



UNIVERSITAT DE
BARCELONA

*Number of claims and number of
near-misses for **telematics pricing**
in automobile insurance*

Montserrat Guillen

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Fundación

BBVA



Contents

1. Introduction (3 minutes)
2. Data & methods (5 minutes)
3. Results (4 minutes)
4. Conclusions (3 minutes)

Companies selling motor insurance based on telematics around the world

Postnet

Play a low base rate
Then just pennies per mile

$\$29 + (150 \times 68) = \56

Make your offer

Detailed description: This is a screenshot of the Postnet website. It features a teal car partially covered by a white car cover. The main headline reads 'Play a low base rate Then just pennies per mile'. Below the car, a calculation shows a base rate of \$29 plus 150 miles multiplied by 68 cents per mile, resulting in a total of \$56. A 'Make your offer' button is visible at the bottom.

By Miles

Pay-by-mile car insurance for savvy drivers.

Get a quick quote

Detailed description: This is a screenshot of the By Miles website. The headline is 'Pay-by-mile car insurance for savvy drivers.' To the right, there is an illustration of several cars. A prominent blue button says 'Get a quick quote'. The text below the headline describes the policy as 'A simple, straightforward policy that better fits the way you live.'

verti

SEGURO DE COCHE POR KILOMETROS

CONTRATA TU SEGURO DE COCHE POR KILOMETROS Y NO PAGUES DE MÁS

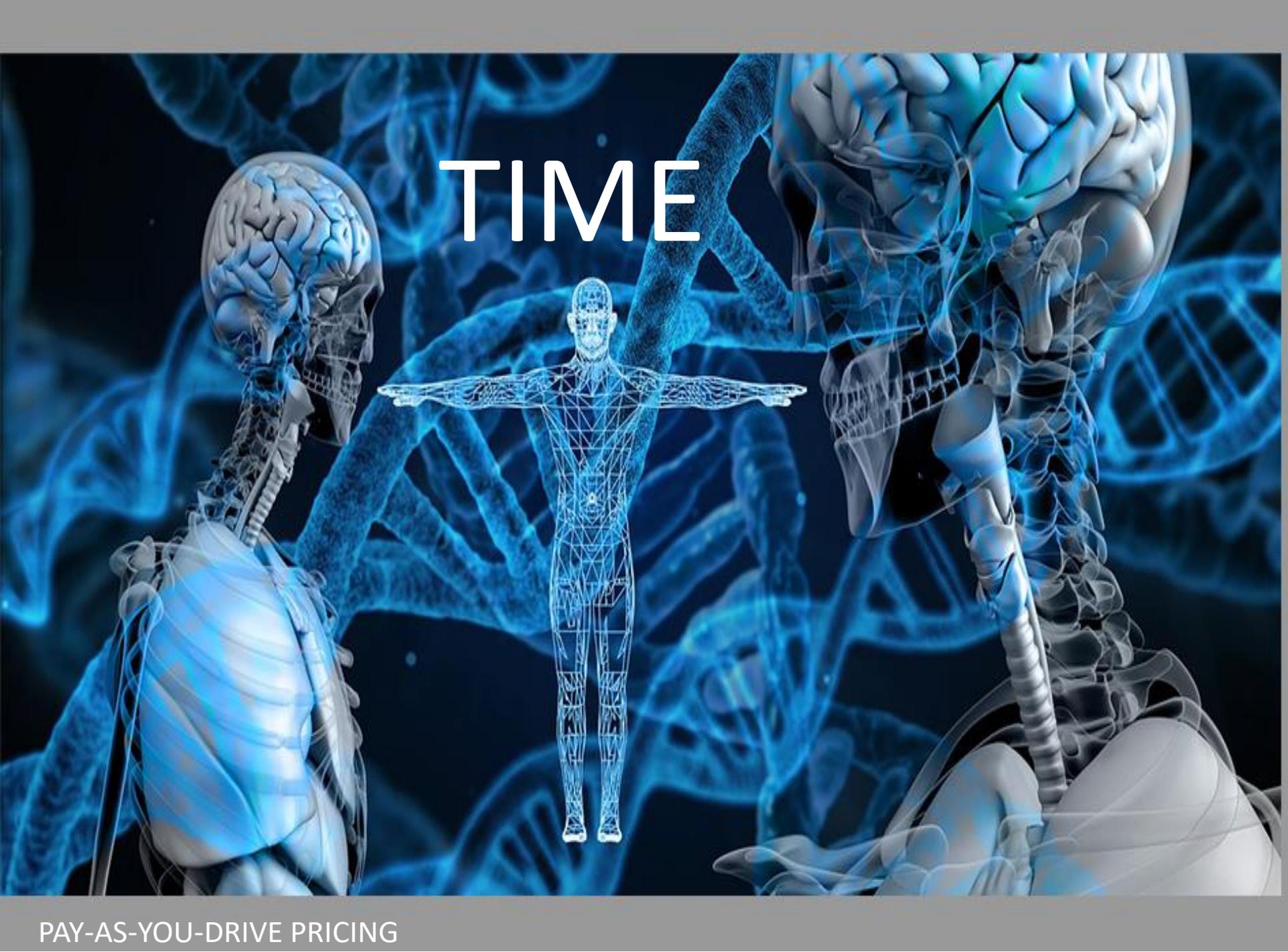
Detailed description: This is a screenshot of the Verti website. The main headline is 'SEGURO DE COCHE POR KILOMETROS'. The background image shows a hand holding a smartphone in front of a car's dashboard. At the bottom, there is a call to action: 'CONTRATA TU SEGURO DE COCHE POR KILOMETROS Y NO PAGUES DE MÁS'.

