## The business case for autonomous deliveries: does it exist?

An economic analysis of the use of autonomous vehicle technology for last mile deliveries

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# Urban Population as a Percentage of Total Population 

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Urban Population (\% of total population)


## The rise of e-commerce

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Delivery Amount (Unit : Million Box)

Unit Price
(Unit : \$/Box)

## Renault, LCVs and AVs

■ Groupe Renault manufacturers Light Commercial Vehicles. These

■ are used extensively for Last Mile Deliveries (LMDs).

- contribute significantly to the revenue of the firm.

■ A new technology arises; Autonomous Vehicles (AVs).

- These vehicles may be used for LMDs.
- This may impact LCV sales.
- Do these vehicles have a business case for LMDs?
- Removal of driver creates value.
- Private perspective.


## Autonomous Single Delivery Vehicle

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ASDVs make a single delivery at a time. Here is Amazon Scout. It goes slow, can use sidewalks.


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AMDVs make multiple deliveries in 1 round. Here is Nuro; it makes 4. It goes fast, uses roads.


## Methodology

- Prospective analysis
- A cost structure of last mile deliveries is modelled, and then extended.
- Data was gathered from three sources

■ Academic Literature and Professional Reports

- Interviews
- Field Visits

■ Operational context/constraints identified, understood and modelled.

## Modelling: Overview

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Smaller warehouse at $\mathbf{1 k m}$ from Customers

## Modelling: Summary

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| Distance from Warehouse | Size of Warehouse | Average distance b/w customers | Vehicle Used ${ }^{1}$ | Market Segment | Output |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1 km 10 km 30 km | $\begin{aligned} & 7500 \mathrm{~m} 2 \\ & 15000 \mathrm{~m} 2 \\ & 25000 \mathrm{~m} 2 \\ & 40000 \mathrm{~m} 2 \end{aligned}$ | 0.5 km 0.6 km 0.7 km <br> 5 km | Diesel Van <br> Electric Van <br> Cargo Bike <br> ASDV ${ }^{2}$ <br> AMDV ${ }^{3}$ | Parcels <br> Groceries <br> B2B ${ }^{4}$ | Cost/Delivery |

${ }^{1}$ Assumption: Fleet is unimodal
${ }^{2}$ Autonomous Single Delivery Vehicle
${ }^{3}$ Autonomous Multiple Delivery Vehicle
${ }^{4}$ Business to Business

## Literature Overview

- Previous Literature
- Total Cost of Ownership (TCO) approach for vehicle choice
- (Lebeau et al.,2019), (Figenbaum,2018), (Camilleri,2017)
- Warehouse Location - (Combes, 2019)
- Autonomous Vehicles - (Figliozzi, 2019, 2020)
- Contribution
- Creation of integrated model (warehouse location AND vehicle choice)
- Include real world operational constraints

■ Account for heterogeneity of LMDs - different Market Segments

- Level of service penalty

■ Driver/Deliverer experience

## Level of Service Penalty

Table: Level of Service Penalty

|  | Task | Handled by in <br> Conventional <br> Delivery | Handled by in <br> Autonomous De- <br> livery |
| :--- | :--- | :--- | :--- |
| 1. | Navigating | Driver | Vehicle |
| 2. | Calling and notifying cus- <br> tomer of arrival | Driver | Vehicle |
| 3. | Locating merchandise in <br> storage | Driver | Customer |
| 4. | Unloading merchandise | Driver | Customer |
| 5. | Delivering merchandise <br> to end customer | Driver | Customer (collects <br> it himself) |
| 6. | Getting proof of success- <br> ful delivery from cus- <br> tomer | Driver | Vehicle (registers <br> opening/closing of <br> door) |

## Market Segments

Table: Market Segments considered for Analysis.

| Variables | Unit | Parcels | Groceries | B2B $^{5}$ |
| :---: | :---: | :---: | :---: | :---: |
| Deliveries per Round | - | $100+$ | 20 | 20 |
| Avg. Weight/Delivery | Kg | 0.3 | 25 | 50 |
| Time per Delivery | Minutes | 3 | 12 | 12 |
| Vehicle Refrigerated | - | No | Yes | No |
| Level of Service Penalty $^{6}$ | $€ /$ Delivery | 1.5 | 3 | 5 |

