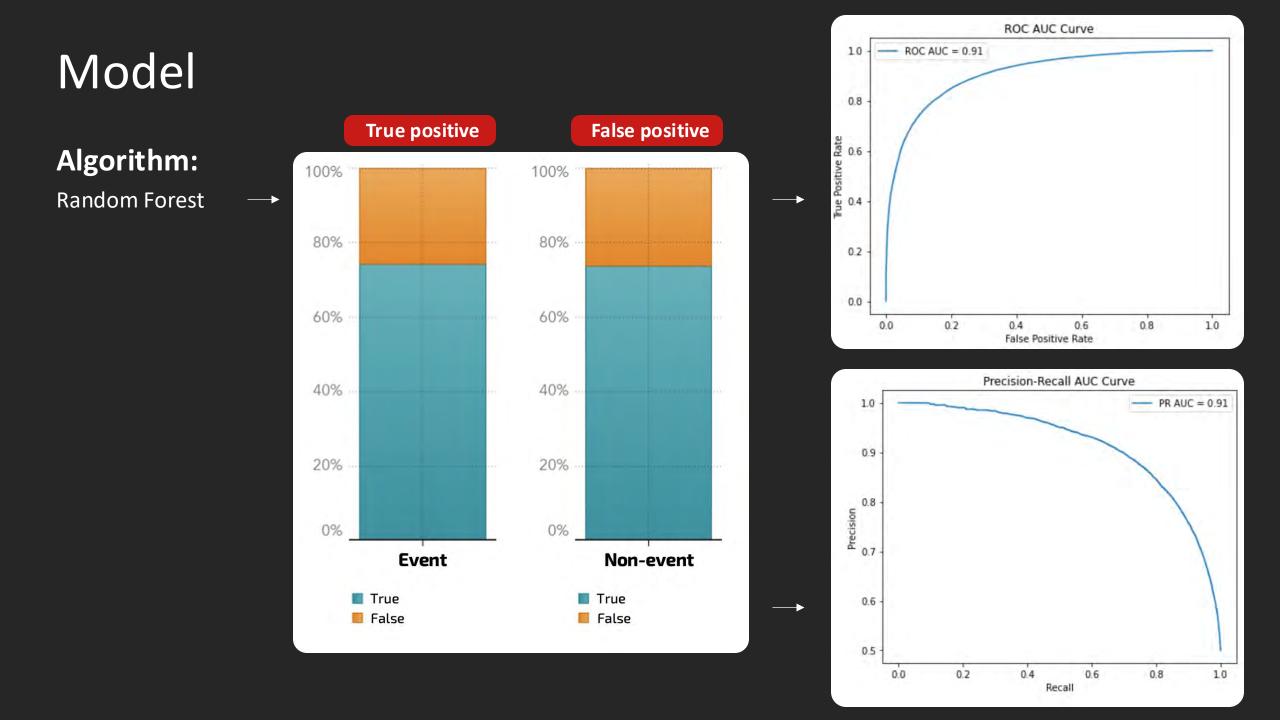
2 approaches Reduce the number of non-events Bad practice
Dealing with class imbalance
Best idea
Best idea
Will allow us to better account for our
events in the prediction model

Model

Algorithm:

Random Forest





Probability of freight car derailment

Based on the probability, we could select the most "high-risk" freight cars that require additional inspection

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§ Результаты анализа 8/26/2020 🗙

