

Sales Pipeline Overview

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Business to business (B2B) sales data has thus far been neglected in the machine learning revolution over the past 5 years. B2B sales teams hold one of the greatest responsibilities in modern global business. They hold the key to the relationship to their largest customers. Our goal here in this section is to look at the pipeline and data and understand it.

Is Sales an Art or a Science?

Sales is neither an art or a science, but a process that can be managed just like any other business process. There are many moving parts to not only an individual deal, but take that deal and multiply it by a few thousand over global territories and cultures you have the modern large global sales force.

With a proper customer relationship management (CRM) system companies can collect millions of data points. That's just what is in the CRM system. Now companies have the ability to monitor sales calls with video through meeting software and voice, now giving us billions of data points. Sales needs the attention of your analytics or data science department.

About the data

This dataset is sample data from IBM's Watson Analytics. The data is from a CRM system that has been run through a process to make it somewhat model ready for analysis. The purpose of the data is to discover patterns with sales wins and losses. This could allow a Sales Operations team to communicate to executive leadership what factors contribute to winning and losing. That allows the organization to be proactive and possibly change the outcome.

This is a flat file that you could get out of a relational database through a data warehouse. Let's take a look at the data.

```
pipeline <- read.csv("pipelineAnalytics.csv")
head(pipeline)
```

##	Opportunity.Number	Supplies.Subgroup	Supplies.Group	Region
## 1	1641984	Exterior Accessories	Car Accessories	Northwest
## 2	1658010	Exterior Accessories	Car Accessories	Pacific
## 3	1674737	Motorcycle Parts	Performance & Non-auto	Pacific
## 4	1675224	Shelters & RV	Performance & Non-auto	Midwest
## 5	1689785	Exterior Accessories	Car Accessories	Pacific
## 6	1692390	Shelters & RV	Performance & Non-auto	Pacific
##	Route.To.Market	Elapsed.Days.In.Sales.Stage	Opportunity.Result	
## 1	Fields Sales	76	Won	
## 2	Reseller	63	Loss	
## 3	Reseller	24	Won	
## 4	Reseller	16	Loss	
## 5	Reseller	69	Loss	
## 6	Reseller	89	Loss	
##	Sales.Stage.Change.Count	Total.Days.Identified.Through.Closing		
## 1	13	104		
## 2	2	163		

```

## 3          7          82
## 4          5         124
## 5         11          91
## 6          3         114
##   Total.Days.Identified.Through.Qualified Opportunity.Amount.USD
## 1          101          0
## 2          163          0
## 3           82        7750
## 4          124          0
## 5           13       69756
## 6           0      232522
##   Client.Size.By.Revenue Client.Size.By.Employee.Count
## 1           5          5
## 2           3          5
## 3           1          1
## 4           1          1
## 5           1          1
## 6           5          1
##   Revenue.From.Client.Past.Two.Years Competitor.Type
## 1           0      Unknown
## 2           0      Unknown
## 3           0      Unknown
## 4           0       Known
## 5           0      Unknown
## 6           0      Unknown
##   Ratio.Days.Identified.To.Total.Days Ratio.Days.Validated.To.Total.Days
## 1           0.69636          0.113985
## 2           0.00000          1.000000
## 3           1.00000          0.000000
## 4           1.00000          0.000000
## 5           0.00000          0.141125
## 6           0.00000          0.000877
##   Ratio.Days.Qualified.To.Total.Days Deal.Size.Category
## 1           0.154215          1
## 2           0.000000          1
## 3           0.000000          1
## 4           0.000000          1
## 5           0.000000          4
## 6           0.000000          5

```

```
dim(pipeline)
```

```
## [1] 78025    19
```

The pipeline data is 78K rows with 19 variables. Let's explore some of the variables. Opp is short for opportunity.

