

Case Study Report

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Data Wrangling

```
# Import the Dataset
kick.2018 <- read_csv("ks-projects-201801.csv")

# Sort out failed and successful projects
kick.2018 <- kick.2018 %>% filter(state %in% c("failed", "successful")) %>%
  mutate(diff_date = as.numeric(as.Date(deadline) - as.Date(str_extract(launched,
    "^.{10}")))))

# Random Sampling of the Data

## index <- sample(nrow(kick.2018), 2000, replace=FALSE)

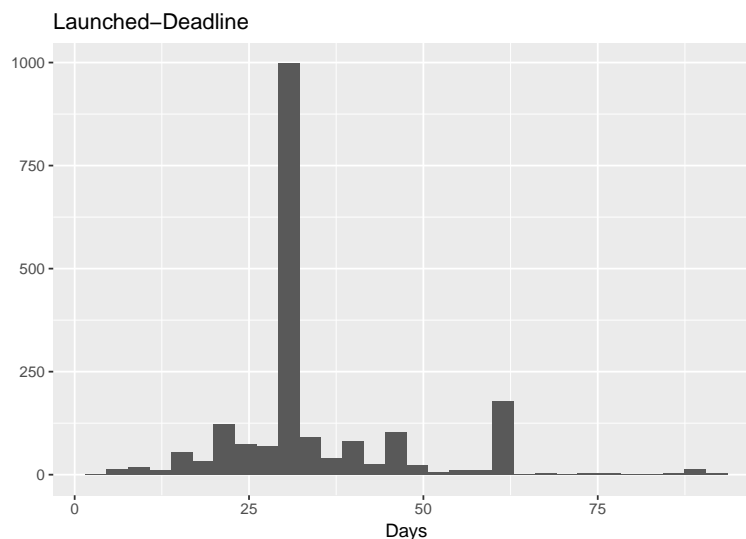
## kick.sample <- kick.2018[index, ]

## write.csv(kick.sample, file = 'Kickstarter_sample.csv', row.names = FALSE)
```

Launched-Deadline and Success/Failure

```
kickstarter <- read_csv("Kickstarter_sample.csv") ## pull in sample data

ggplot(kickstarter) +
  geom_histogram(aes(x = diff_date), bins = 30) +
  xlab("Days") +
  ylab("") +
  ggtitle("Launched-Deadline")
```



```

kickstarter$state <- factor(kickstarter$state)

failed <- subset(kickstarter, select = diff_date, subset = state == "failed", drop = T)
successful <- subset(kickstarter, select = diff_date, subset = state == "successful", drop = T)

ggplot(kickstarter, aes(x = diff_date, fill = state, color = state)) +
  geom_histogram(alpha=0.2, position="identity", bins = 30) +
  xlab("Days") +
  ylab("") +
  scale_fill_discrete(name="Successful/Failure",
                      breaks=c("failed", "successful"),
                      labels=c("Failed", "Successful")) +
  scale_color_discrete(name="Successful/Failure",
                      breaks=c("failed", "successful"),
                      labels=c("Failed", "Successful")) +
  ggtitle("Launched-Deadline") +
  theme(legend.position="bottom")

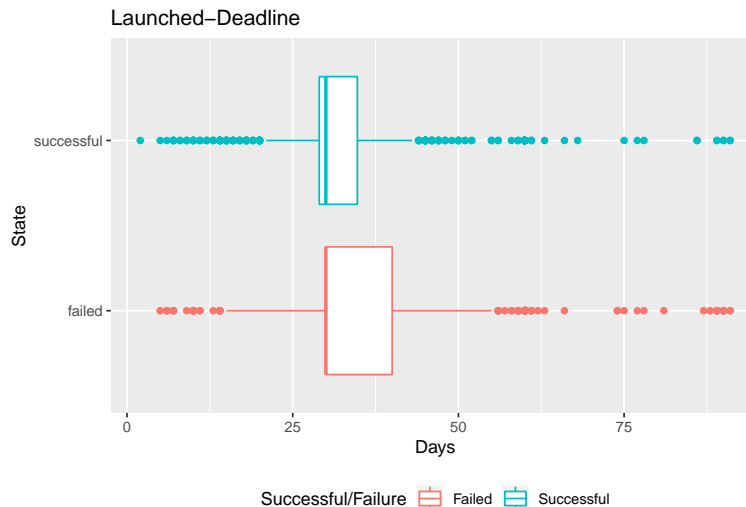
```



```

ggplot(kickstarter, aes(x = state, y = diff_date, color = state)) +
  geom_boxplot() +
  xlab("State") +
  ylab("Days") +
  scale_color_discrete(name="Successful/Failure",
                      breaks=c("failed", "successful"),
                      labels=c("Failed", "Successful")) +
  ggtitle("Launched-Deadline") +
  theme(legend.position="bottom") +
  coord_flip()

