

Substandard Deviations Analysis

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Data Wrangling

Load relevant R libraries

```
library(tidyverse)
library(readxl)
library(lubridate)
library(ggplot2)
library(hrbrthemes)
library(gcookbook)

sales <- read_excel("product_sales.xlsx", sheet = 1)

head(sales)

## # A tibble: 6 x 4
##   month                company  product  total_sales_USD
##   <dtm>                <chr>    <chr>         <dbl>
## 1 2009-03-01 00:00:00 Voltswitch Nova 1         127411
## 2 2009-03-01 00:00:00 Voltswitch Nova 1 M         138302
## 3 2009-03-01 00:00:00 Voltswitch Nova 2         131839
## 4 2009-03-01 00:00:00 Voltswitch Mycro 1             0
## 5 2009-03-01 00:00:00 Voltswitch Nova 3             0
## 6 2009-03-01 00:00:00 Voltswitch Mycro 2             0
```

Data transformation of Voltswitch sales

We see the structure of the data and wish to transform it into a format that can be merged with other licensee sales data for comparison and regression. Specifically we want to know their total sales, total sales of products using InfiniCharge and what percentage of their revenue came from infringing technology.

```
mycro_line <- c("Mycro 1", "Mycro 2", "Mycro 3")

total_sales_volt <- sales %>%
  filter(company == "Voltswitch") %>%
  mutate(year = year(month)) %>%
  group_by(year) %>%
  summarise(total_sales = sum(total_sales_USD))

total_sales_volt_mycro <- sales %>%
  filter(product %in% mycro_line) %>%
  mutate(year = year(month)) %>%
  group_by(year) %>%
  summarise(total_sales_mycro = sum(total_sales_USD))

volt_sales <- merge(total_sales_volt, total_sales_volt_mycro)
```

```
volt_sales1 <- volt_sales %>%
  mutate(percent_patent = total_sales_mycro / total_sales)

head(volt_sales1)
```

```
##   year total_sales total_sales_mycro percent_patent
## 1 2009      4817169           853537      0.1771864
## 2 2010      8626491          2566663      0.2975327
## 3 2011     11949673          4302745      0.3600722
## 4 2012     13539102          5151277      0.3804741
## 5 2013     13848538          4419741      0.3191486
## 6 2014      9490083          2146065      0.2261376
```

Data transformation of licensee sales

From here we load, clean and merge the sales data from the other licensees to merge with the Voltswitch data.

```
companies <- c("Year", "Rekall", "Tyrell Corp.", "Sirius Cybernetics",
              "Cyberdyne Systems", "Nakatomi Corp.")

licensed_sales <- read_excel("sales_data.xlsx", sheet = 1, cell_rows(7:14))

licensed_sales_clean <- licensed_sales[, c(1, 3, 4, 5, 6, 7)]
names(licensed_sales_clean) <- companies

licensed_patent_clean <- licensed_sales[, c(1, 9, 10, 11, 12, 13)]
names(licensed_patent_clean) <- companies

lsc <- licensed_sales_clean %>%
  gather(company, sales, -Year)

lpc <- licensed_patent_clean %>%
  gather(company, percent, -Year)

clean_sales_patents <- merge(lsc, lpc) %>%
  arrange(Year, company)

head(clean_sales_patents)
```

```
##   Year      company      sales  percent
## 1 2008 Cyberdyne Systems 103638.3 0.7368421
## 2 2008   Nakatomi Corp.      0.0 0.0000000
## 3 2008      Rekall    660776.3 0.5188679
## 4 2008 Sirius Cybernetics      0.0 0.0000000
## 5 2008   Tyrell Corp.   382268.2 1.0000000
## 6 2009 Cyberdyne Systems 4718794.4 0.7473684
```

