eDream Odigeo Baggage Likelihood Model

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"The simulacrum is never that which conceals the truth — it is the truth which conceals that there is none. The simulacrum is true."

Jean Baudrillard, Simulacra and Simulation, 1988

Exploratory Data Analysis

Plotting and visualising the distributions of different variables

Overall proportion of people having booked extra baggage:

```
## ## No Extra Baggage Extra Baggage ## 80.4 19.6
```

How to make the model as parsimonious as possible? Let's see which values do not have such relevance

I think the best would be to try a good old mulitple logisitic regression model with several dummy variables, but omit data which at first you find useless.

Data which seems irrelevant at first sight:

- TIMESTAMP
- DEPARTURE
- ARRIVAL

As there is very strong class imbalance within the TRAIN booking binary variable (99.5% in the training set did not book a train).

```
## ## False True
## 99.5 0.5
```

Similarly within the PRODUCT variable (98.1% booked a Trip compared to a Dynpack) - both these variables were not considered.

```
## ## DYNPACK TRIP ## 1.9 98.1
```

Investigating further, comparing baggage selection rates among different variables

Feature engineering

Mapping of the booking company and encoding

It would be interesting to see if there are significant variations in baggage booking between eDreams (ED), Opodo (OP) or Go Voyage (GO) - should use a string operation on this

To simplify I assume that there is no local variability between bookings in UK, Italy, Spain, France etc.. Also, extracting different countries would just lead to a categorical factor variable with potentially many levels - which is not so good for a machine learning algorithm.

It seems that however there is nothing interesting there - the proportions are virtually exactly the same everywhere

##						
##		No	Extra	Baggage	${\tt Extra}$	Baggage
##	EDREAMS			80.6		19.4
##	GO VOYAGE			81.2		18.8
##	OPODO			79.9		20.1
##	OTHER			75.7		24.3

The website variable can therefore be ommitted

Don't know what to do with GDS variables, I remove them for now and come back later

Synthetic variable family size

After creating a synthetic variable combining ADULTS + CHILDREN + INFANTS called FAMILY_SIZE Maybe those who pick SMS as an extra are more likely to pick other extras? To investigate

Not much going on there actually

Could it be that with certain devices more customers book devices?

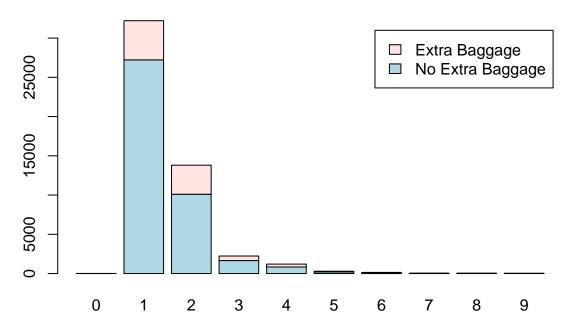
##						
##		No	Extra	Baggage	${\tt Extra}$	Baggage
##	COMPUTER			79.5		20.5
##	OTHER			76.6		23.4
##	SMARTPHONE			83.2		16.8
##	TABLET			79.6		20.4

Not so much actually really, a bit of a face value judgement but let's omit the DEVICE variable for now and investigate later if we have time

Adults travelling alone I would assume would be less likely to book luggage, but with one or more children much more likely to get luggage, especially with infants

Indeed from a small table you can see that:

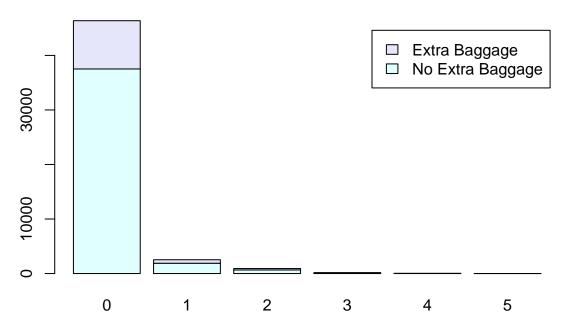
Adult Booking Distribution



Number of Adults in Booking

##						
##		No	${\tt Extra}$	Baggage	Extra	Baggage
##	0			100.0		0.0
##	1			84.5		15.5
##	2			73.2		26.8
##	3			74.1		25.9
##	4			69.5		30.5
##	5			71.7		28.3
##	6			63.3		36.7
##	7			69.4		30.6
##	8			80.0		20.0
##	9			73.9		26.1

