

Airport-activity

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Airport activity:

Building a quick profile of daily this and that.

(Code chunk shortcut: **Command+Option+I**)

Lemme create a palette. Clear my working space, load some useful packages, and set my working drive.

```
cat("\014") # clear the console

rm(list=ls())
dev.off()

## null device
##          1
library(reshape2)
library(stringr)
library(plyr)
library(dplyr)

## Warning: package 'dplyr' was built under R version 3.4.2
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:plyr':
##
##   arrange, count, desc, failwith, id, mutate, rename, summarise,
##   summarize
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

library(formattable)
library(doBy)
library(ggplot2)

setwd("~/Dropbox/Work and research/Airport-activity/aviation-activity")
```

Colleagues in the Aviation Department drummed up data on weekday activity for a few random weekdays and weekends across a couple of typical years (2015 and 2016) for activity at the airport I'm looking at.

The weekday data:

```
list.files()
```

```
## [1] "Airport-activity.Rmd"
## [2] "All runway data 20190818.csv"
## [3] "aviation-activity.Rproj"
## [4] "EWR data.R"
## [5] "EWR_FD_150914_150916.csv"
## [6] "EWR_FD_160914_160916.csv"
## [7] "EWR_FD_2015_WKND.csv"
## [8] "EWR_FD_2016_WKND.csv"
## [9] "EWR_FD_20160917-18_20150919-20.xlsx"
## [10] "README.md"
## [11] "Save weekday 20190818.csv"
## [12] "Save weekend 20190818.csv"
## [13] "Time of day.csv"

a1 = read.csv("./EWR_FD_150914_150916.csv", skip = 7)
b1 = read.csv("./EWR_FD_160914_160916.csv", skip = 7)
ab1 = rbind(a1,b1)
      names(ab1) = tolower(names(ab1))
ab1$runway.occupancy.time=NULL
ab1$taxiway.used.to.enter.exit.runway=NULL
ab1$daytype = "weekday"
rm(a1,b1)
```

Locked and loaded.

The weekend data:

```
a = read.csv("./EWR_FD_2016_WKND.csv") #read.csv("./EWR_FD_150914_150916.csv", skip = 7)
head(a)
```

```
##      Carrier.Group Call.Sign Registration Model Operation
## 1  United Airlines   UAL1960      N34131 B752  Arrival
## 2  United Airlines   UAL213      N587UA B752  Arrival
## 3 General Aviation   AAL1384      N845NN B738  Arrival
## 4  United Airlines   RPA3540      N644RW E170  Arrival
## 5  United Airlines   UAL1261      N14228 B738  Arrival
## 6  United Airlines   UAL1186      N411UA A320  Arrival
##      Origination.Airport Destination.Airport Gate.Assigned..Aerobahn.
## 1                LAX                EWR                Gate_110
## 2                SFO                EWR                Gate_104
## 3                CLT                EWR                Gate_37
## 4                MEM                EWR                Gate_87
## 5                ORD                EWR                Gate_134
## 6                LAS                EWR                Gate_132
##      Gate Runway.Assigned..Aerobahn. Runway First.Fix
## 1 Gate_110                22L      22L      DAG
## 2 Gate_104                22L      22L      TIPRE
## 3 Gate_37                 22L      22L      AUDII
## 4 Gate_87                 22L      22L      DIYAB
## 5 Gate_134                22L      22L      DUFEE
## 6 Gate_132                22L      22L      DVC
##      International.or.Domestic.Indicator Event.Time Flight.Origination.Date
## 1                Domestic 9/17/16 0:02
## 2                Domestic 9/17/16 0:05
## 3                Domestic 9/17/16 0:08
## 4                Domestic 9/17/16 0:11
```

```
## 5 Domestic 9/17/16 0:12
## 6 Domestic 9/17/16 0:25
## Scheduled.Off.Block.Time..Aerobahn. Actual.Off.Block.Time..Aerobahn.
## 1 9/16/16 19:06 9/16/16 19:11
## 2 9/16/16 19:00 9/16/16 18:55
## 3 9/16/16 22:15 9/16/16 22:15
## 4 9/16/16 20:46 9/16/16 21:55
## 5 9/16/16 22:10 9/16/16 22:30
## 6 9/16/16 19:59 9/16/16 19:59
## Movement.Area.Entrance.Time Actual.Take.Off.Time..Aerobahn.
## 1 9/16/16 19:23
## 2 9/16/16 19:11
## 3 9/16/16 22:48
## 4 9/16/16 22:04
## 5 9/16/16 22:41
## 6 9/16/16 20:15
## Actual.Landing.Time..Aerobahn. Movement.Area.Exit.Time
## 1 9/17/16 0:02 9/17/16 0:06
## 2 9/17/16 0:05 9/17/16 0:12
## 3 9/17/16 0:08 9/17/16 0:11
## 4 9/17/16 0:11 9/17/16 0:15
## 5 9/17/16 0:12 9/17/16 0:18
## 6 9/17/16 0:25 9/17/16 0:32
## Scheduled.In.Block.Time..Aerobahn. Actual.In.Block.Time..Aerobahn.
## 1 9/17/16 0:21 9/17/16 0:07
## 2 9/17/16 0:13 9/17/16 0:13
## 3 9/17/16 0:04 9/17/16 0:12
## 4 9/16/16 23:38 9/17/16 0:15
## 5 9/17/16 0:19 9/17/16 0:19
## 6 9/17/16 0:49 9/17/16 0:33
## Total.Taxi.Time
## 1 0:05:28
## 2 0:07:32
## 3 0:03:48
## 4 0:04:28
## 5 0:06:41
## 6 0:07:48
```

```
b = read.csv("./EWR_FD_2015_WKND.csv") #read.csv("./EWR_FD_160914_160916.csv", skip = 7)
ab = rbind(a,b)
names(ab) = tolower(names(ab))
ab$daytype = "weekend"
rm(a,b)
```