Merge Voltswitch with other licensees

We now are able to merge the Voltswitch data with other licensees to view a combined data set ready for regression.

```

volt_sales2 <- volt_sales1 %>%
  mutate(Year = year) %>%
  mutate(company = "Voltswitch") %>%
  mutate(sales = total_sales) %>%
  mutate(percent = percent_patent) %>%
  select(Year, company, sales, percent)

total_sales_df <- rbind(clean_sales_patents, volt_sales2) %>% arrange(Year, company)

royalty_schedule <- tibble(
  company = companies[-1],
  royalty_percent = c(0.05, 0.07, 0.02, 0.03, 0.01))

reg_tot1 <- total_sales_df %>%
  merge(royalty_schedule) %>%
  mutate(royalties_due = sales * percent * royalty_percent)

reg_tot2 <- reg_tot1 %>%
  group_by(company) %>%
  summarise(tot_sales = sum(sales),
            avg_percent = mean(percent),
            royalty_percent = mean(royalty_percent))

head(reg_tot2)

## # A tibble: 5 x 4
##   company          tot_sales avg_percent royalty_percent
##   <chr>              <dbl>      <dbl>         <dbl>
## 1 Cyberdyne Systems 60708026.    0.746         0.03
## 2 Nakatomi Corp.    85491386.    0.122         0.01
## 3 Rekall            6340256.    0.488         0.05
## 4 Sirius Cybernetics 45640221.    0.202         0.02
## 5 Tyrell Corp.      11729122.    1             0.07

```

Analysis and Modeling

Estimate reasonable royalties

Here we use linear regression to estimate the best fit given the sales data and share of sales by InfiniCharge products. We also calculate the royalties due.

```

roy_mod <- lm(royalty_percent ~ tot_sales + avg_percent, reg_tot2)

volt_sales3 <- volt_sales2 %>%
  summarise(company = "Voltswitch",
            tot_sales = sum(sales),
            avg_percent = mean(percent))

pred_roy_percent <- predict(roy_mod, volt_sales3)

volt_sales3$royalty_percent <- pred_roy_percent

```

```
reg_tot3 <- rbind(reg_tot2, volt_sales3)

reg_tot3 %>%
  mutate(royalties_due = tot_sales * avg_percent * royalty_percent)

## # A tibble: 6 x 5
##   company          tot_sales avg_percent royalty_percent royalties_due
##   <chr>              <dbl>      <dbl>          <dbl>          <dbl>
## 1 Cyberdyne Systems 60708026.    0.746          0.03        1358399.
## 2 Nakatomi Corp.    85491386.    0.122          0.01        104422.
## 3 Rekall            6340256.    0.488          0.05        154661.
## 4 Sirius Cybernetics 45640221.    0.202          0.02        184174.
## 5 Tyrell Corp.      11729122.    1              0.07        821039.
## 6 Voltswitch        62271056    0.293          0.0200       365528.
```

Validation of reasonable royalties to defeat Georgia-Pacific defense.

We look to validate that the estimated royalty given the presence of similar licenses is profitable to demonstrate mutual gains to trade for the license. We should expect the price to be above the cost of the license since there are no close substitutes. We use the split- apply-combine methodology.

```
results <- read_excel("survey_results.xlsx", sheet = "Survey Results", cell_rows(4:54))

key <- read_excel("survey_results.xlsx", sheet = "Question Key", cell_rows((4:28)))

results1 <- results[, -1]

split_prices <- split(results1, seq(1:50))
prices_lists <- lapply(split_prices, t)
prices_lists <- lapply(prices_lists, cbind, key)
tester <- bind_rows(prices_lists)

tester_names <- c("Price", "Question", "Talk_Time", "Weight", "PPI", "Camera")
names(tester) <- tester_names

head(tester)
```

```
##   Price Question Talk_Time Weight PPI Camera
## 1   205         1         10     4 300     12
## 2   185         2         10     4 300      8
## 3   210         3         10     4 400     12
## 4   200         4         10     4 400      8
## 5   185         5         10     5 300     12
## 6   170         6         10     5 300      8
```

```
tester_model <- lm(Price ~ Talk_Time + Weight + PPI + Camera, tester)
```

```
tester_model
```

```
##
## Call:
## lm(formula = Price ~ Talk_Time + Weight + PPI + Camera, data = tester)
##
## Coefficients:
```

```
## (Intercept)    Talk_Time      Weight      PPI      Camera
##      3.1750      9.9954     -10.5750     0.2477     5.0062
```

```
anova(tester_model)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Response: Price
```

```
##           Df    Sum Sq Mean Sq  F value    Pr(>F)
## Talk_Time   1 18649337 18649337 227329.12 < 2.2e-16 ***
## Weight      1   33549    33549    408.95 < 2.2e-16 ***
## PPI         1   184140   184140   2244.61 < 2.2e-16 ***
## Camera      1   120300   120300   1466.42 < 2.2e-16 ***
## Residuals 1195    98034      82
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Predicting prices for different products

Given the model for prices based on the survey results, we now use that to price best alternatives to satisfy the but-for process. But-for the infringement what would Voltswitch's alternatives have been? We price those and analyze whether those alternatives were as profitable as using InfiniCharge.