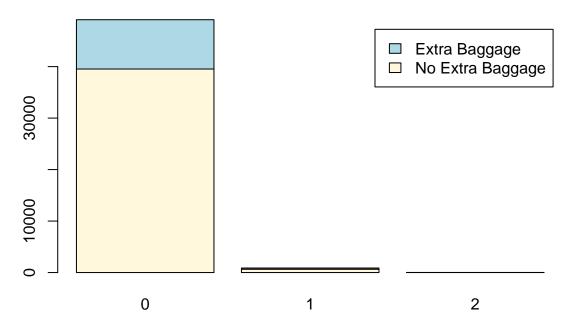
Children Booking Distribution



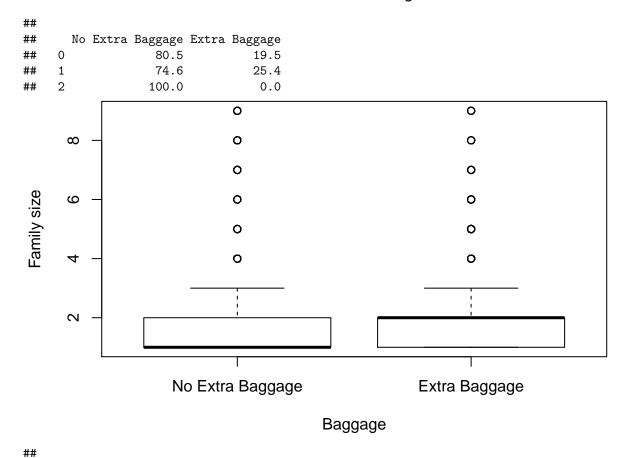
Number of Children in Booking

##						
##		No	${\tt Extra}$	Baggage	${\tt Extra}$	Baggage
##	0			80.9		19.1
##	1			74.6		25.4
##	2			71.8		28.2
##	3			72.6		27.4
##	4			88.5		11.5
##	5			0.0		100.0

Infants Booking Distribution



Number of Infants in Booking

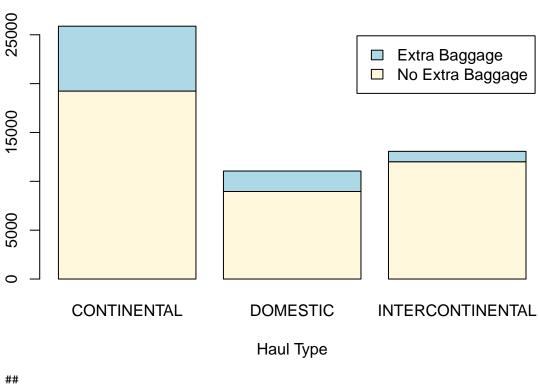


No Extra Baggage Extra Baggage

##	1	84.7	15.3
##	2	74.3	25.7
##	3	74.8	25.2
##	4	69.1	30.9
##	5	71.2	28.8
##	6	69.9	30.1
##	7	63.5	36.5
##	8	77.8	22.2
##	9	62.9	37.1

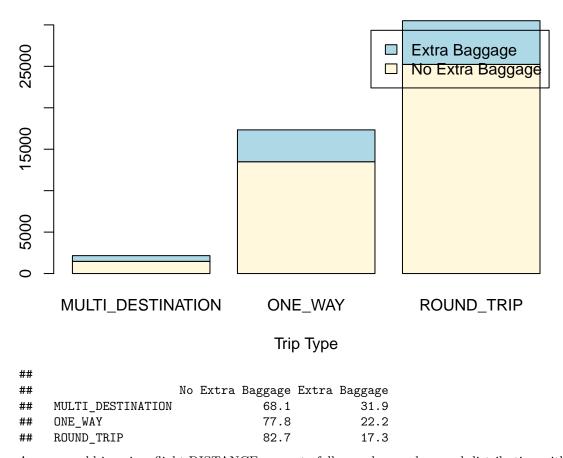
I would imagine that flight distance would account for a lot of the probability of luggage selection (high R2), as people who travel further I would assume need to carry more than if they are doing a short weekend trip within Europe