A little messy, so rename variables and then consolidate.

```
names(ab1) = names(ab)
ab = rbind(ab1,ab)
rm(ab1)
```

Preparation.

The airport tracks all kind of activity. But I'm interested explicitly in arrivals and departures.

```
table(ab$operation)
```

```
##
##           Arrival           Departure Missed Approach Rejected Take-off
##           5573             5551             154             8
```

```
ab = subset(ab, ab$operation=="Arrival" | ab$operation == "Departure")
```

The underlying .csvs and the imports didn't recognize time for what it was. (Times and dates usually require a little extra attention.)

```
ab$day.time = strptime(ab$event.time, "%m/%d/%y %H:%M", tz = "EST5EDT")
```

```
## Warning in strptime(ab$event.time, "%m/%d/%y %H:%M", tz = "EST5EDT"):
## unknown timezone 'zone/tz/2019b.1.0/zoneinfo/America/New_York'
```

```
ab$date = format(as.POSIXct(ab$day.time,format='%m/%d/%Y %H:%M:%S'),format='%m/%d/%Y')
```

```
## Warning in as.POSIXct.POSIXlt(ab$day.time, format = "%m/%d/%Y %H:%M:%S"):
## unknown timezone 'zone/tz/2019b.1.0/zoneinfo/America/New_York'
```

```
## Warning in as.POSIXlt.POSIXct(x, tz): unknown timezone 'zone/tz/2019b.1.0/
## zoneinfo/America/New_York'
```

```
ab$date = as.Date(ab$date, format="%m/%d/%Y")
```

```
ab$weekday = weekdays(ab$date)
```

Specifically, I want to be able to treat time down to the minute, so I'll just create a time variable that lets me do it easily.

```
ab$time = as.POSIXct(as.numeric(as.POSIXct(ab$day.time)) %% 86400, origin = "2000-01-01")
```

```
## Warning in as.POSIXct.POSIXlt(ab$day.time): unknown timezone 'zone/tz/
## 2019b.1.0/zoneinfo/America/New_York'
```

```
ab$time2 = as.numeric(as.POSIXct(ab$day.time)) %% 86400 # 60*60*24
```

```
## Warning in as.POSIXct.POSIXlt(ab$day.time): unknown timezone 'zone/tz/
## 2019b.1.0/zoneinfo/America/New_York'
```

```
benchpoints = c("2001-01-01 05:30:00", "2001-01-01 06:30:00", "2001-01-01 9:30:00", "2001-01-01 16:30:00")
benchpoints2 = as.numeric(as.POSIXct(benchpoints)) %% 86400
```

```
# Effective hours are everything but 11pm-6am. 86400 seconds in a day.
ab$timecat = ifelse(ab$time2>=82800 | ab$time2<=21600,"Night","SixAMtoElevenPM") #18000
ab = droplevels(ab) #Quick cleanup to trim unused data levels and
```