```
desc <- read_excel("product_descriptions.xlsx", sheet = 1, cell_rows(6:23))
```

```
marg <- read_excel("product_margins.xlsx", sheet = 1, cell_rows(6:23))
```

```
desc1 <- desc[, -c(2,5)]
```

```
marg1 <- marg[, -2]
```

```
desc_marg <- merge(desc1, marg1)
```

```
names(desc_marg) <- c("Product", "Parent_Co", "Release_Date", "Talk_Time", "Weight",
  "Screen", "Memory", "Camera", "PPI", "Margin")
```

```
head(desc_marg)
```

```
##   Product Parent_Co Release_Date Talk_Time Weight Screen Memory Camera
## 1 Blade 1 Voltswitch 2011-03-01      20   11.8   4.70      10    5.0
## 2 Blade 2 Voltswitch 2012-07-01      25    8.0   5.00      10    6.7
## 3 Blade 3 Voltswitch 2013-03-01      20    6.1   5.40      12    6.7
## 4 Blade 4 Voltswitch 2013-07-01      25    5.7   5.40      12    9.2
## 5 Blade 5 Voltswitch 2014-03-01      20    5.4   5.60      18    8.0
## 6 Blade 6 Voltswitch 2014-07-01      15    4.0   5.75      16    8.0
##   PPI Margin
## 1 198    0.4
## 2 223    0.6
## 3 265    0.4
## 4 265    0.6
## 5 297    0.4
## 6 297    0.6
```

```
dm_data <- desc_marg %>%
```

```
  select(Talk_Time, Weight, PPI, Camera)
```

```
prod_preds <- predict(tester_model, dm_data)
```

```
desc_marg$pred <- prod_preds
```

```

desc_marg <- desc_marg %>%
  mutate(profits = pred * Margin) %>%
  mutate(pred_royalty = pred * 0.02) %>%
  mutate(profit_over_license = profits - pred_royalty) %>%
  arrange(profits)