Generali

Generali Sei in Auto PPU

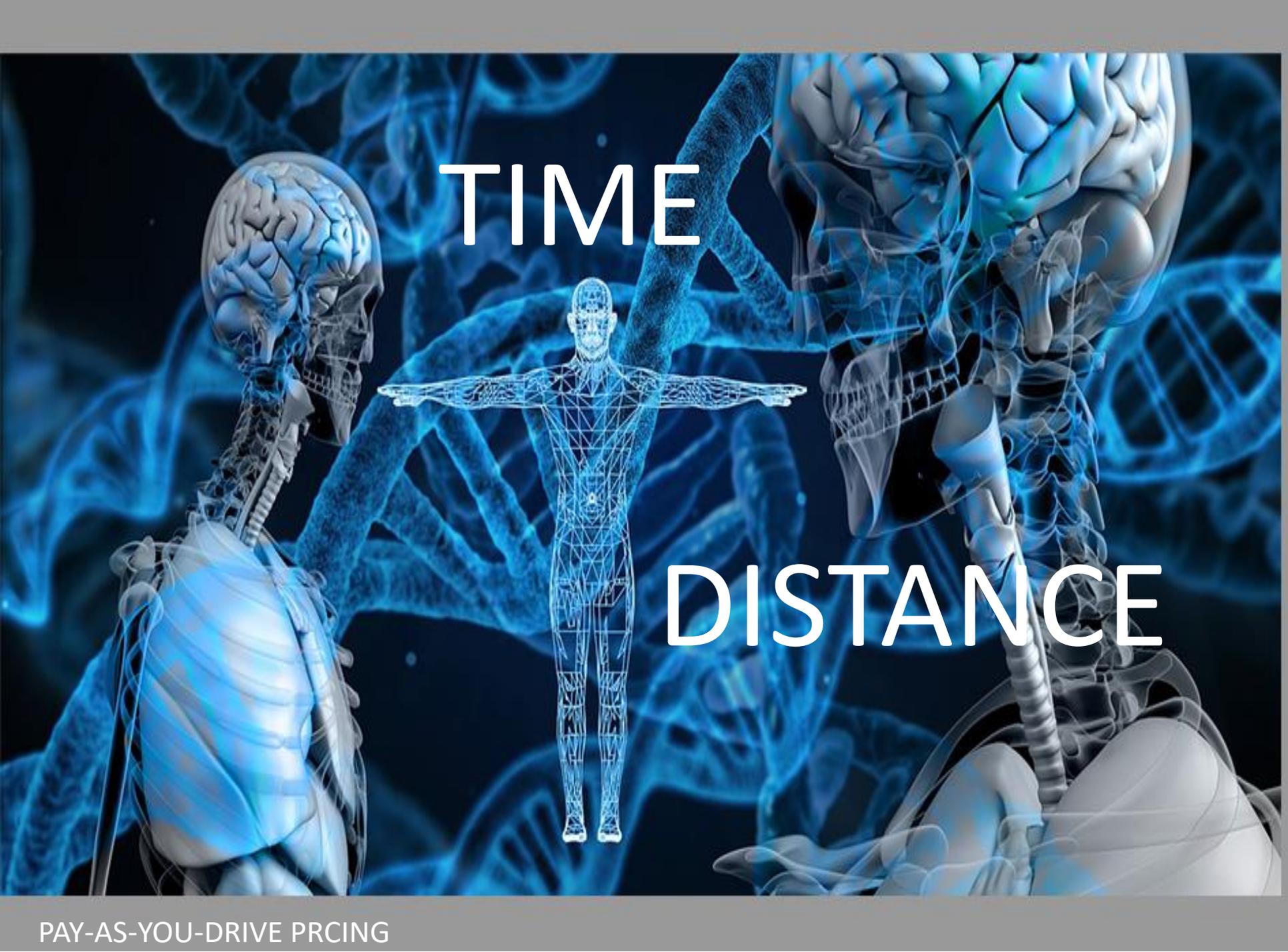
Detailed description: This is a screenshot of the Generali website. The main headline is 'Generali Sei in Auto PPU'. The background image shows a hand on a steering wheel. The page includes a navigation menu at the top and a list of features or benefits at the bottom.

