# Take home challenge

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# Objectives

Goal: higher driver engagement & reliability EU-wide

- 1. UX improvement Price Freeze
- 2. Supply health track and lift driver engagement
- 3. Campaign AB test

**Data snap:** 37 k drivers · 1.8 M rides · 10 k in test

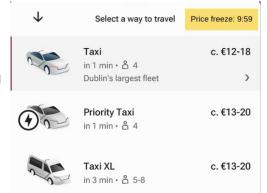
### 1. UX improvement - Price Freeze

#### What:

When a passenger taps "Book a ride" we lock the quoted fare (e.g. €12.40) for the next 10 minutes and show a small countdown badge beside the price.

#### Why:

Protect the passengers from surprising price changes while browsing and allowing the company to adjust the prices without negative impact on customer experience.



#### **Benefits**

- Eliminates the frustration of watching a quote jump while the passenger is still picking pickup / destination.
- Builds trust and speeds decisions ("I have ten minutes—no surprise, no hidden surge").
- Similar actions are visible in airline industry.
- Deliveroo price-lock A/B lifted conversion +3 pp (public case study)

#### How we know this matters?

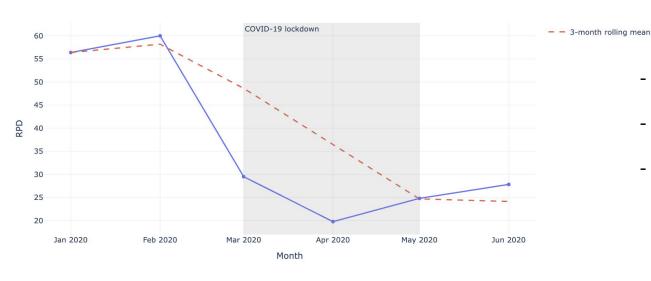
Signal	What we'll look at	Why it's convincing
App-store / Reddit keywords	"price changed", "surge", "misleading fare"	Consistent top-3 complaint in sentiment-scan
Funnel analytics	Ratio of sessions with a fare update → confirm	Fare jumps correlate with a >30 % higher drop-off
Competitive gap	Bolt gives 5-min fare hold in 8 markets	Passengers expect parity

### How we'll prove success (4-week A/B, 50/50 traffic)?

Metric (KPI)	Definition	Baseline
Search-to-Confirm Conversion	% rides that reach "Confirm" after quote shown	46 %
Session Drop-offs with Fare Change	Users who see a price change & abandon	8 %
Average GMV / Session	Total fare booked per app session	€4.60
Guard rails	Cancellation rate, driver acceptance, CSAT	

# 2. Supply health KPI - Rides-per-active-driver

Rides per Active Driver-Day — Last 24 Months

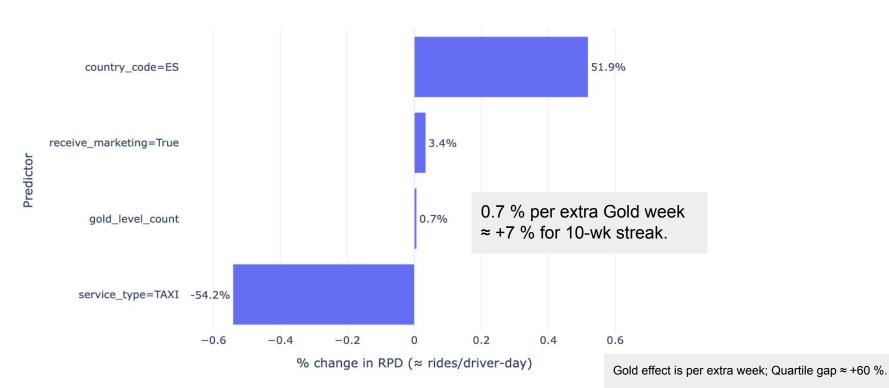


- Normalises fleet size, links to revenue
- 3-mo MA smoothes seasonality
- COVID dip −50 % (Mar '20)

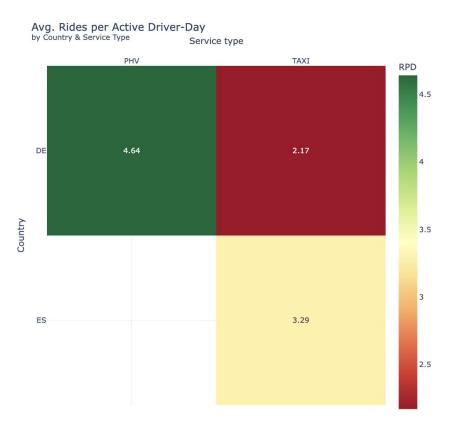
$$RPD = \frac{rides}{active driver days}$$

# 2. What drives higher rides-per-active-driver?

Largest model effects on RPD (percent change)



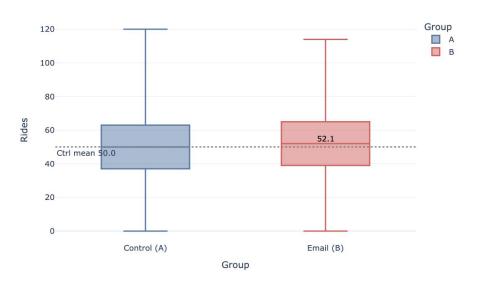
# 2. Segment heat-map



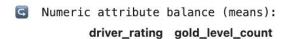
- Germany PHV top performers (RPD ≈ 4.6)
- Germany Taxi lags (RPD ≈ 2.2)
- Spain Taxi only with moderate results (RPD ≈ 3.3)

# 3. Email campaign lift

#### Rides in the week after email



- Group B got weekly surge email → **+2.1 rides** / driver
- **+4 %** lift (95 % CI +3–5 %)
- Welch ttest: p < 0.001, n = 5 000 each



#### test\_group

Α	4.89	20.17
В	4.90	17.36

Service-type counts:

test\_group A B

#### service\_type

PHV 4 6

### 4. Final notes

### Suggestions - not additive

- **Gold-streak gamification** (+0.3-0.6 rides)
- Marketing re-opt-in toggle (+0.6 rides)
- **Push hot-spot alerts** (+2 rides)
  - as in app or push notifications

### **Risks**

- Retention < 12w</li>
- Taxi backlash → monitor earnings spread
- Weekly RPD dashboard monitoring