tail(desc_marg, n= 9)

```

##	Product	Parent_Co	Release_Date	Talk_Time	Weight	Screen	Memory
## 9	Blade 6	Voltswitch	2014-07-01	15	4.0	5.75	16
## 10	Blade 2	Voltswitch	2012-07-01	25	8.0	5.00	10
## 11	Blade 4	Voltswitch	2013-07-01	25	5.7	5.40	12
## 12	Mycro 1	Voltswitch	2009-07-01	35	8.0	4.50	8
## 13	Mycro 3	Voltswitch	2011-07-01	35	8.2	5.00	10
## 14	Mycro 2	Voltswitch	2010-07-01	40	8.0	4.20	10
## 15	HAL 1000	Lyte Solutions	2012-05-01	50	9.0	4.50	10
## 16	HAL 2000	Lyte Solutions	2013-05-01	50	9.0	4.50	10
## 17	HAL 3000	Lyte Solutions	2014-05-01	60	9.0	4.50	12

##	Camera	PPI	Margin	pred	profits	pred_royalty	profit_over_license
## 9	8.0	297	0.6	224.4371	134.6623	4.488742	130.1735
## 10	6.7	223	0.6	257.2491	154.3494	5.144981	149.2045
## 11	9.2	265	0.6	304.4927	182.6956	6.089854	176.6058
## 12	2.0	174	0.6	321.5335	192.9201	6.430670	186.4894
## 13	5.0	207	0.6	342.6130	205.5678	6.852260	198.7155
## 14	4.2	180	0.6	384.0105	230.4063	7.680211	222.7261
## 15	5.0	200	0.5	482.3491	241.1746	9.646982	231.5276
## 16	5.0	223	0.5	488.0474	244.0237	9.760947	234.2627
## 17	5.0	264	0.5	598.1587	299.0793	11.963174	287.1162

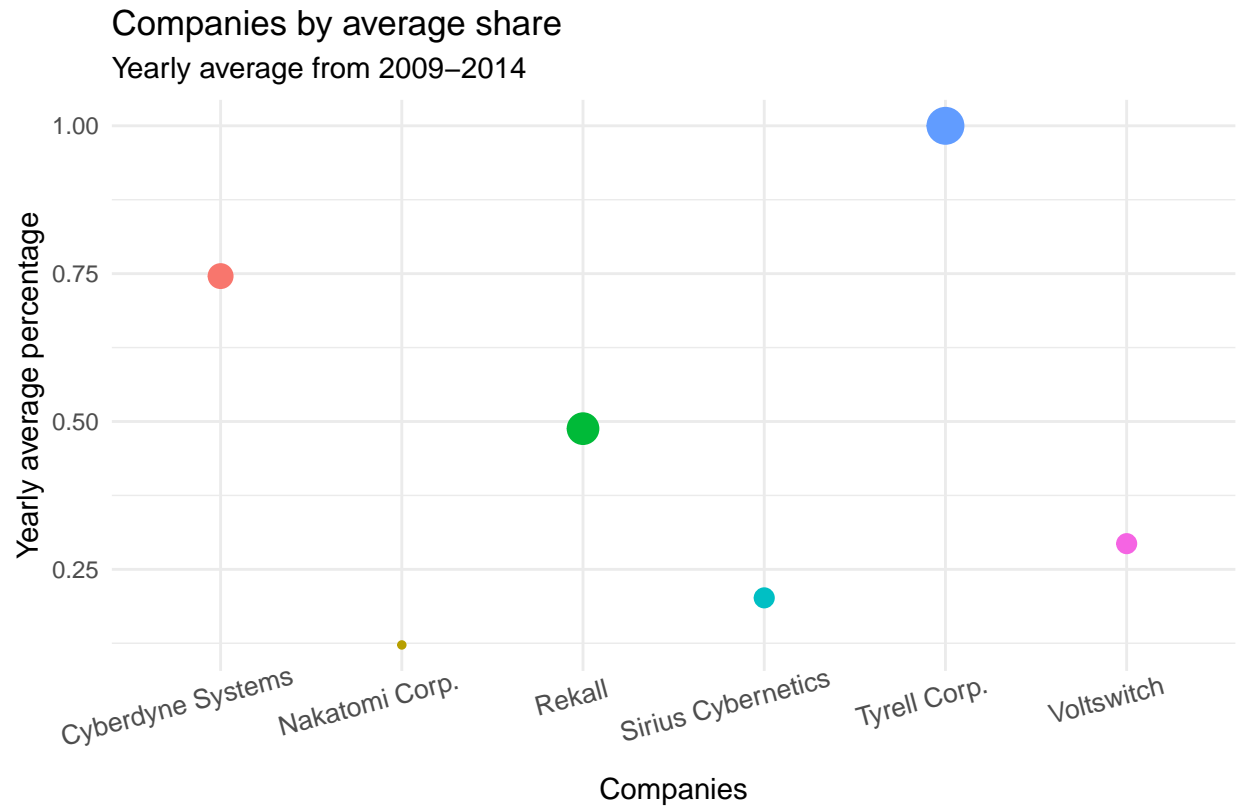
Validation through visualization

We create some visuals which convey the structure of the data to non-technical stakeholders