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	Results of analysis								2 🗉 🛇	* 🖻 ארי אין	
lodel Accuracy	v Evaluation								0.60		
Wagon ID	Wagon Model	Region	Road	Current Station	Previous Station	Probability of Derailment 🔻 Event	t / Non-event	D.00		1,00	
24421554	11-217	Южный	C-KAB	ГРАЧИ	БЕЛАЯ КАЛИТВА	0.6594310339	1	And the second	and a second		
62546569	12-1302	Волжский	Ю-УР	ЧЕЛЯБИНСК-ГЛАВНЫЙ	КИСЕГАЧ	0.6568745665	1	100%	100%		
24026882	11-217	Восточный	В-СИБ	АНГАСОЛКА	СЛЮДЯНКА II	0.6552774847	1				
52111952	11-1807	Волжский	Ю-УР	БЕРДЯУШ	КРОПАЧЕВО	0.6548417112	1	80%	80%		
24480766	11-217	Южный	C-KAB	ТИМАШЕВСКАЯ	БАТАЙСК	0.65157868	1	00.0	00.0		
95261715	19-7053-02	Волжский	Ю-УР	КРОПАЧЕВО	иглино	0.6374430049	1				
24566457	11-217	Восточный	В-СИБ	АНГАСОЛКА	СЛЮДЯНКА II	0.6349571651	1	60%	60%		
24568974	11-270	Восточный	В-СИБ	СЛЮДЯНКА ІІ	АНГАСОЛКА.	0.6312880598	1	9010			
95320701	19-3054-04	Южный	ПРИВ	ЕЛХОВКА	СУХОЙ КАРАБУЛАК	0.6307843157	1				
29767696	11-7038	Волжский	Ю-УР	златоуст	КИСЕГАЧ	0.6302977058	1	40%	40%		
52664737	11-280	Восточный	ЗАБ	ХАРАГУН	РОССОШЬ	0.6207333829	1				
54477385	13-401	Волжский	КБШ	АБДУЛИНО	САРАЙ-ГИР	0.6191433973	1				
90851874	16-3002-03	Волжский	Ю-УР	КРОПАЧЕВО	ИГЛИНО	0.6191288791	1	-20%	20%		
29046695	11-1807-01	Южный	C-KAB	ТИМАШЕВСКАЯ	БАТАЙСК	0.6188766269	1				
24217069	11-217	Волжский	КБШ	ДЕМА	чишмы	0.6176351838	1				
52401718	11-217	Волжский	КБШ	ЧЕРНИКОВКА	ДЕМА	0.6164444534	1	0%	0%		
52461084	11-217	Волжский	КБШ	ИГЛИНО	КРОПАЧЕВО	0.6149420286	1	Event	Non	-event	
52664992	11-280	Восточный	В-СИБ	СЛЮДЯНКА.11	КАНАШ	0.6117293848	1	True	True		
24579542	11-217	Северо-Западный	ГОРЬК	ПЫЧАС	КАРАМБАЙ	0.6099320931	1	False	False		
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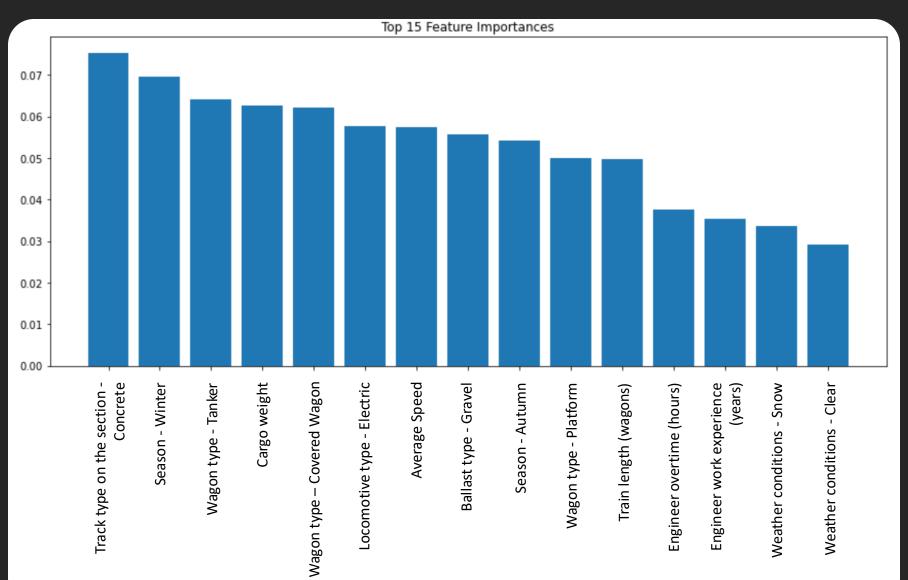
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Variable Importance