OK, the analysis is going to consider expected delays by time of day. There are some rule-of-thumb level windows of time that the airport uses as categories for high-level analysis. I'll borrow them and shape the data into tables. I'll treat 6 AM to 11 PM as peak and the remaining hours as "night".

```
table(ab$timecat)
```

```
##
##           Night SixAMtoElevenPM
##           2784             8340
```

```
table(ab$timecat,ab$operation)
```

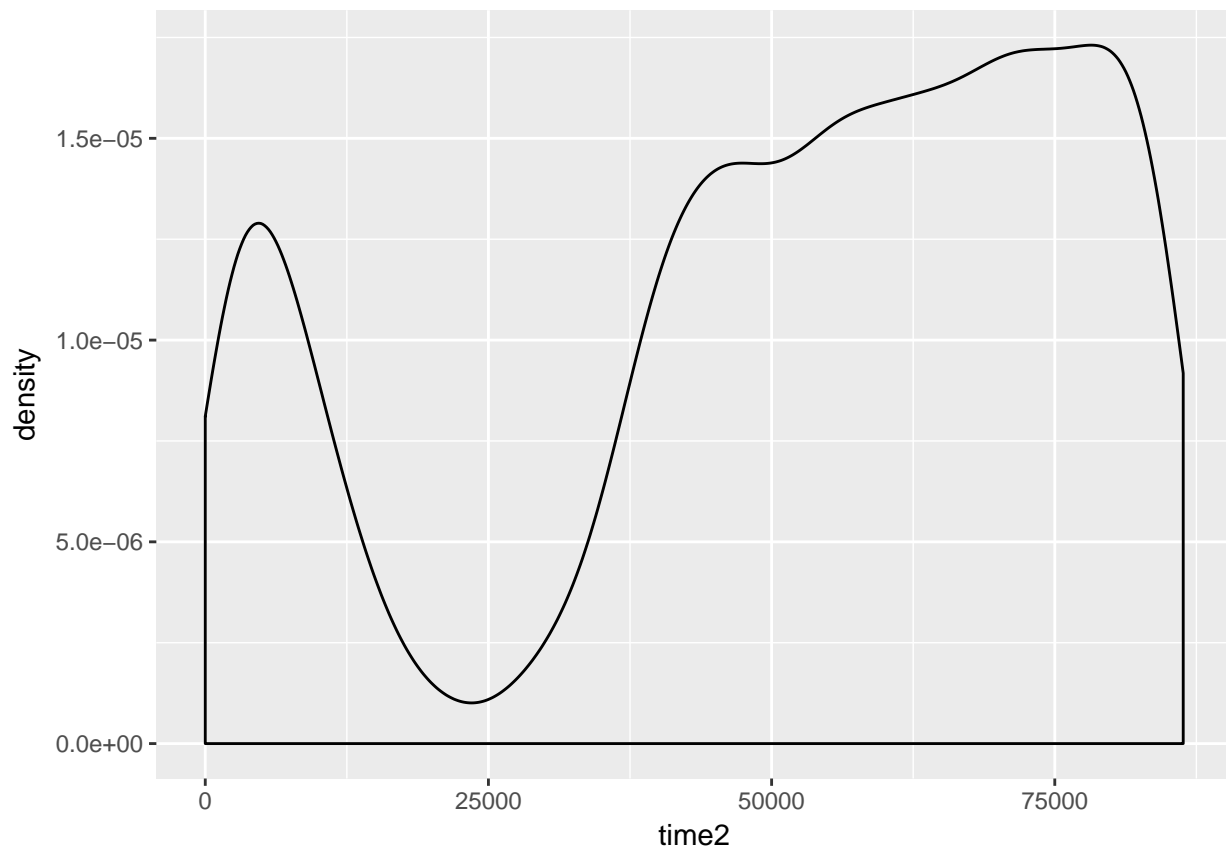
```
##
```

```
##           Arrival Departure
##   Night           1437     1347
##   SixAMtoElevenPM  4136     4204
```