Haul Booking Distribution



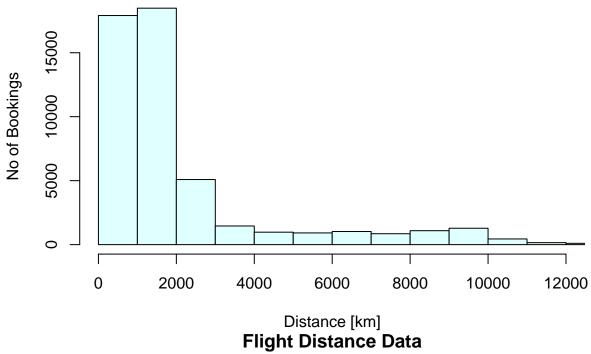
#	#						
#	#		No	Extra	Baggage	${\tt Extra}$	Baggage
#	#	CONTINENTAL			74.3		25.7
#	#	DOMESTIC			81.1		18.9
#	#	INTERCONTINENTAL			91.9		8.1

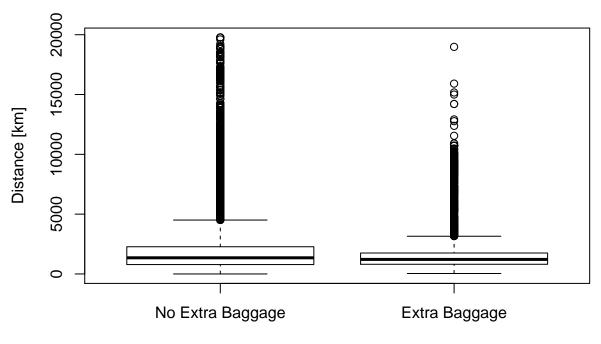
Trip Booking Distribution



As one would imagine, flight DISTANCE seems to follow a skew eved normal distribution with alot of short flights between 0-3000 km and then drastic reductions from then onwards.

Air Travel Distance Distribution





Building the model

Logisitic Regression

Although our first assumption that number of adults was a predictor of baggage selection - indeed fitting it to our general linear model it would seem so as it's hightly significant in terms of p-value:

```
##
## Call:
  glm(formula = EXTRA_BAGGAGE ~ DISTANCE + factor(HAUL_TYPE) +
       factor(TRIP_TYPE) + ADULTS + CHILDREN + INFANTS, family = binomial(link = "logit"),
##
       data = train)
##
## Deviance Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
## -1.6971 -0.7341 -0.6139 -0.3449
                                        2.6021
##
## Coefficients:
##
                                       Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                                     -8.539e-01
                                                 5.373e-02 -15.892
## DISTANCE
                                     -5.518e-05
                                                 8.476e-06 -6.510 7.52e-11
## factor(HAUL_TYPE)DOMESTIC
                                     -4.215e-01
                                                 2.887e-02 -14.602
                                                                    < 2e-16
                                                 4.366e-02 -26.092
## factor(HAUL_TYPE)INTERCONTINENTAL -1.139e+00
                                                                     < 2e-16
## factor(TRIP_TYPE)ONE_WAY
                                     -4.411e-01
                                                 5.141e-02 -8.580
                                                                     < 2e-16
## factor(TRIP_TYPE)ROUND_TRIP
                                     -7.569e-01
                                                 5.022e-02 -15.073
## ADULTS
                                      2.770e-01
                                                 1.266e-02
                                                             21.873
                                                                    < 2e-16
## CHILDREN
                                      2.368e-01
                                                 2.774e-02
                                                              8.535
                                                                     < 2e-16
## INFANTS
                                      2.438e-01 8.133e-02
                                                             2.998 0.00272
##
## (Intercept)
                                     ***
## DISTANCE
                                     ***
## factor(HAUL_TYPE)DOMESTIC
                                     ***
## factor(HAUL_TYPE)INTERCONTINENTAL
## factor(TRIP_TYPE)ONE_WAY
                                     ***
## factor(TRIP_TYPE)ROUND_TRIP
## ADULTS
                                     ***
## CHILDREN
## INFANTS
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
  (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 49478
                             on 49999
                                       degrees of freedom
## Residual deviance: 46638
                             on 49991 degrees of freedom
## AIC: 46656
##
## Number of Fisher Scoring iterations: 5
```