## Values above for Diesel Van.

[^0]
## Modelling: Math

Deliveries $/$ Week $=f($ Vehicle, MarketSegment, CustomerDensity, SpeedBetweenDeliveries, LocationofWarehouse, ApproachSpeed)

## Results: Scenarios

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| Scenarios | Purchase Costs(Euros) |  | Losses <br> Linked to Driver Experience | Costs <br> Linked to <br> Remote <br> Opera- <br> tors | Level of Service Penalty | Driver Wage per Hour |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Scenario 1 | SDAV $=250,000$ | MDAV $=300,000$ | None | None | None | 11.17 € |
| Scenario 2 | SDAV $=250,000$ | MDAV $=300,000$ | 20\% loss of deliveries per day | $\begin{aligned} & 1 \text { remote } \\ & \text { operator } \\ & \text { for } 20 \\ & \text { vehicles } \end{aligned}$ | 1.5,3 and 5 Euros acc. to Market Segment | $11.17 €$ |
| Scenario 3 | SDAV = Price of Cargo Bike (7,890) | MDAV $=$ Price of Electric Van $(64,643)$ | 20\% loss of deliveries per day | $\begin{aligned} & 1 \text { remote } \\ & \text { operator } \\ & \text { for } 20 \\ & \text { vehicles } \end{aligned}$ | 1.5,3 and 5 Euros acc. to Market Segment | $11.17 €$ |
| Scenario 4 | SDAV = Price of Cargo Bike (7,890) | MDAV $=$ Price of Electric Van $(64,643)$ | 20\% loss of deliveries per day | $\begin{aligned} & 1 \text { remote } \\ & \text { operator } \\ & \text { for } 20 \\ & \text { vehicles } \end{aligned}$ | None | $11.17 €$ |
| Scenario 5 | SDAV = Price of Cargo Bike (7,890) | MDAV $=$ Price of Electric Van $(64,643)$ | 20\% loss of deliveries per day | $\begin{aligned} & \hline 1 \text { remote } \\ & \text { operator } \\ & \text { for } 20 \\ & \text { vehicles } \end{aligned}$ | None | 80 cents |
| Scenario 6 (Green Vehicles Only) | SDAV = Price of Cargo Bike (7,890) | MDAV $=$ Price of Electric Van $(64,643)$ | 20\% loss of deliveries per day | $\begin{aligned} & 1 \text { remote } \\ & \text { operator } \\ & \text { for } 20 \\ & \text { vehicles } \end{aligned}$ | None | 80 cents |

Table: Scenarios

## Cost per delivery: S1, Parcels

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Cost Per Delivery as a function of Distance between Deliveries (Market Segment = Parcel). Scenario 1.



## Cost per delivery: S1, Groceries

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Cost Per Delivery as a function of Distance between Defiveries (Market Segment = Groceries). Scenario 1.
Vehicles compored = Diesel Van, Electric Van, Cargo Bike, Autanomous Single Delivery Vehicle, Autonomous Mutiple Deliwery Vehicle Warehouse located at 1,10 or 30 kn away from customers.



## Cost per delivery: S1, B2B

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Cost Per Delivery as a function of Distance between Deliveries (Market Segment = B2B). Scenario 1.
Vehicles compered = Diesel Van, Electric Van, Electric Carros Bke, Mutiple Dehivery Autonomous Vehicle. Warehouse located at 1,10 or 30 km away from customers


## Optimal Warehouse Location/Vehicle Combination

- Warehouse location and vehicle choice are not independent:

■ Some vehicles - ex. Cargo Bikes - are severely limited by their autonomy

- Comparing TCO over different vehicles is not appropriate if some vehicles have different operational constraints
■ A logistics firm will choose a combination of warehouse location/vehicle type that offers the least cost per delivery.


## Optimal Warehouse Location/Vehicle type: S1

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Most competive vehicle-warehouse location combination per market segment. Scenario 1 Three wa ehouse loc:aions ase compared; $1 \mathrm{~km}, 10 \mathrm{~km}$, and 30 km from the frist deflery, Market Segment $=$ Parcels


## Market Segment $=$ Groceries

Distaxce Eetween Detweries (km)



Market Sogment $=\mathrm{B} 2 \mathrm{~B}$


## Optimal Warehouse Location/Vehicle type: S4

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Most competitive vehicle-warehouse location combination per market segment. Scenario 4: Customers indifferent to level of sarvice. Thre warehouse loc:aions ase compared, $1 \mathrm{~km}, 10 \mathrm{~km}$, and 30 km from the frost defleery.