PAY-AS-YOU-DRIVE PRICING = BASE PREMIUM + DISTANCE * COST per UNIT

A blue-toned 3D anatomical illustration of the human body. In the center is a wireframe figure of a human standing with arms outstretched. To the left is a detailed view of a human brain and skull. To the right is a detailed view of a human hand and forearm. In the background, there are faint, glowing DNA double helix structures. The word "TIME" is written in large, white, sans-serif capital letters across the upper middle of the image.

TIME

PAY-AS-YOU-DRIVE PRICING



TIME

DISTANCE

Telematics data: early 2000s

Total Distance
Km. Night
Km. Excess speed
Km. Urban

Telematics data: 2020+



[MORE DETAILS](#)



Super Easy Octo Telematics

2. DATA & METHODS

How do raw telematics data look like?

The image shows a screenshot of Microsoft Excel with the 'Datos' (Data) tab selected. The spreadsheet contains raw telematics data with columns for ID, GPS.time, Accessories.s, Longitude, Latitude, GPS.speed.kn, X1.ECU.fuel.c, X2.Integral.fu, and Current.fuel.c. The data is organized in a table with 17 rows and 9 columns.

	B	C	D	E	F	G	H	I	J	K
1	ID	GPS.time	E	Accessories.s	Longitude	Latitude	GPS.speed.kn	X1.ECU.fuel.c	X2.Integral.fu	Current.fuel.c
2	14705559598	27/06/2018 0:00	#	on	120048127	29481834	77	22784	8268.6	NA
3	14705559598	27/06/2018 0:01	#	on	120045677	29476968	62	22784	8268.68	NA
4	14705559598	27/06/2018 0:01	#	on	120043469	29471833	74	22784	8268.85	NA
5	14705559598	27/06/2018 0:02	#	on	120041206	29466663	70	22784,5	8268.99	NA
6	14705559598	27/06/2018 0:02	#	on	120038345	29461290	77	22784,5	8269.03	NA
7	14705559598	27/06/2018 0:03	#	on	120034947	29456416	70	22784,5	8269.12	NA
8	14705559598	27/06/2018 0:03	#	on	120032211	29451699	70	22784,5	8269.24	NA
9	14705559598	27/06/2018 0:04	#	on	120030847	29445850	81	22784,5	8269.4	NA
10	14705559598	27/06/2018 0:04	#	on	120029000	29440235	66	22785	8269.48	NA
11	14705559598	27/06/2018 0:05	#	on	120026396	29435530	66	22785	8269.49	NA
12	14705559598	27/06/2018 0:05	#	on	120024521	29430690	70	22785	8269.62	NA
13	14705559598	27/06/2018 0:06	#	on	120024579	29425772	61	22785	8269.72	NA
14	14705559598	27/06/2018 0:06	#	on	120026846	29420668	77	22785	8269.84	NA
15	14705559598	27/06/2018 0:07	#	on	120031278	29416523	64	22785.5	8269.97	NA
16	14705559598	27/06/2018 0:07	#	on	120035508	29413164	66	22785.5	8270.11	NA

Raw data file. Sun et al. (2020)

DATA IN MOTOR INSURANCE, EVERY 30''

What is the problem with the low frequency of claims?