However, considering adults childrens and infants as levels, it seems that having two children or one infant highly increases the change of selecting luggage. It might me interesting for the the sake of parsimony to remove the adult category.

```
##
## Call:
```

```
## glm(formula = EXTRA_BAGGAGE ~ DISTANCE + factor(HAUL_TYPE) +
##
       factor(TRIP_TYPE) + factor(ADULTS) + factor(CHILDREN) + factor(INFANTS),
##
       family = binomial(link = "logit"), data = train)
##
## Deviance Residuals:
                    Median
##
      Min
                1Q
                                  3Q
                                          Max
## -1.3694 -0.7246 -0.5959 -0.3293
                                        2.6493
##
## Coefficients:
##
                                       Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                                     -1.187e+01 1.008e+02 -0.118 0.90624
## DISTANCE
                                     -5.727e-05 8.502e-06 -6.736 1.62e-11
## factor(HAUL_TYPE)DOMESTIC
                                     -4.314e-01
                                                2.897e-02 -14.891 < 2e-16
## factor(HAUL_TYPE)INTERCONTINENTAL -1.113e+00
                                                4.385e-02 -25.375 < 2e-16
                                     -3.974e-01 5.177e-02 -7.676 1.64e-14
## factor(TRIP_TYPE)ONE_WAY
## factor(TRIP_TYPE)ROUND_TRIP
                                     -7.448e-01
                                                5.051e-02 -14.747 < 2e-16
## factor(ADULTS)1
                                                1.008e+02
                                     1.116e+01
                                                            0.111 0.91183
## factor(ADULTS)2
                                     1.180e+01
                                                1.008e+02
                                                            0.117 0.90676
## factor(ADULTS)3
                                     1.172e+01 1.008e+02
                                                            0.116 0.90744
## factor(ADULTS)4
                                     1.197e+01
                                                1.008e+02
                                                            0.119 0.90544
## factor(ADULTS)5
                                     1.184e+01
                                                1.008e+02
                                                            0.117
                                                                   0.90651
## factor(ADULTS)6
                                     1.225e+01 1.008e+02
                                                            0.122 0.90324
## factor(ADULTS)7
                                     1.186e+01 1.008e+02
                                                            0.118 0.90632
## factor(ADULTS)8
                                     1.138e+01
                                                1.008e+02
                                                            0.113
                                                                   0.91010
## factor(ADULTS)9
                                     1.168e+01 1.008e+02
                                                            0.116 0.90776
## factor(CHILDREN)1
                                    2.991e-01 5.037e-02
                                                            5.938 2.89e-09
## factor(CHILDREN)2
                                                7.818e-02
                                                            6.102 1.05e-09
                                     4.770e-01
## factor(CHILDREN)3
                                     3.966e-01
                                                1.890e-01
                                                            2.099 0.03584
## factor(CHILDREN)4
                                                6.272e-01 -1.133 0.25734
                                    -7.105e-01
## factor(CHILDREN)5
                                    1.354e+01 3.247e+02
                                                            0.042 0.96673
## factor(INFANTS)1
                                     2.346e-01 8.371e-02
                                                            2.803
                                                                   0.00507
## factor(INFANTS)2
                                    -1.169e+01 8.089e+01 -0.145 0.88504
##
## (Intercept)
## DISTANCE
## factor(HAUL TYPE)DOMESTIC
                                     ***
## factor(HAUL TYPE)INTERCONTINENTAL
## factor(TRIP_TYPE)ONE_WAY
                                     ***
## factor(TRIP TYPE)ROUND TRIP
                                     ***
## factor(ADULTS)1
## factor(ADULTS)2
## factor(ADULTS)3
## factor(ADULTS)4
## factor(ADULTS)5
## factor(ADULTS)6
## factor(ADULTS)7
## factor(ADULTS)8
## factor(ADULTS)9
## factor(CHILDREN)1
                                     ***
## factor(CHILDREN)2
                                     ***
## factor(CHILDREN)3
## factor(CHILDREN)4
## factor(CHILDREN)5
## factor(INFANTS)1
```

```
## factor(INFANTS)2
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