Variable	Data Type	Description
Opportunity Number	Integer	Unique identifier some opps may have multiple products
Supplies Subgroup	Factor	Product subgroup

Variable	Data Type	Description
Supplies Group	Factor	Product grouping
Region	Factor	Business region in which opp was sold
Route to Market	Factor	The channel in which owns the opp
Elapsed Days In Sales Stage	Integer	The number of days an opp has been in the sales stage
Opportunity Result	Factor	Tells if the opp was won or lost
Sales Stage Change Count	Factor	How many times has the stage name changed
Total Days Identified Through Closing	Integer	The age of an opp
Total.Days.Identified.Through.Qualified	Integer	How long it took in days to qualify an opp
Opportunity Amount USD	Integer	The revenue over a 12 month period the opp represents
Client Size By Revenue	Integer	Client size by the clients yearly revenue
Client Size by Employee Count	Integer	Client size by the number of employees
Revenue From Client Past Two Years	Integer	Client spend past two years
Competitor Type	Factor	Indicator if a competitor has been identified
Ratio Days Identified To Total Days	Integer	Ratio of total days the opportunity has spent in sales stage: Identified/Validating over total days in sales process

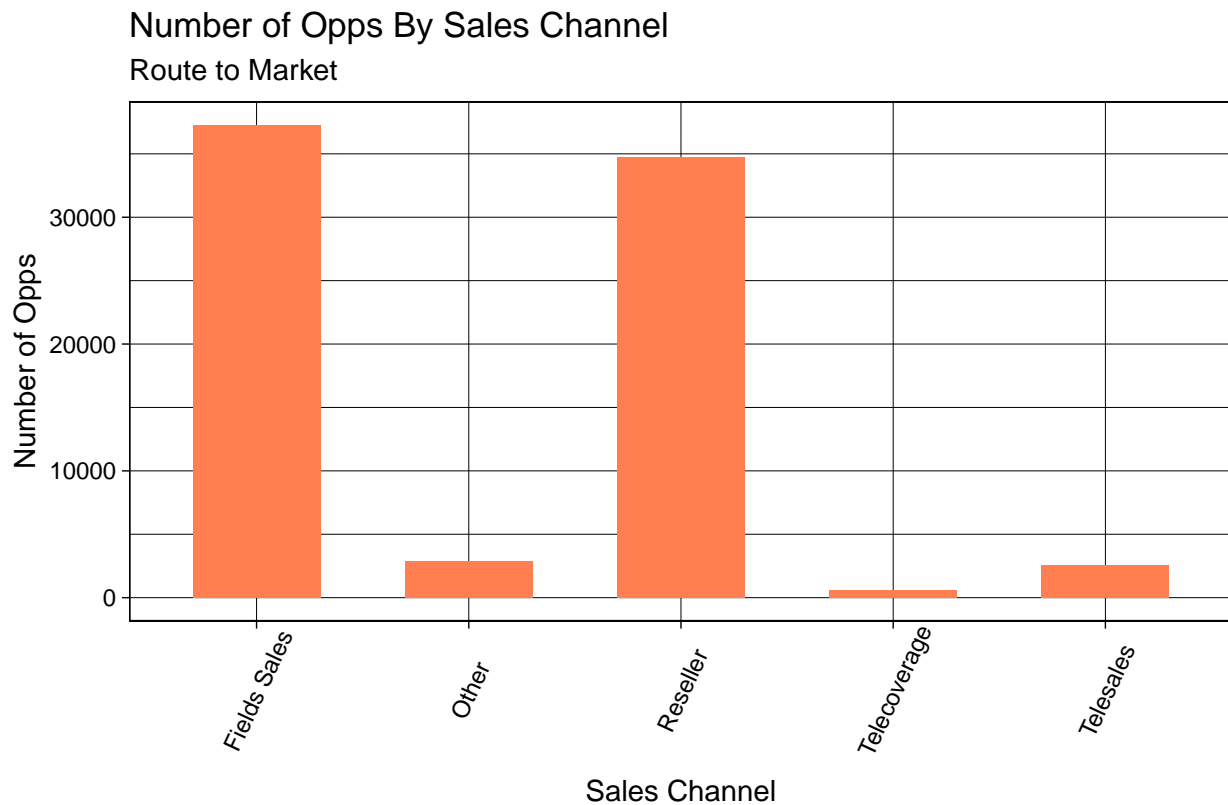
Variable	Data Type	Description
Ratio Days Validated To Total.Days	Integer	Ratio of total days the Opportunity has presence in sales stage: Validated/Qualifying over total days in sales process
Ratio Days Qualified To Total Days	Integer	Qualified/Gaining Agreement over total days in sales process
Deal Size Category	Integer	Categorical size of the opportunity size

