



# Taking Google Analytics to the next level

Johann de Boer, digital analytics consultant and trainer

# Johann de Boer

- Digital Analytics consultant and trainer
- Loves Data
- Master of Analytics (Marketing)
- Bachelor of Engineering (Computer Systems)
- From user experience and accessibility
- To analytics implementation and insights



# Purpose of this talk

Encourage web data analysts to move to R and away from Excel!

Help those wanting to learn R to get started.

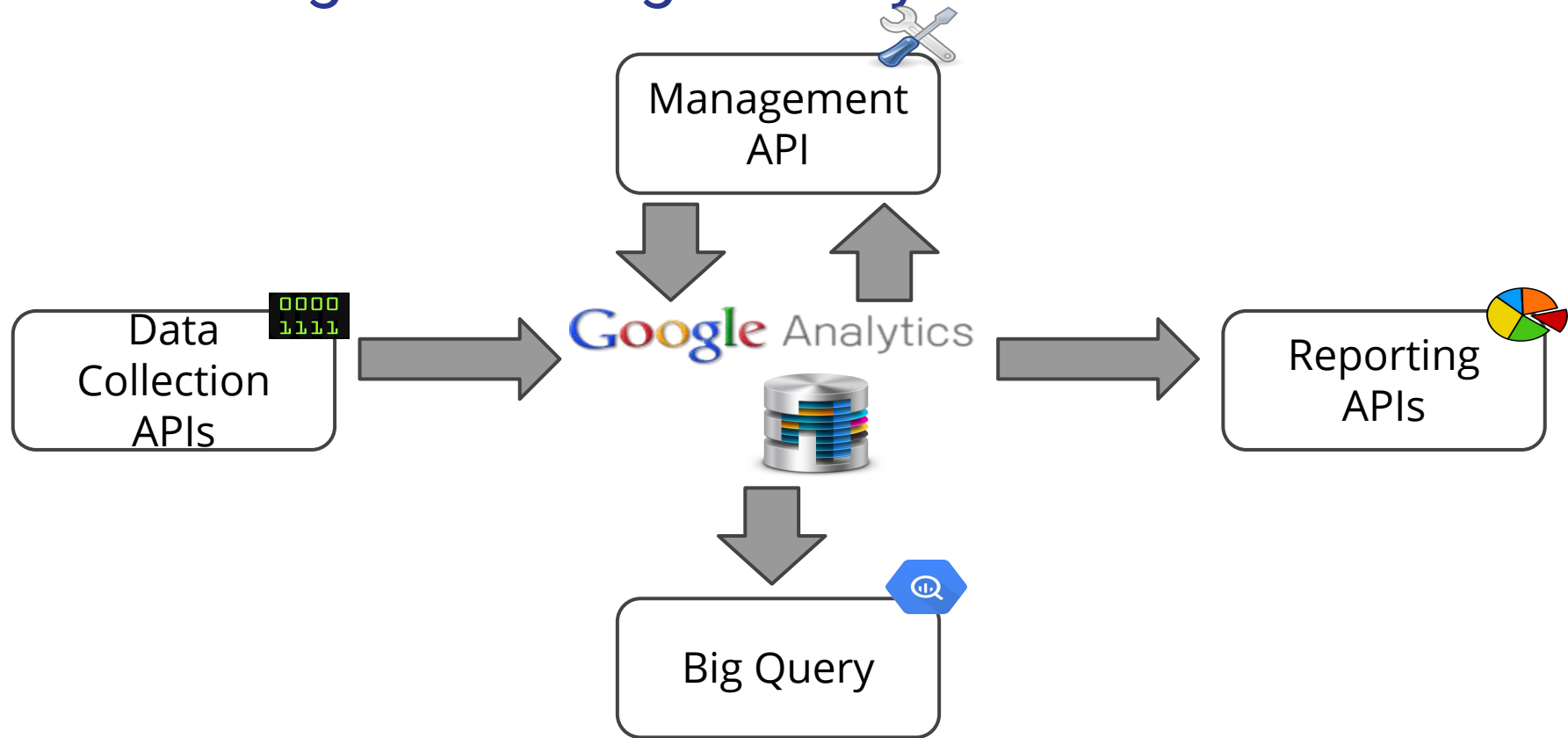


# What is R?

- open-source programming language
- purpose built for data analysis
- World domination!