- Telematics data have detailed information:
 - Years > Months > Weeks > Days > Hours > Minutes > Seconds
 - But the **accident event phenomenon is unlikely.**
 - $1/10 > 1/120 > 1/480 > 1/365 > 1/87720 > 1/5263200 > 1$
accident every 315 millions of seconds
- Statistical methods for rare events
 - “Excess of zeros”
 - *Corrections for imbalanced classes: SMOTE*
 - Rarity

What is a *near-miss*?

- A **near-miss** is a term borrowed from aviation safety – a situation in which **an accident is narrowly avoided**, such as when a driver brakes suddenly in order to avoid a crash (Arai et al., 2001).

Near-misses (or incidents) have been shown to be **correlated** with claims in auto insurance

*Ma, Y. L., Zhu, X., Hu, X. and Chiu, Y. C. (2018). The use of context-sensitive insurance telematics data in auto insurance ratemaking, **Transportation Research Part A** 113, 243–258.*

Guillen et al. (2021) Near-miss telematics in motor insurance. ***Journal of Risk and Insurance*** <https://onlinelibrary.wiley.com/doi/epdf/10.1111/jori.12340>

Examples: near-misses

- **Acceleration:** $>6\text{m/s}^2$, (Hynes & Dickey, 2008).
- **Braking:** $<-6/\text{s}^2$
- **Dangerous Turns:** speed combined with angle
- **Use of smart phone while driving**

North American Actuarial Journal (2019) we proposed modeling *near-miss events*

Problem: *(at fault near-misses?)*

Key concept that changes automobile insurance pricing

- Frequency and severity of **claims**
- **New approaches that take into account "near-miss" incidents**
 - Frequency models :
 - Claims (too rare)
 - Near-misses (difficulty to price them)
 - Claims (with near-misses as inputs)

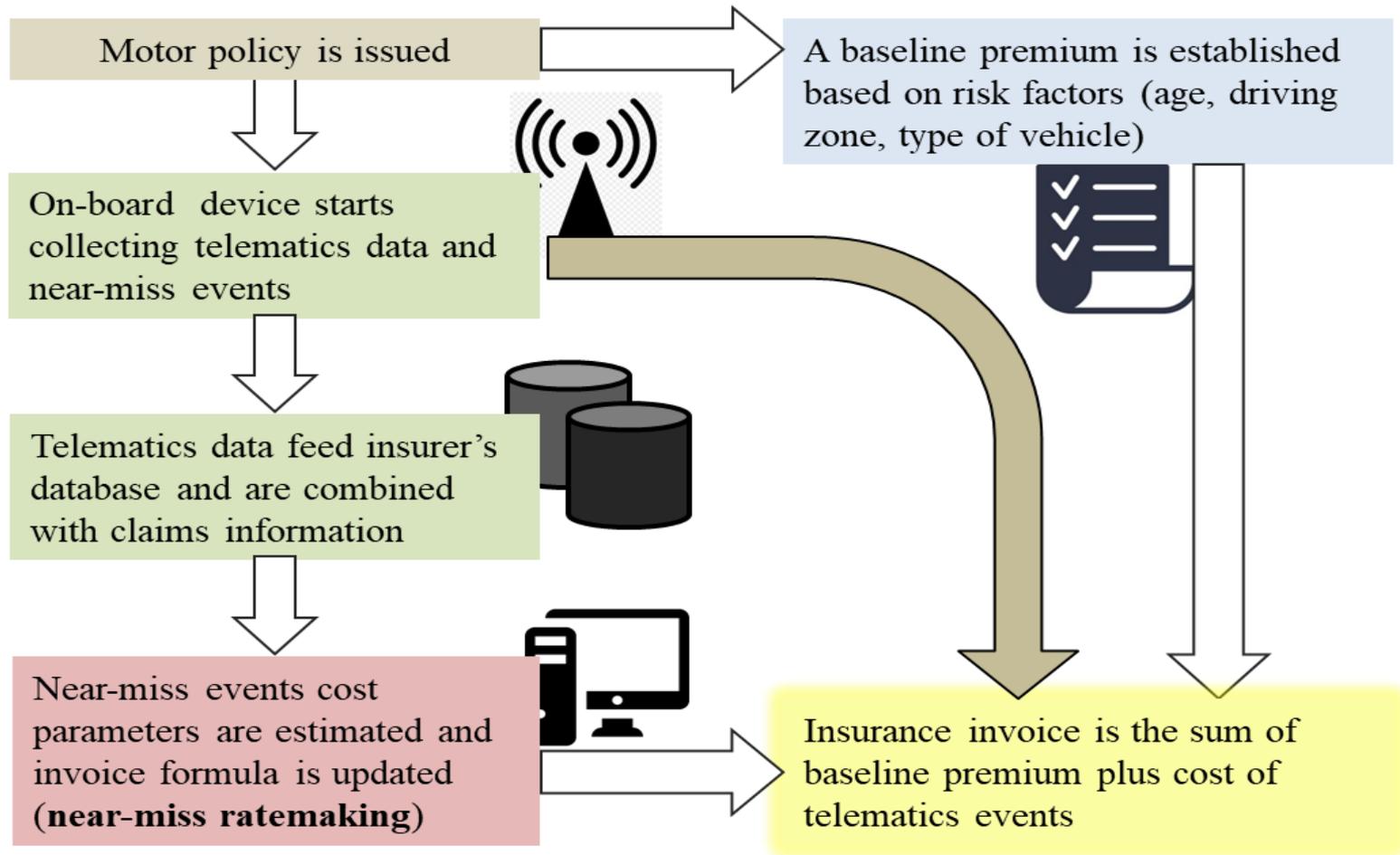
New models *Number of CLAIMS where near-misses are used as risk factors, i.e. new pricing tools*