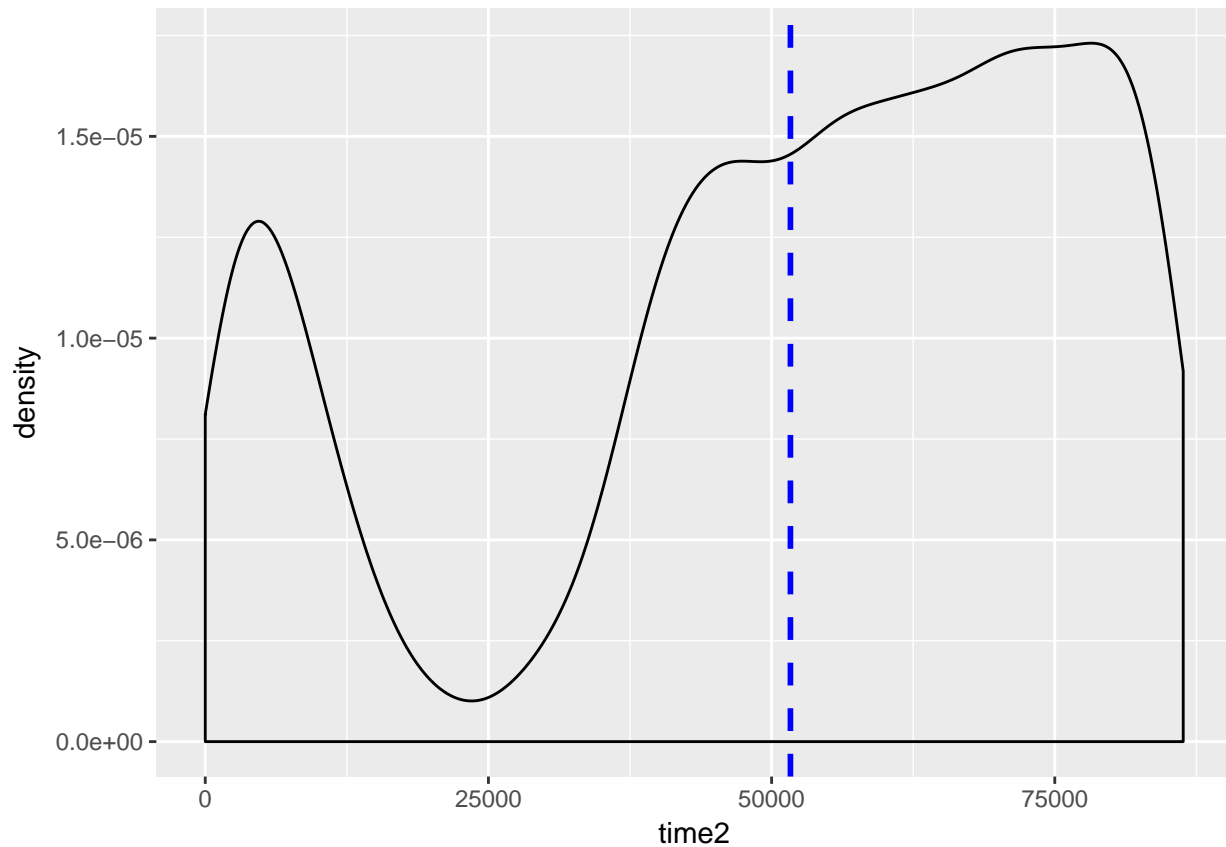
```
ab_wknd = subset(ab, ab$daytype=="weekend")
ab_wkdy = subset(ab, ab$daytype=="weekday")
ab_wknd = table(ab_wknd$timecat, ab_wknd$runway)
ab_wkdy = table(ab_wkdy$timecat, ab_wkdy$runway)
```

So what DOES activity at the airport look like, anyway?

```
# Basic density
p = ggplot(ab, aes(x=time2)) +
  geom_density()
p
```