Analysis Overview

Lets visualize some of this information to do this we are going to use ggplot.

```
library(ggplot2)
theme_set(theme_linedraw())
options(scipen = 999)

w <- ggplot(pipeline, aes(Route.To.Market))
w + geom_bar(width = 0.6, fill = "coral") +
  labs(title = "Number of Opps By Sales Channel",
        subtitle = "Route to Market",
        caption = "Source: IBM Watson Win Loss Analysis dataset",
        y = "Number of Opps",
        x = "Sales Channel") +
  theme(axis.text.x = element_text(angle = 65, vjust = 0.6))
```



Source: IBM Watson Win Loss Analysis dataset

```
table(pipeline$Route.To.Market)
```

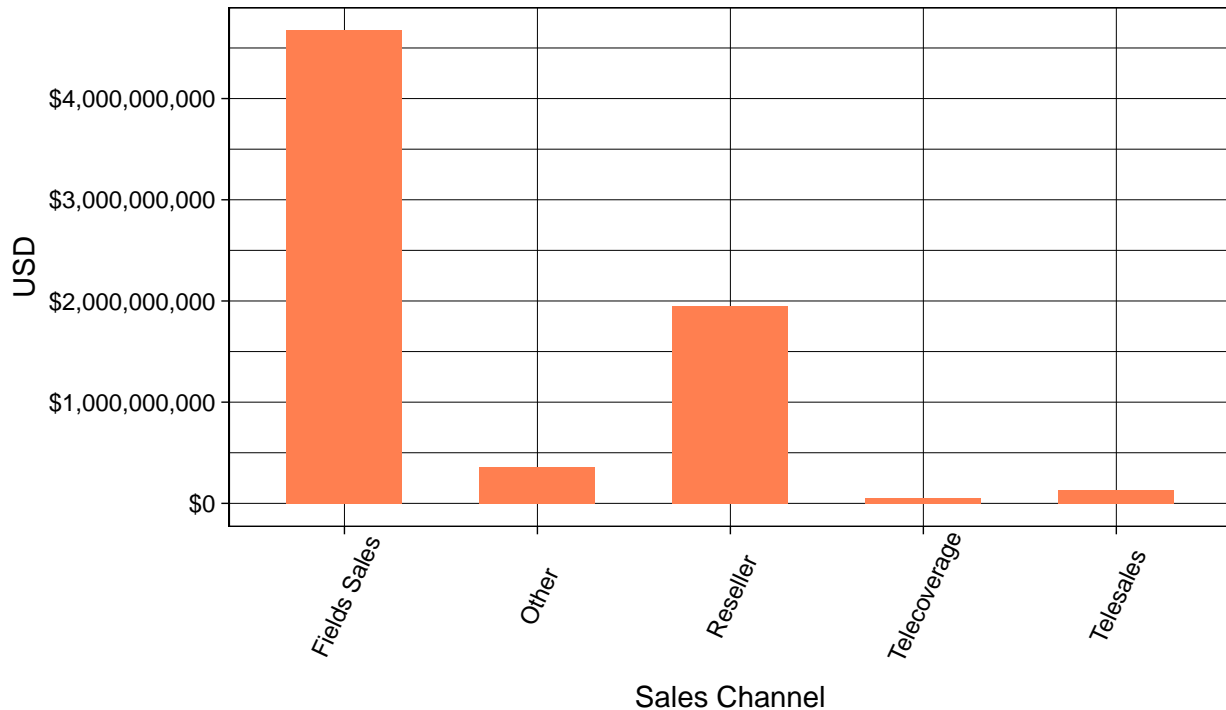
```
##
## Fields Sales      Other      Reseller Telecoverage  Telesales
##      37262        2856      34758         619      2530
```

Most of the opportunities come from field sales and the reseller channel.

```
w1 <- ggplot(pipeline, aes(x = Route.To.Market, y = Opportunity.Amount.USD))
w1 + geom_bar(stat = "identity", width = 0.6, fill = "coral") +
  labs(title = "Total Opportunity $ By Sales Channel",
        subtitle = "Route to Market",
        caption = "Source: IBM Watson Win Loss Analysis dataset",
        y = "USD",
        x = "Sales Channel") +
  scale_y_continuous(labels = scales::dollar) +
  theme(axis.text.x = element_text(angle = 65, vjust = 0.6))
```