## Market Segment $=$ Parcels

$$
\begin{aligned}
& +20 \times 2
\end{aligned}
$$



Distance fetween Defiveries (kmi)
Market Segment $=$ B2B


## Optimal Warehouse Location／Vehicle type：Low Wage regions

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Most competitive vehicle－warehouse location combination per market segment．Scenario：Low Wage Geographical Regions． Three warehouse locations are compared； $1 \mathrm{~km}, 10 \mathrm{~km}$, and 30 km from the frsct defleery．
Market Segment＝Parcels

$$
\begin{aligned}
& \text { Vehicle } \\
& \text { onan }
\end{aligned}
$$

Ditance Between Delireries $(\mathrm{km})$
Market Segment = Groceries


> Market Segment = B2B

Distance Between Delireries（kn）


$$
\begin{aligned}
& \text { 高落 }
\end{aligned}
$$

## Conclusion

■ Model which compares cost per delivery over different vehicles, warehouse locations, size, customer densities and market segments is developed.

- Competitiveness domains for different vehicles are determined.
- Conventional vehicles extremely efficient under current operating scenario.
- If AV s become cheaper, and service penalty is borne by customers, AV s are competitive for certain market segments, under certain conditions.
- Future work: Model to be extended to mixed fleets, question of lead time to be addressed.


## Contact

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Thanks for your time and insights!
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## Model: Overview

■ In this theses, a microeconomic model of cost per delivery is developed. Its components are

■ Warehouse

- Vehicles

■ Elements specific to the use of AECS vehicles

## Appendix 1: Warehouse Costs

A warehouse consists of

- A physical structure at a location

■ Employees
■ Equipment, Electricity, Maintenance etc.
warehouse $_{\text {costs }}=$ warehouse $_{\text {rent }}+$

$$
\text { warehouse }_{\text {employeecost }}+\text { warehouse }_{\text {othercosts }}
$$

The following assumptions are made;

- A logistic firm requires a warehouse for its cross-docking operations.
- This warehouse is rented.
- Firm operates only 1 warehouse with unimodal fleet.
- Equipment, Electricity, Maintenance and other costs are assumed to be dependant on Market segment to which warehouse caters.


## Warehouse Rent: Size, Location

The rent of a warehouse depends on two criteria;

- its size and,
- its distance from the city center.

Based on data from property rental sites, this can be expressed as;

$$
\begin{equation*}
\text { warehouse }_{\text {rent }}=\left(0.0003 *\left(w h_{d}^{4}\right)-0.04 *\left(w h_{d}^{3}\right)+2.15 *\left(w h_{d}^{2}\right)-43.1 *\left(w h_{d}\right)+392.3\right. \tag{6}
\end{equation*}
$$

Where warehouse $_{\text {rent }}$ is the rent per square meter and $w h_{d}$ is the distance of the warehouse from city center.

## Warehouse Rent: Size, Location (contd.)

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Warehouse Rent per Square Meter as a Function of Distance from City Center for lle de France region. City center is taken as Place du Chatelet.


## Warehouse Size and Number of Employees

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Based on available (limited) data, the number of employees as a function of the size of the warehouse follows a linear relation.

$$
\begin{equation*}
\text { warehouse }_{\text {employeecost }}=\text { warehouse }_{\text {size }} * \text { employees }_{m 2} * \text { cost }_{\text {employee }} \tag{7}
\end{equation*}
$$

Number of Employees as a function of Warehouse Size


## Warehouse: other costs

Apart from rental costs, and wage costs to employees, there are other costs in running a warehouse. These include, but are not limited to;

■ Equipment costs
■ Electricity costs
■ Maintenance costs
Equipment costs include the costs of sorting machines, or Automated Storage and Retrieval Machine, which are extremely expensive.

## Warehouse: other costs (contd.)

All other costs, not indicated by rent or employment costs are represented by the greek letter $\delta$.

$$
\begin{equation*}
\text { warehouse }_{\text {othercosts }}=\delta \tag{8}
\end{equation*}
$$

In this model it is assumed that for a given market segment, for a given warehouse size, these other costs are the same. These costs are not known. This creates a problem that can be tackled in at least two ways;

1 calibrate model such that cost/delivery reflect real world values. ${ }^{7}$
2 subtract the min. cost/delivery from all other costs/delivery for each configuration such that $\delta$ is removed.