# Interfacing with Google Analytics



# GA API platform overview

- **Collection APIs**
  - Collect and send data to Google Analytics for storage and processing
- **Management API**
  - Configure Google Analytics accounts
  - Import off-line data
- **Reporting APIs**
  - Query and extract processed data from Google Analytics for reporting

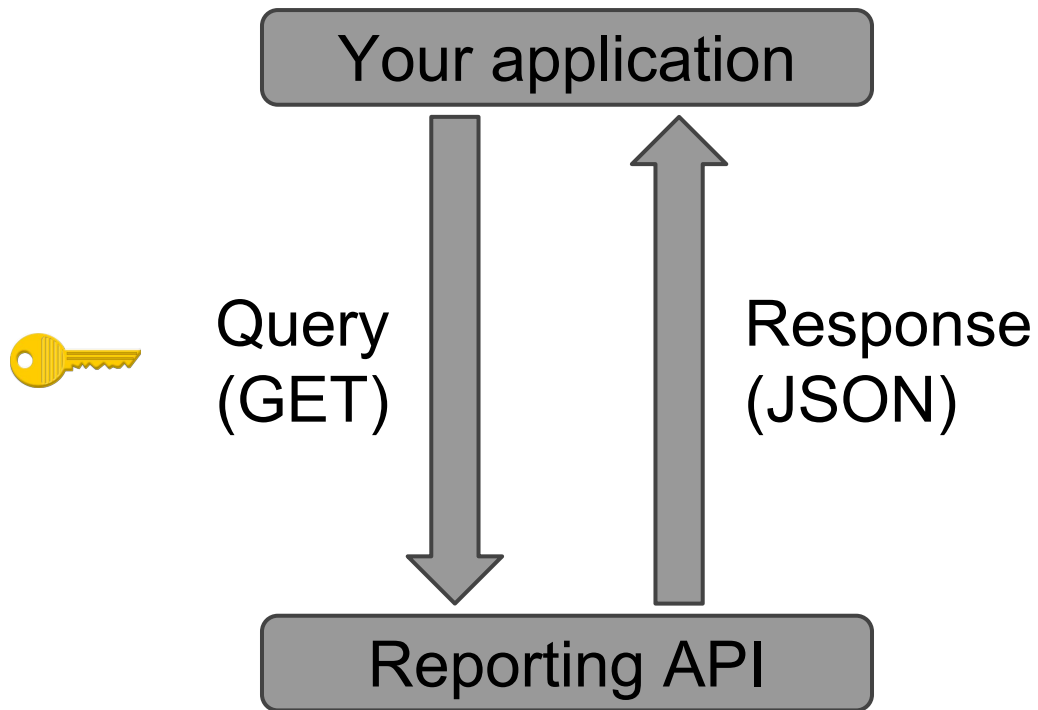


# Why use the Reporting APIs?

- Surpass the limits of the GA user interface
- Automate repetitive tasks
- Perform more powerful data analysis
- Easily reproduce and reuse analyses
- Combine with other data sources
- Develop innovative tools
- Feed into other tools and applications



# API concepts

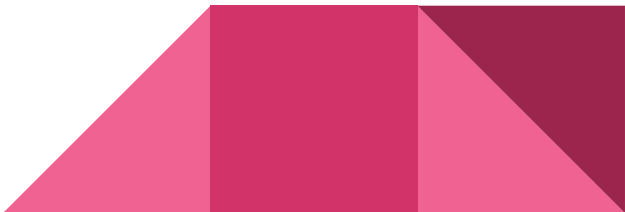




# API Query

Authorization: Bearer {oauth2-token}

GET https://www.googleapis.com/analytics/v3/data/ga?  
ids=ga:98765432&  
start-date=2014-01-01&  
end-date=2014-01-31&  
metrics=ga:sessions&  
dimensions=ga:landingPagePath&  
sort=-ga:sessions&  
filters=ga:medium==organic&  
segment=users::condition::ga:goalCompletionsAll>0&  
samplingLevel=HIGHER\_PRECISION&  
start-index=1&  
max-results=10000



# API Response

```
{
  "kind": "analytics#gaData",
  "id": string,
  "selfLink": string,
  "containsSampledData": boolean,
  "query": {
    ...
  },
  "itemsPerPage": integer,
  "totalResults": integer,
  "previousLink": string,
  "nextLink": string,
  "profileInfo": {
    ...
  },
```

```
  "columnHeaders": [
    {
      "name": string,
      "columnType": string,
      "dataType": string
    }, ...
  ],
  "rows": [
    [string, ...], ...
  ],
  "sampleSize": string,
  "sampleSpace": string,
  "totalsForAllResults": [{
    metricName: string,
    ...
  }]
}
```

# Query Explorer

* ids	=	ga:98765432
dimensions	=	ga:hour,ga:dayOfWeekName
* metrics	=	ga:sessions
segment	=	
filters	=	
sort	=	
* start-date	=	2014-08-03
* end-date	=	2014-08-09
start-index	=	
max-results	=	10000