Total Opportunity \$ By Sales Channel

Route to Market



Source: IBM Watson Win Loss Analysis dataset

```
tapply(pipeline$Opportunity.Amount.USD, pipeline$Route.To.Market, FUN = sum)
```

```
## Fields Sales      Other      Reseller Telecoverage      Telesales
## 4669514062      357806122 1948586216      46515153      127575717
```

This is a pretty large dollar value of pipeline. The total pipeline for this sample for field sales for example is 4.7 billion dollars. The reseller channel is worth 1.95B billion dollars. This looks like a typically B2B pipeline in that our field sales generate the most opportunity while we have other channels that might and this is a guess that smaller opportunities are managed by Resellers. Let's check that guess.

```
rtmTBL <- table(pipeline$Route.To.Market, pipeline$Client.Size.By.Revenue)
rtmTBL
```

```
##
##           1      2      3      4      5
## Fields Sales 27112 1837 2075 2678 3560
## Other        2160  81   146  178  291
## Reseller     27417 1880 2469 1690 1302
## Telecoverage  558   10   16   13   22
## Telesales    2257  33   50   77  113
```

```
prop.table(rtmTBL, 1)
```

```
##
##           1      2      3      4      5
## Fields Sales 0.72760453 0.04929955 0.05568676 0.07186946 0.09553969
## Other        0.75630252 0.02836134 0.05112045 0.06232493 0.10189076
## Reseller     0.78879682 0.05408827 0.07103401 0.04862190 0.03745900
## Telecoverage 0.90145396 0.01615509 0.02584814 0.02100162 0.03554120
```

```
## Telesales 0.89209486 0.01304348 0.01976285 0.03043478 0.04466403
```

The relationship is not with Client Size By Revenue.

```
cltTBL <- table(pipeline$Route.To.Market, pipeline$Revenue.From.Client.Past.Two.Years)
cltTBL
```

```
##
##           0      1      2      3      4
## Fields Sales 32446  590   861  1233  2132
## Other       2564   48    54   61   129
## Reseller    31332 1082  1091   731   522
## Telecoverage 591   10    12    1    5
## Telesales   2275   52    65   66   72
```

```
prop.table(cltTBL, 1)
```

```
##
##           0           1           2           3           4
## Fields Sales 0.870753046 0.015833825 0.023106650 0.033090011 0.057216467
## Other       0.897759104 0.016806723 0.018907563 0.021358543 0.045168067
## Reseller    0.901432764 0.031129524 0.031388457 0.021031130 0.015018125
## Telecoverage 0.954765751 0.016155089 0.019386107 0.001615509 0.008077544
## Telesales   0.899209486 0.020553360 0.025691700 0.026086957 0.028458498
```

Nor is it by spend...

```
oppTBL <- table(pipeline$Route.To.Market, pipeline$Deal.Size.Category)
oppTBL
```

```
##
##           1      2      3      4      5      6      7
## Fields Sales 4697  4235  3794  6316 12986  3592 1642
## Other       449   304   293   486   907   288   129
## Reseller    6081  9965  7442  6251  3640   970   409
## Telecoverage 97   107    89   144   142    32    8
## Telesales   771   512   350   431   399    52   15
```

```
prop.table(oppTBL, 1)
```

```
##
##           1           2           3           4           5
## Fields Sales 0.126053352 0.113654662 0.101819548 0.169502442 0.348505180
## Other       0.157212885 0.106442577 0.102591036 0.170168067 0.317577031
## Reseller    0.174952529 0.286696588 0.214108982 0.179843489 0.104724092
## Telecoverage 0.156704362 0.172859451 0.143780291 0.232633279 0.229402262
## Telesales   0.304743083 0.202371542 0.138339921 0.170355731 0.157707510
##
##           6           7
## Fields Sales 0.096398476 0.044066341
## Other       0.100840336 0.045168067
## Reseller    0.027907244 0.011767075
## Telecoverage 0.051696284 0.012924071
## Telesales   0.020553360 0.005928854
```

It looks like there is some relationship between the channel and how large the opportunity is. Not enough information to go off of since we could have a global business and they use resellers even with large opportunities. It would appear that telesales is used for smaller opportunities.

Basic Sales Metrics

Here we are going to take a look at some basic sales metrics like Close Rate and time to close. Lets start with Close Rates.

Close Rates

Close Rates in sales is very important. Its often how we start a basic forecast in B2B sales. Close rate is simple the number of opportunties sold or won divided by the total number of oportunties.