[^1]
## Warehouse: other costs (contd.)

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Method 2 can be represented as:

$$
\text { warehouse }_{\text {costs }}^{\text {configuration } 1}=\text { warehouse }_{\text {rent }}^{\text {configuration } 1}+
$$

warehouse employeecost

$$
\text { warehouse }_{\text {costs }}^{\text {configuration2 }}=\text { warehouse }_{\text {rent }}^{\text {configuration2 }}+
$$

$$
\text { warehouse }_{\text {employeecost }}^{\text {configuration } 2}+\text { warehouse othercosts }
$$

## Warehouse: other costs (contd.)

As for a given configuration (warehouse location, warehouse size, customer density, vehicle, market segment),

$$
\begin{equation*}
\text { warehouse }_{\text {othercosts }}^{\text {configuration } 1}=\text { warehouse }_{\text {othercosts }}^{\text {configuration } 2}=\delta \tag{11}
\end{equation*}
$$

From 9, 10 and 11 we have;
warehouse $_{\text {costs }}^{\text {configuration } 1}-$ warehouse $_{\text {costs }}^{\text {configuration2 }}=$ warehouse $_{\text {rent }}^{\text {configuration }}$ - warehouse rent $_{\text {configuration }}+$ warehouse employeecost - warehouse employeecost

## Warehouse Size and Throughput

Throughput of a warehouse is assumed proportional to its size.

$$
\begin{array}{r}
\text { warehouse }_{\text {throughput } \propto \text { warehouse }_{\text {size }}}^{\text {warehouse }_{\text {throughput }} \propto \text { fleet }_{\text {size }}} \begin{array}{r}
\text { warehouse } \\
\text { size }
\end{array} \propto \text { fleet }_{\text {size }} \\
\text { warehous }_{\text {size }}=k * \text { fleet }_{\text {size }}
\end{array}
$$

■ Width and number of bays are determined using Google Maps.
■ Area of warehouse is determined using Google Maps.

■ Number of vehicles per square meter is thus calculated.

Warehouse Size and Throughput (contd.)

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From above images;
■ Area of building $=48 \mathrm{~m} * 248 \mathrm{~m}=11904 \mathrm{~m} 2$

- Length required by 1 vehicle $=3 \mathrm{~m}$

■ Length of warehouse $=248^{*} 2=496 \mathrm{~m}$
■ No. of vehicles $=496 / 3=165.33$
■ No. of vehicles $/ \mathrm{m} 2=165.33 / 11904=0.014$ vans $/ \mathrm{m} 2$
■ No. of vehicles $/ \mathrm{m} 2($ only 1 side $)=0.007$

## Warehouse Size and Throughput (contd.)

■ No.vehicles/m2 depends on the type of vehicle.

- an ASDV is not as wide as a Cargo Bike is not as wide as an AMDV is not as wide as a Van.
- Area reqd(ASDV) $=0.5^{*}$ Area reqd(Van)
- Area reqd(Cargo Bike) $=0.7^{*}$ Area reqd(Van)
- Area reqd (AMDV) $=0.8^{*}$ Area reqd(Van)


## Total Cost of Ownership (TCO) of a Vehicle

- The TCO is a widely used approach to compare different cost structures over differing vehicle technologies.
■ It involves comparing actualised costs for each period over the life of a vehicle. These costs include;
- Costs independent of distance travelled

■ Purchase, resale, insurance, subsidies
■ Costs dependent on distance travelled
■ Fuel, maintenance

## Vehicles Compared for TCO Analysis

■ Maximum number of deliveries a vehicle can accomplish in a day is determined.

- This is used to determine - using TCO, Warehouse Costs and Driver Wages - the Cost per Delivery
- The following vehicles are compared;
- Diesel Vans (often the base case)
- Electric Vans
- Electric Cargo Bikes
- Single Delivery Autonomous Vehicles (SDAV)
- Multiple Delivery Autonomous Vehicles (MDAV)


## Vehicles Compared: TCO Data I

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TCO per day as a function of Capacity Volume


Energy source of Van
$\boxminus$ Diesel
追 Electric
$\rightleftarrows$ Petrol

## Vehicles Compared: TCO Data II

■ Analysis based on data of 50 Vans of differing energy source, capacity, manufacturer, etc.
■ Vans of different volume capacities used in different market segments. Parcels use van with capacity of 3.5 m 3 , Groceries and B2B of 8m3.
■ Purchase and Energy costs for Grocery Market Segments are 1.2 times that of B 2 B segment due to refrigeration.