But these are still not risk profiles



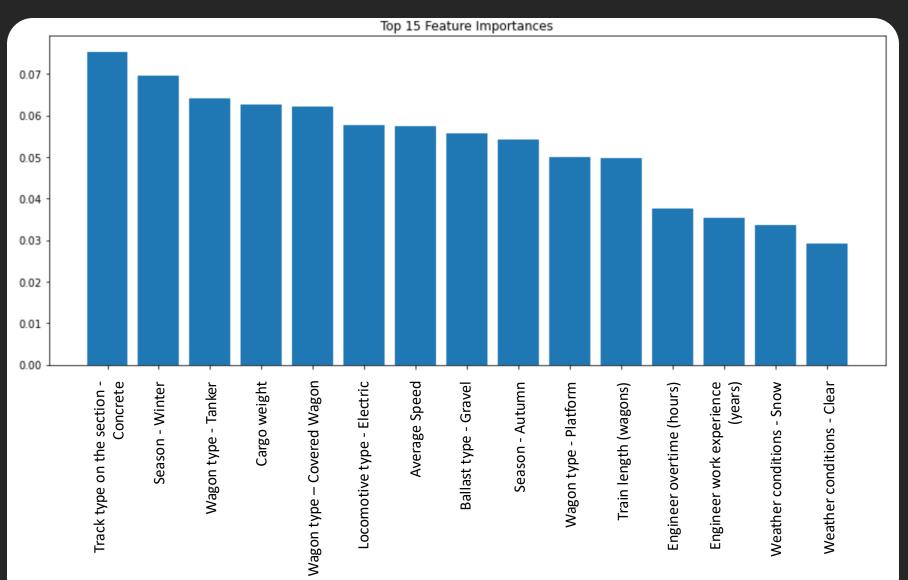


Variable Importance

The challenge №2

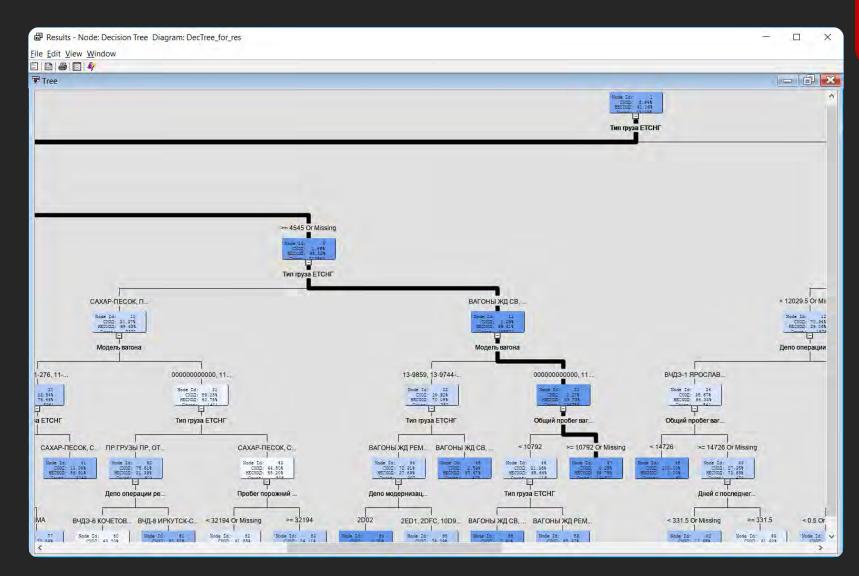
But these are still not risk profiles





Risk Profiles

Let's build a Decision Tree based on Random Forest results



OMG

Risk Profiles

IF Number of wagons < 38.5 AND Years from issue < 51.5 AND Ratio: speed on the last section to average speed < 1.0385 94,5% Probability (Share 21%)

IF Number of wagons >= 38.5 AND Type of dispatch «Group of Wagon» AND Condition (Wagon Weight) = Empty AND Train Weight >= 175425 AND Ratio: speed on the last hauls to average speed >= 0.94 78,6% Probability (Share 15%)

IF Остаток ресурса в км до нормы >= 1014.5 AND Кол-во вагонов >= 38.5 AND Type of dispatch «Group of Wagon» AND Condition (Wagon Weight) = With Cargo AND Days to scheduled repairs >= 437.5 99,9% Probability of **not fall** (Share 47%)

Project Results





Minimizing the risks of environmental disasters