In **Journal of Risk and Insurance** (2021) we have proposed using *near-miss events as covariates for pricing*

- **Acceleration: $>6\text{m/s}^2$**
- **Braking: $<-6\text{/s}^2$**
- **Use of smart phone while driving**

Near-miss telematics

motor insurance pricing



Notation and Poisson model specification

- Y_i number of claims at fault policy i ,
 $i = 1, \dots, n$
- T_i risk exposure, offset for policy i
- x_i, E_i ratemaking factors (traditional, telematics)

$$\begin{aligned} E(Y_i | x_i, E_i) &= T_i \exp(x_i' \beta + E_i' \alpha) \\ &= T_i \exp(x_i' \beta) \exp(E_i' \alpha) \end{aligned}$$

3. RESULTS

- SPANISH YEARLY DATA (MAPFRE):
 - Number of claims (what happens when we introduce telematics informaton?) PAYD/PHYD scheme.
 - Distance driven (can we identify factors that affect exposure?)
 - Percentile charts (can we score drivers?)
- CYPRUS DATA (EMERGENT):
 - Near-misses are correlated with Claims
 - Near-miss telematics pricing
- SPANISH TRIP DATA:
 - Finding patterns for near-misses
- CHINESE TICK DATA:
 - Driver's ECG. (New concept simliar to eletrocardiogram)