```

reg_tot3 %>%
  ggplot(aes(x = company, y = avg_percent, size = royalty_percent)) +
  guides(size = FALSE, colour = FALSE) +
  labs(x = "Companies",
       y = "Yearly average percentage",
       title = "Companies by average share",
       subtitle = "Yearly average from 2009-2014",
       caption = "We see that generally the royalty rate increases as the percentage increases.") +
  geom_point(aes(colour = company)) +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 15, hjust = 0.7, size = 10))

```

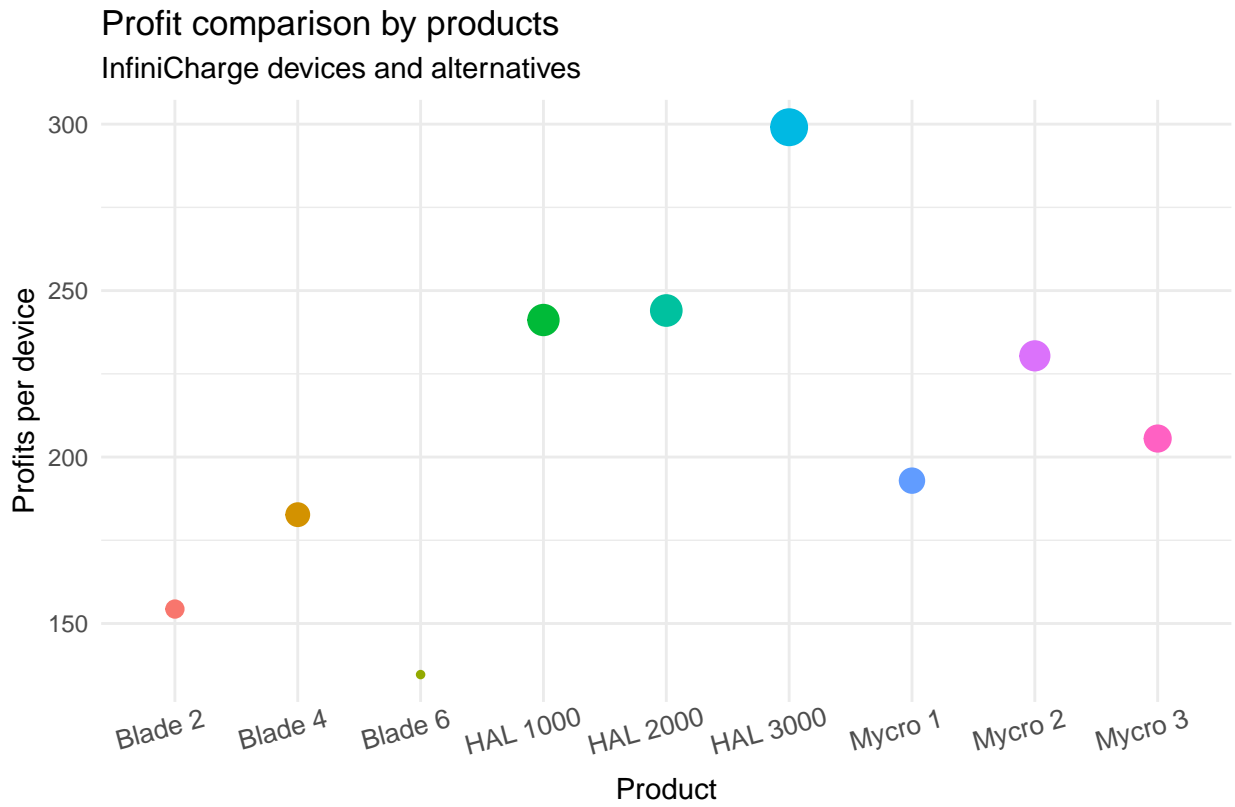


We see that generally the royalty rate increases as the percentage increases.

```
prof_dev <- desc_marg[9:17,]

pd_viz_df <- prof_dev[, c(1, 12)]

pd_viz_df %>%
  mutate(Product = as.factor(Product)) %>%
  ggplot(aes(x = Product, y = profits)) +
    guides(size = FALSE, colour = FALSE) +
    labs(x = "Product",
         y = "Profits per device",
         title = "Profit comparison by products",
         subtitle = "InfiniCharge devices and alternatives",
         caption = "We see that the profits of using InfiniCharge are greater per device than alternatives"),
  geom_point(aes(colour = Product, size = profits)) +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 15, hjust = 0.7, size = 10))
```



We see that the profits of using InfiniCharge are greater per device than alternative