   (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 49478 on 49999 degrees of freedom
## Residual deviance: 46313 on 49978 degrees of freedom
## AIC: 46357
##
## Number of Fisher Scoring iterations: 11
If we remove the categories using the family size feature, we get:
##
## Call:
## glm(formula = EXTRA_BAGGAGE ~ DISTANCE + factor(HAUL_TYPE) +
       factor(TRIP_TYPE) + factor(DEVICE) + factor(COMPANY) + factor(FAMILY_SIZE),
##
##
       family = binomial(link = "logit"), data = train)
##
## Deviance Residuals:
##
      Min
                10
                    Median
                                  30
                                          Max
## -1.3171 -0.7225 -0.5624 -0.3321
                                        2.5300
##
## Coefficients:
##
                                      Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                                    -8.204e-01 5.954e-02 -13.777 < 2e-16
## DISTANCE
                                    -4.363e-05 9.452e-06 -4.616 3.92e-06
## factor(HAUL_TYPE)DOMESTIC
                                    -4.007e-01 3.271e-02 -12.249
                                                                   < 2e-16
## factor(HAUL_TYPE)INTERCONTINENTAL -1.205e+00 5.070e-02 -23.768 < 2e-16
## factor(TRIP_TYPE)ONE_WAY
                                    -3.155e-01 5.853e-02 -5.390 7.04e-08
## factor(TRIP_TYPE)ROUND_TRIP
                                    -7.060e-01 5.709e-02 -12.367 < 2e-16
## factor(DEVICE)OTHER
                                    -1.215e+01 8.917e+01 -0.136 0.89159
## factor(DEVICE)SMARTPHONE
                                    -2.078e-01 3.296e-02 -6.303 2.91e-10
## factor(DEVICE)TABLET
                                    -5.104e-02 5.501e-02 -0.928 0.35350
## factor(COMPANY)GO VOYAGE
                                     3.616e-01 4.402e-02
                                                            8.214
                                                                   < 2e-16
## factor(COMPANY)OPODO
                                     1.871e-01 3.022e-02
                                                            6.193 5.90e-10
## factor(COMPANY)OTHER
                                     1.239e+01 8.917e+01
                                                            0.139 0.88948
## factor(FAMILY_SIZE)2
                                     6.127e-01 2.936e-02 20.865 < 2e-16
## factor(FAMILY SIZE)3
                                     6.410e-01 5.105e-02 12.556
                                                                   < 2e-16
## factor(FAMILY_SIZE)4
                                     8.750e-01 6.096e-02 14.353 < 2e-16
## factor(FAMILY SIZE)5
                                     8.928e-01 1.035e-01
                                                            8.625 < 2e-16
## factor(FAMILY SIZE)6
                                     9.211e-01 1.558e-01
                                                            5.912 3.38e-09
## factor(FAMILY SIZE)7
                                     1.236e+00 2.609e-01
                                                            4.738 2.16e-06
## factor(FAMILY_SIZE)8
                                     6.004e-01 3.555e-01
                                                            1.689 0.09127
## factor(FAMILY_SIZE)9
                                     1.047e+00 4.039e-01
                                                            2.592 0.00955
##
## (Intercept)
                                    ***
## DISTANCE
                                     ***
## factor(HAUL_TYPE)DOMESTIC
## factor(HAUL_TYPE)INTERCONTINENTAL ***
## factor(TRIP_TYPE)ONE_WAY
                                    ***
## factor(TRIP_TYPE)ROUND_TRIP
## factor(DEVICE)OTHER
## factor(DEVICE)SMARTPHONE
                                    ***
```

```
## factor(DEVICE)TABLET
## factor(COMPANY)GO VOYAGE
                                      ***
## factor(COMPANY)OPODO
## factor(COMPANY)OTHER
## factor(FAMILY_SIZE)2
## factor(FAMILY_SIZE)3
## factor(FAMILY_SIZE)4
## factor(FAMILY_SIZE)5
## factor(FAMILY_SIZE)6
## factor(FAMILY_SIZE)7
## factor(FAMILY_SIZE)8
## factor(FAMILY_SIZE)9
##
                     '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
##
   (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 39500
                             on 39889
                                       degrees of freedom
## Residual deviance: 36964
                             on 39870
                                       degrees of freedom
     (110 observations deleted due to missingness)
## AIC: 37004
##
## Number of Fisher Scoring iterations: 12
## Area under the curve: 0.6895
```

ROC Curve 1.0 1.0 False Positive Rate

First iteration of the logistic regression model gives an AUC of 0.6895

Overall one of the challenges of building this model is that there is strong class imbalance - indeed it might

be interesting to try an xgboost model with the data encoded as levels rather than as dummy variables Preliminary 80/20 train/validation split to have internal validation mechanism

${\bf Graph~Boosted~Machine~with~XGBoost}$