Near-miss telematics

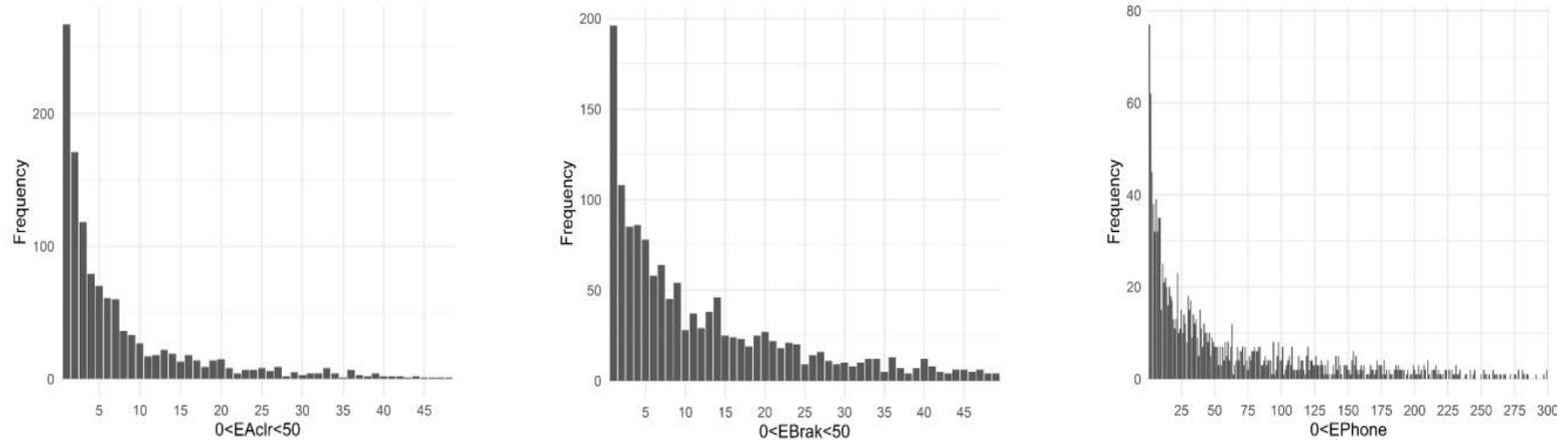


Figure 1, 2 and 3 show the histogram of EBrak, EAclr and EPhone. Due to the large frequency of zeroes we decided to remove them from the graphs, therefore only positive observations are represented. The data present a long right tail, so we also decided to limit the representation up to a maximum value, specifically 50 for EAclr and EBrak, and 300 for EPhone. Note that EAclr has 83.66% of zeroes, and 0.62% are equal or greater than 50. EBrak has 80.91% of zeroes, and 0.82% are equal or greater than 50, and finally EPhone has 79.78% of zeroes and 0.65% are higher than 300.

Claims frequency using near-miss events as covariates

Table 3. Parameter estimates of the Poisson model of the weekly rate of at fault claims for the telematics and claims data set

Parameter	Estimate	Standard Error	p-value
Intercept	-8.0637	0.0673	<.0001
EAc1r1	-0.0825	0.0265	0.0019
EAc1r2	0.3069	0.1277	0.0162
EAc1r3	0.0095	0.0390	0.8072
EBrak1	0.0268	0.0086	0.0018
EBrak2	-0.4966	0.0770	<.0001
EBrak3	0.0984	0.0336	0.0034
EPhon	0.0004	0.0002	0.0776
EngineCapacity	0.3644	0.0287	<.0001

The AIC equals 7345.00 and the BIC equals 7407.39. The pseudo-R² equals 21.83%.

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Near-miss telematics ratemaking

- Basic rate plus additional cost of near misses.

Table 1. Weekly breakdown of a total bill per week. Pure premium in motor insurance as a function of near-miss events for a driver of a car with engine capacity 1,769 cc. Basic weekly rate 1.95 Eur.

Week	Distance driven (km)	Number of near-miss brakes (a)	Number of near-miss accelerations (b)	Minutes of smart phone use (c)	Cost of near-misses (Eur) (d)	Bill per week (Eur) (e)
1	30	0	0	0	0,00	1,95
2	73	0	0	2	0,37	2,32
3	104	2	2	2	6,59	8,54
4	260	6	2	1	9,40	11,35
5	705	19	4	21	27,51	29,46

Total bill for five weeks: 53.61 Eur.

(e)=1.95+(d)

(d)=0.75(a)+2.36(b)+0.18(c)

Near-miss telematics ratemaking

- Basic rate with a reward for safe driving and additional charge for near misses.

Table 2. Weekly bill of pure premium in motor insurance as a function of near-miss events for a driver of a car with engine capacity 1,769 cc). Basic weekly rate (6.66 Eur) minus discounts for safe driving, or plus penalizations for near misses.

Week	Distance driven (km)	Number of near-miss brakes (a)	Number of near-miss acceleration (b)	Minutes of smart phone use (c)	Cost of near-misses (Eur) (d)	Total weekly bill (Eur) (e)
1	30	0	0	0	-5.65	1.01
2	73	0	0	2	-5.29	1.37
3	104	2	2	2	0.93	7.59
4	260	6	2	1	9.00	15.66
5	705	19	4	21	54.94	61.60

Total bill for five weeks: 87.23 Eur

(e)=6.66+(d)

(d)=if ((a)>2, 1.5(a), -0.75(1-(a)),)+if ((b)>2, 4.71(b), -2.36(2-(b)))+if ((c)>2, 0.36(c), -0.18(1-(c)))