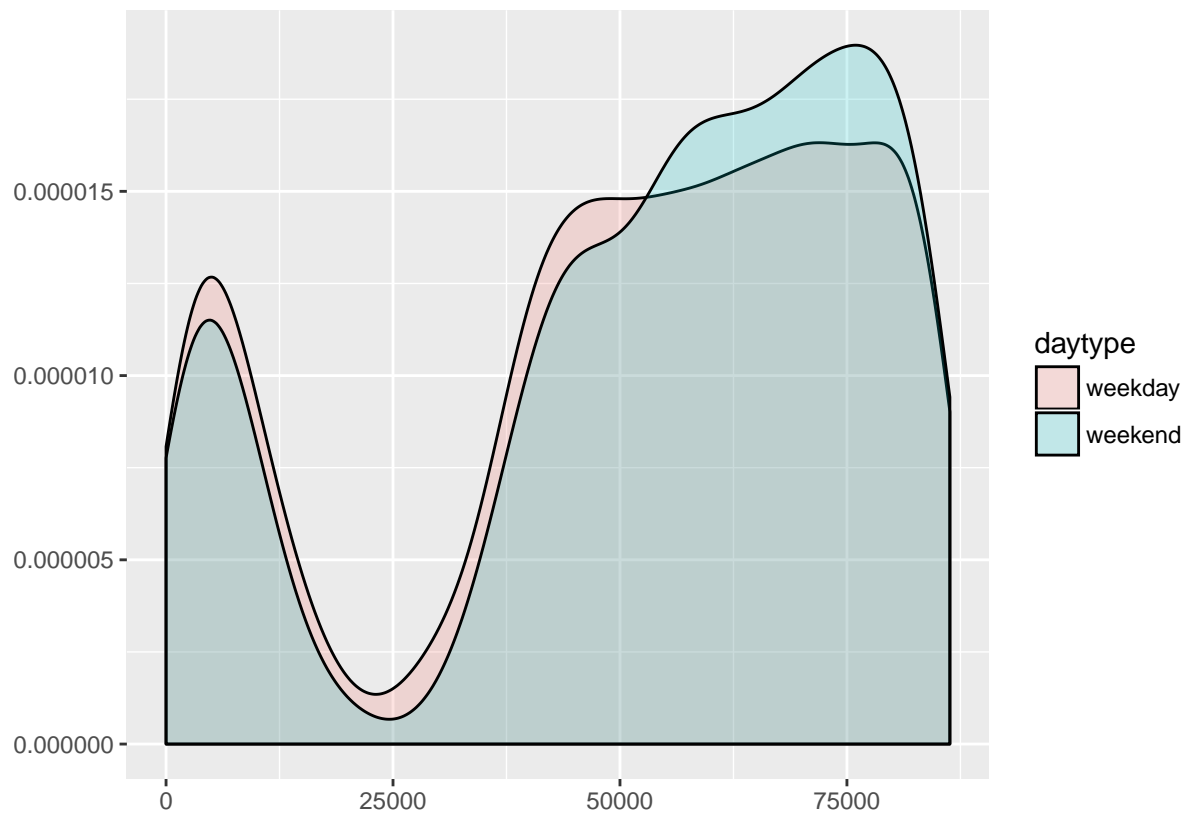


```
# Add mean line
p + geom_vline(aes(xintercept=mean(time2)),
  color="blue", linetype="dashed", size=1)
```



These will be the tables I use to apply expected delays, which we will estimate separately using the same data but during periods of similar delays in other years. First I'm curious as to how different the airport's weekday profile is from the weekend:

```
options(scipen=5)
pq = ggplot(ab, aes(x=time2, fill=daytype)) +
  geom_density(alpha=0.2) + labs(x = "", y = "")
pq
```



```
ggsave(pq, file="./Activity_EWR.png", height=4,width=6)
```

The distributions are fairly similar.

Saving the data:

```
write.csv(ab_wknd,"./Save weekend 20190818.csv")  
write.csv(ab_wkdy,"./Save weekday 20190818.csv")  
write.csv(ab,"./All runway data 20190818.csv")
```