- ASDVs cost $2 x$, and $A M D V$ s cost $3 x$ diesel vans. ${ }^{8}$

[^2]
## Methodology: determining Number of Deliveries per day

Table: Variables used

| Variable | Description | Unit |
| :---: | :---: | :---: |
| $n$ | number of deliveries in a day | - |
| $t_{\text {lod }}$ | length of day | hours |
| $t_{\text {loading }}$ | time to load vehicle | hours |
| $t_{\text {delivery }}$ | time per delivery | hours |
| $d_{\text {wfd }}$ | distance of warehouse to first delivery | kilometers |
| $d_{b d}$ | average distance between deliveries | kilometers |
| $s_{w f d}$ | average speed between warehouse and delivery area | kilometers per hour |
| $s_{b d}$ | average speed between deliveries | kilometers per hour |
| $d p s$ | deliveries per stop | - |
| $d p s_{\text {coeff }}$ | time coefficient if more than 1 deliveries per stop | - |
| $v_{\text {autonomy }}$ | vehicle autonomy | kilometers |
| $v_{\text {volumecapacity }}$ | vehicle volume capacity | cubic meters |
| $v_{\text {weightcapacity }}$ | vehicle weight capacity | kilograms |
| $p_{\text {vol }}$ | average parcel volume | cubic meters |
| $p_{w t}$ | average parcel weight | kilograms |

Methodology: Equations - Number of Deliveries per day I

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To find number of deliveries per day, the following steps are implemented;

1 Solve for n subject to
1 Time constraint

$$
\begin{equation*}
\mathrm{n}_{1}=\left(\frac{\left(t_{\text {lod }}-\frac{2 * d_{\text {wfd }}}{s_{\text {wfd }}}+\frac{d_{b d}}{s_{b d}}-t_{\text {loading }}\right) * d p s}{\frac{d_{b d}}{s_{b d}}+\left(\left(1+(d p s-1) * d p s_{\text {coeff }}\right) * t p d\right)}\right) \tag{17}
\end{equation*}
$$

2 Autonomy Constraint

$$
\begin{equation*}
n_{2}=\left(v_{\text {autonomy }}-\frac{2 * d_{w f d}}{s_{w f d}}\right) * \frac{d_{b d}}{s_{b d}}+1 \tag{18}
\end{equation*}
$$

3 Volume Constraint

$$
\begin{equation*}
n_{3}=\frac{v_{\text {volumecapacity }}}{p_{\text {vol }}} \tag{19}
\end{equation*}
$$

## Methodology: Equations - Number of Deliveries per day II

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2 Determine time of round with $n_{\text {round }}$

$$
\begin{equation*}
\mathrm{t}_{\text {round } 1}=2 * \frac{d_{w f d}}{s_{w f d}}+\left(\frac{n_{\text {round } 1}}{d p s}-1\right) * \frac{d_{b d}}{s_{b d}}+\left(\frac{n_{\text {round } 1}}{d p s}\right) *\left(1+(d p s-1) * d p s_{\text {coeff }}\right) * t_{\text {delivery }}+t_{\text {loading }} \tag{22}
\end{equation*}
$$

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4 Weight Constraint

$$
\begin{equation*}
n_{4}=\frac{v_{\text {weightcapacity }}}{p_{w t}} \tag{20}
\end{equation*}
$$

5 Choose min from 17,18, 19, 20

$$
\begin{equation*}
n_{\text {round } 1}=\min \left(n_{1}, n_{2}, n_{3}, n_{4}\right) \tag{21}
\end{equation*}
$$

3 Determine no. of 'complete' rounds per day

$$
\begin{equation*}
r_{n}=\left\lfloor\frac{t_{\text {lod }}}{t_{\text {round } 1}}\right\rfloor \tag{23}
\end{equation*}
$$

## Methodology: Equations - Number of Deliveries per day III

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4 Determine time left

$$
\begin{equation*}
t_{\text {left }}=t_{\text {lod }}-r_{n} * t_{\text {round } 1} \tag{24}
\end{equation*}
$$