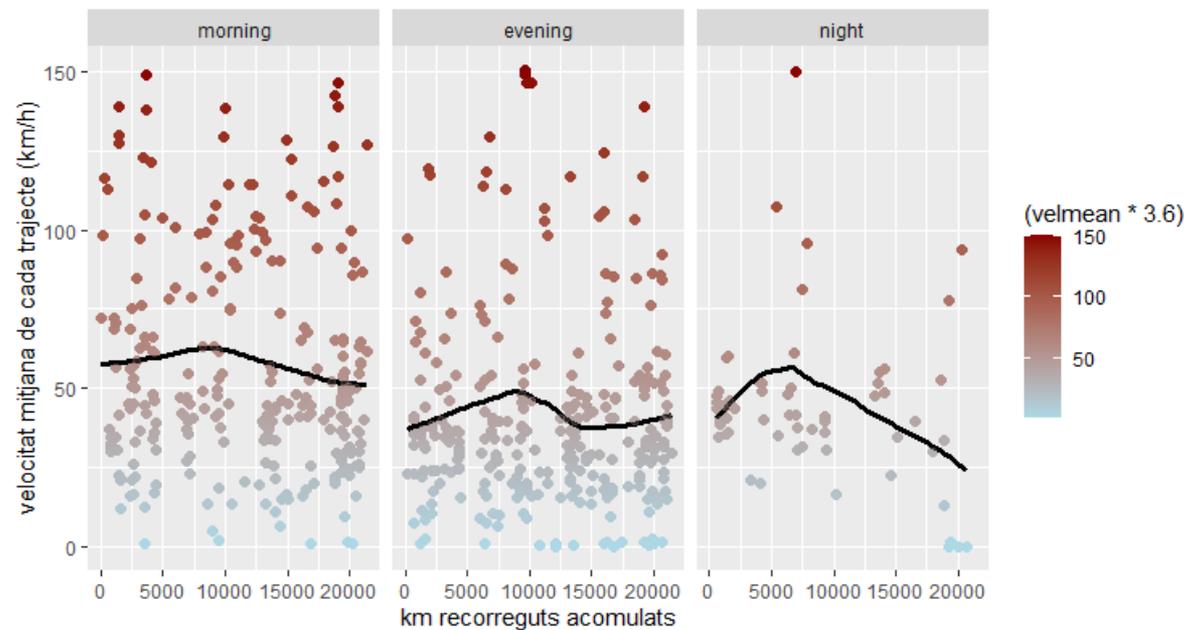
A red triangular warning sign with reflective borders is positioned on a paved road. In the background, a dark-colored car is parked, and a person wearing a bright yellow safety vest is crouching on the left side of the road. The scene is set outdoors with greenery and a clear sky.