```



Confidence Interval

```

N <- 10^4

mean.diff <- mean(failed)-mean(successful)
se <- sqrt(var(failed)/length(failed)+var(successful)/length(successful))

boot.perc <- numeric(N)
Tstar <- numeric(N)

for (i in 1:N){
  failedBoot <- sample(failed, length(failed), replace=TRUE)
  successfulBoot <- sample(successful, length(successful), replace = TRUE)
  SEstar <- sqrt(var(failedBoot)/length(failedBoot)+var(successfulBoot)/length(successfulBoot))
  meanDiffBoot <- mean(failedBoot)-mean(successfulBoot)
  Tstar[i] <- (meanDiffBoot-mean.diff)/SEstar
  boot.perc[i] <- meanDiffBoot
}

formula.t.CI <- t.test(diff_date~state, data = kickstarter)$conf
boot.perc.CI <- quantile(boot.perc, c(0.025,0.975))
boot.t.CI <- mean.diff-quantile(Tstar, c(0.975,0.025))*se

```

Formula t CI (95%): (2.07, 4.357)

Bootstrap percentile CI (95%): (2.072, 4.351)

Bootstrap t CI (95%): (2.067, 4.339)

Hypothesis Testing

$$H_0 : \mu_{investment\ window, failed} = \mu_{investment\ window, successful}$$

$$H_A : \mu_{investment\ window, failed} \neq \mu_{investment\ window, successful}$$

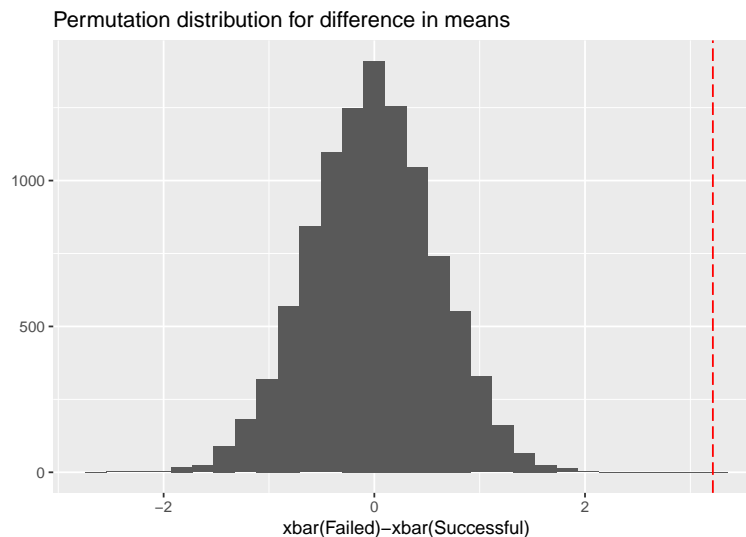
```

pooled.data <- unlist(kickstarter$diff_date)
perm <- numeric(N)

```

```
for (i in 1:N){
  index <- sample(length(pooled.data), size = length(failed), replace = FALSE)
  perm[i] <- mean(pooled.data[index]) - mean(pooled.data[-index])
}
```

```
ggplot(data.frame(perm), aes(x = perm)) +
  geom_histogram(bins = 30) +
  xlab("xbar(Failed)-xbar(Successful)") +
  ylab("") +
  geom_vline(xintercept=mean.diff, color = "red", linetype = "longdash") +
  ggtitle("Permutation distribution for difference in means")
```



```
p.perm <- (sum(perm >= mean.diff)+1)/(N+1)*2
p.t <- t.test(diff_date~state, data = kickstarter)$p.value
```

p-value (perm): $1.9998 \cdot 10^{-4}$

p-value (t): $4.0121047 \cdot 10^{-8}$

Success Rate before and after 2017

```
kick.2018 <- kick.2018 %>% mutate(before_2017 = ifelse(launched < "2017-01-01",
  "Yes", "No"))

before.2017 <- subset(kick.2018, select = state, subset = before_2017 == "Yes",
  drop = T)
after.2017 <- subset(kick.2018, select = state, subset = before_2017 == "No",
  drop = T)

succ.before.2017 <- sum(before.2017 == "successful")
succ.after.2017 <- sum(after.2017 == "successful")

prop.test(c(succ.before.2017, succ.after.2017), c(length(before.2017), length(after.2017)),
  correct = FALSE)
```

```
##
```

```
## 2-sample test for equality of proportions without continuity
```

```
## correction
##
## data:  c(succ.before.2017, succ.after.2017) out of c(length(before.2017), length(after.2017))
## X-squared = 93.689, df = 1, p-value < 2.2e-16
## alternative hypothesis: two.sided
## 95 percent confidence interval:
##  -0.02943464 -0.01946908
## sample estimates:
##      prop 1      prop 2
## 0.4006772 0.4251290
```

We are 95% confident that in average the proportion of successful projects after 2017 is 2 to 3% higher than before 2017.