5 Determine number of other deliveries
1 Time constraint

$$
\begin{equation*}
\mathrm{n}_{5}=\left(\frac{\left(t_{\text {left }}-\frac{2 * d_{\text {wfd }}}{s_{\text {wfd }}}+\frac{d_{b d}}{s_{b d}}-t_{\text {loading }}\right) * d p s}{\frac{d_{b d}}{s_{b d}}+\left(\left(1+(d p s-1) * d p s_{\text {coeff }}\right) * t p d\right)}\right) \tag{25}
\end{equation*}
$$

2 Autonomy Constraint will be same as 18
3 Volume Constraint will be same as 19
4 Weight Constraint will be same as 20
5 Choose min from 25,18,19,20

$$
\begin{equation*}
n_{\text {other }}=\min \left(n_{5}, n_{2}, n_{3}, n_{4}\right) \tag{26}
\end{equation*}
$$

## Methodology: Equations - Number of Deliveries per day IV

6 Thus, from 23, 21,26, total deliveries in a day are given as;

$$
\begin{equation*}
d_{\text {total }}=r_{n} * n_{\text {round } 1}+n_{\text {other }} \tag{27}
\end{equation*}
$$

- From $d_{\text {total }}, d_{\text {total }}^{\text {week }}$ is determined.
- This is applied across whole fleet to determine dfleet total week.
- This is used to determine cost/delivery.


## The Driver/Deliverer

■ In conventional deliveries, the driver/deliverer accomplishes the last meters of the last mile and obtains a confirmation of delivery from the end customer.
■ He is paid an hourly wage of 11 euros/hour. He works 40 hours/week.
■ In deliveries using autonomous vehicles, the driver wage is zero.

## Latent Knowledge I

Table: Latent Knowledge in Drivers

| S.No | Skill/Knowledge | Application |
| :--- | :--- | :--- |
| 1. | Personal <br> relationships with <br> various <br> inhabitants | (ex. If customer is not avail- <br> able, driver delivers to neigh- <br> bour, based on previous agree- <br> ments) |
|  |  | -Accommodating non prepared <br> return on other rounds (Espe- <br> cially true for business deliv- |
|  |  | eries, If a return form a cus- <br> tomer is not prepared, s/he can |
|  |  |  |
| making the round more effi- |  |  |
| cient) |  |  |

## Latent Knowledge II

|  |  | -Concierge |
| :---: | :---: | :---: |
| 2. | Knowledge of Parking spaces | - Reduction of time spent to look for a parking space |
|  |  | - Knowledge of parking time restrictions |
|  |  | Alternate parking spots |
| 3. | Knowledge of traffic conditions and trends | Ex. Higher traffic in a specific repeated delivery address at a particular time (commercial center) |
|  |  | -Reordering delivery order to achieve faster overall delivery |

## Latent Knowledge III

Introduction
Urban Population E-Commerce Context

Methodology
Modelling
Overview
Summary
Literature
Service Penalty
Market Segments
Math
Results
Cost per Delivery
Winning
Combination

| 4. | Knowledge of <br> geographical <br> quirks | Ex. Access codes, GPS Map <br> Failures: Dead Ends |
| :--- | :--- | :--- |
|  | - Driver saves a lot of time <br> by already knowing the access <br> code, or the requirement of it. |  |
|  | -Driver aware of GPS failures <br> in certain specific scenarios, ex. <br> Dead ends, and avoids them. |  |

## Latent Knowledge IV

An experienced driver can be $40 \%$ more effective during his/her rounds. "the difference between two of our drivers (D22 and D24) with similar round sizes and parcel volumes shows a considerable variation in effective- ness, with D22 driving $44 \%$ less distance, spending $35 \%$ less time per parcel, $29 \%$ less driving time per parcel, and $39 \%$ less parking time per parcel. The variation in effectiveness of our drivers relates to better route planning, exploitation of accumulated knowledge of the round, personal relation- ships with other stakeholders, the amount of time spent at the curbside and the influence of walking. These statistics show that more effective drivers achieve higher rate of delivery of parcels per minute while spending less time driving and parking in the van".(Bates et al., 2018)


[^0]:    ${ }^{5}$ Business to Business
    ${ }^{6}$ iff $A V$ s used

[^1]:    ${ }^{7}$ These values are known through field visits.

[^2]:    ${ }^{8}$ arbitrary choice, will soon change it to $2 x, 3 x$ of Electric Vans.