HISTORICAL INFORMATION
ON CLAIMS

**Empirical exercise
with Spanish
insurer MAPFRE
Trip data 2018**

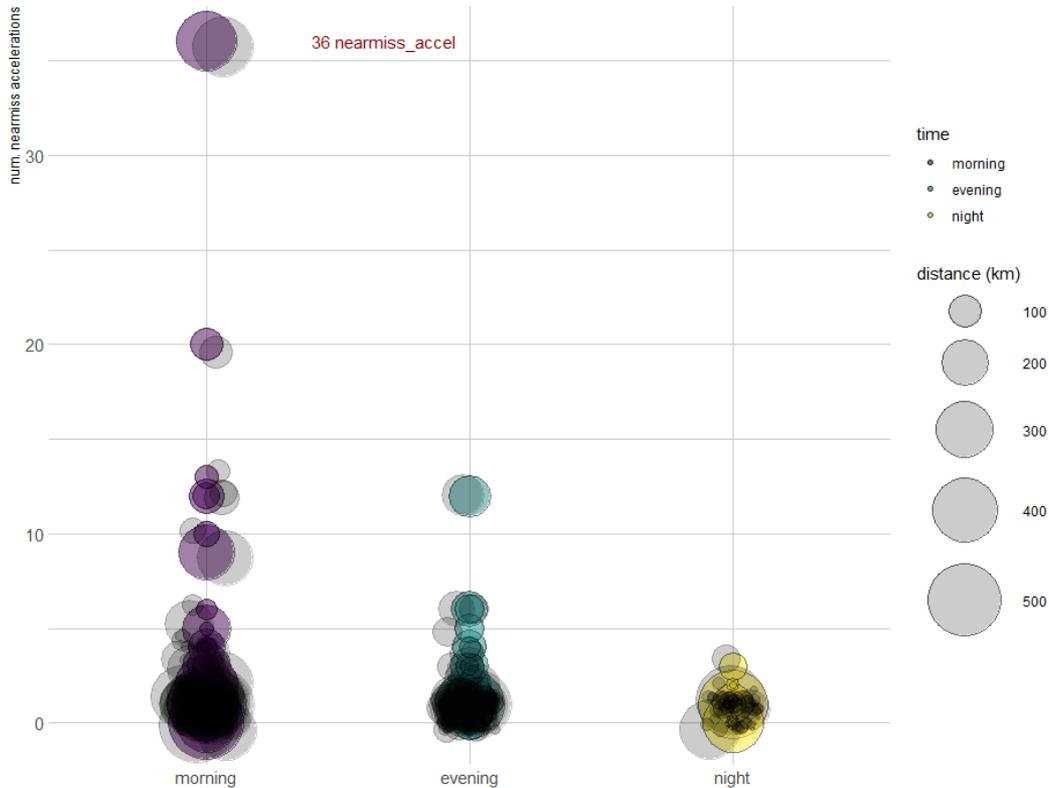
Telematics trip data: More information

Average speed



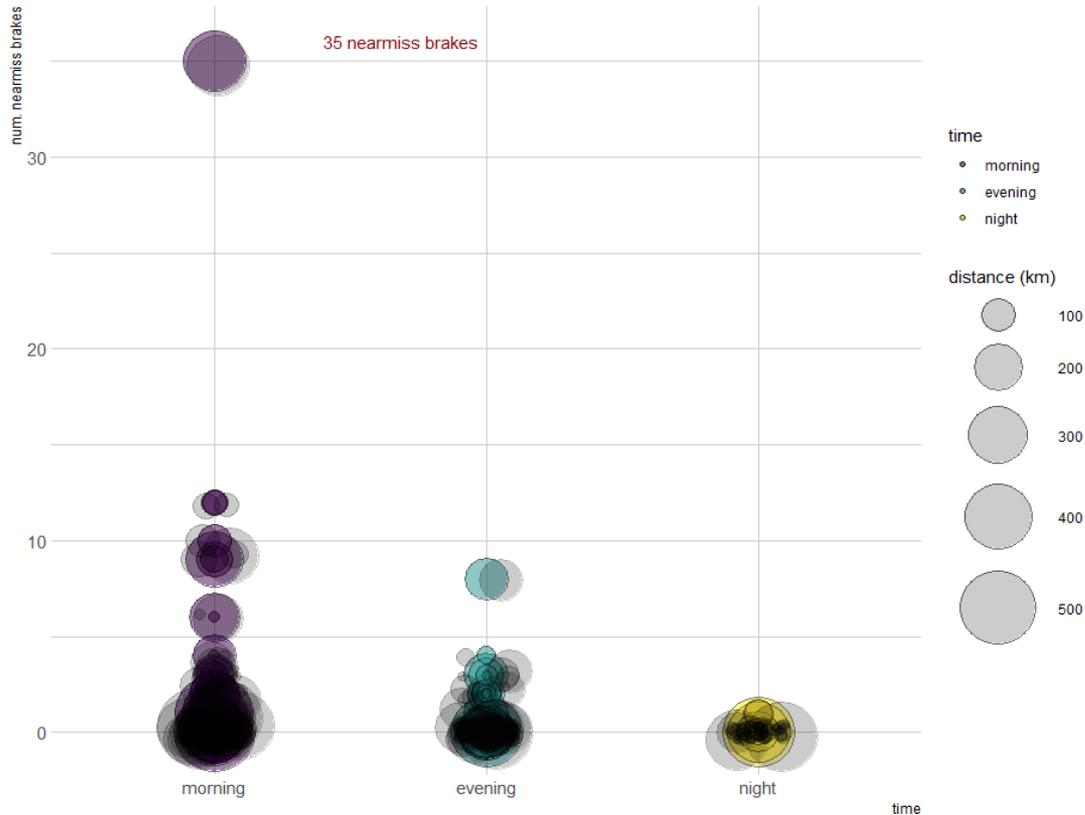
Highest average speeds in the morning (left) According to total km

Telematics data: Acceleration events



Acceleration events (vertical axis) in the morning (left) according to tripdistance (bubble).

Telematics data: Braking events



Braking events (vertical axis) in the morning (left) according to tripdistance (bubble).

4. CONCLUSIONS

Will **motor insurance** change?

- Consumers
 - Personalization
 - More interaction with insurers
- Manufacturers
 - Vehicles will be equipped with telematics and possibly vehicles provide a service (insurance included)
- Insurers
 - Products are more demanding 24/7.
 - Data analysts are needed. **Preprocessing is crucial.**
 - Communication to mass consumers of complex pricing
 - Prevention and service provision.

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