Get Data

Your query matched 168 results and the API only returned the

ga:hour	ga:dayOfWeekName	ga:sessions
00	Friday	2143
00	Monday	615
00	Saturday	1884
00	Sunday	540
00	Thursday	2283
00	Tuesday	2115
00	Wednesday	2284
01	Friday	2343
01	Monday	585
01	Saturday	1931
01	Sunday	521
01	Thursday	2237
01	Tuesday	2209
01	Wednesday	2257

# Expressing yourself with GA

Explaining the steps you followed to put together a data analysis report:

- precisely and accurately
- using natural language but without ambiguity
- so that others can reproduce that report without error, and
- can reuse your steps to save time in pursuing other analysis ideas



# Google Analytics Reporting Query

1. **Account Property View**
2. **Date Range: Start Date** and **End Date**
3. **Variables** to form rows and columns: **Dimensions** and **Metrics**
4. **Maximum Rows** and which variables to **Sort By**
5. **Table Filters** to remove unwanted rows
6. One or more **Segments** to compare





## New Custom Report

Some data in this report may have been removed when a threshold was applied. [Learn more](#)

Edit Email Export Add to Dashboard Shortcut

Segmentation



All Sessions  
100.00%

+ Add Segment

Date Range

27 Oct 2015 - 26 Nov 2015

This report is based on 24

Control the number of sessions used to calculate this report:

Sampling Level

Faster Processing

Higher Precision

[Learn More](#)

Table Filter

Report Tab

 advanced

Gender ?	Device Category ?	Default Channel Grouping	Users ↓	Sessions	Page Views
1. female	desktop	Organic Search	98,032 (23.01%)	147,463 (22.24%)	917,940 (26.17%)
2. male	desktop	Organic Search	76,848 (18.04%)	114,314 (17.24%)	736,503 (21.00%)
3. female	tablet	Organic Search	38,171 (8.96%)	70,718 (10.66%)	317,577 (9.06%)
4. female	mobile	Organic Search	29,989 (7.04%)	42,513 (6.41%)	156,637 (4.47%)
5. male	tablet	Organic Search	17,440 (4.09%)	29,298 (4.42%)	132,240 (3.77%)
6. male	desktop	Direct	14,483 (3.40%)	23,971 (3.61%)	144,344 (4.12%)
7. female	desktop	Direct	14,160 (3.32%)	22,332 (3.37%)	141,068 (4.02%)
8. male	mobile	Organic Search	13,065 (3.07%)	17,301 (2.61%)	62,469 (1.78%)
9. female	desktop	Referral	9,322 (2.19%)	15,416 (2.32%)	103,243 (2.94%)
10. female	mobile	Direct	9,138 (2.15%)	12,526 (1.89%)	60,566 (1.73%)

Dimensions

Metrics

Maximum Rows

Show rows: 10

Go to: 1

1-10 of 37

GA Custom Report		
File Edit View Insert Format Data Tools Add-ons Help All ch		
£ % .00 123 Arial 10 B		
ga:users		
	A	B
1	Configuration Options	Your Google Analytics Reports
2	Report Name	Report1
3	Type	core
4	View (Profile) ID / ids	ga:67809522
5	Start Date	27/10/2015
6	End Date	26/11/2015
7	Last N Days	30
8	Metrics	ga:users ga:sessions ga:pageviews
9	Dimensions	ga:deviceCategory ga:userGender ga:channelGrouping
10	Sort	
11	Filters	
12	Segment	
13	Sampling Level	
14	Start Index	
15	Max Results	
16	Spreadsheet URL	



GA Custom Report

File Edit View Insert Format Data Tools Add-ons Help All changes saved in Drive

Comments Share

Report1

	A	B	C	D	E	F
1	Report1					
2	Last Run On	27/11/2015 11:27:47				
3	View (Profile) Name	villagecinemas.com.au [PRIMARY]				
4	Total Results Found	37				
5	Total Results Returned	37				
6	Contains Sampled Data	No				
7						
8						
9						
10	Totals For All Results					
11				ga:users	ga:sessions	ga:pageviews
12				125780	166590	901075
13						
14	Results Breakdown					
15	ga:deviceCategory	ga:userGender	ga:channelGrouping	ga:users	ga:sessions	ga:pageviews
16	desktop	female	(Other)	10	16	64
17	desktop	female	Direct	4277	5567	35742
18	desktop	female	Email	2215	2924	12423
19	desktop	female	Organic Search	30482	38715	245775
20	desktop	female	Paid Search	2302	2688	14160
21	desktop	female	Referral	2706	3688	26030
22	desktop	female	Social	74	96	578
23	desktop	male	(Other)	13	25	170
24	desktop	male	Direct	4262	5913	35782

Report Configuration Report1

```

library(ganalytics)
my_creds <- GoogleApiCreds("johann@lovesdata.com")

q <- GaQuery(view = 67809522, creds = my_creds)

DateRange(q) <- c("2015-10-27", "2015-11-26")
Dimensions(q) <- c("userGender", "deviceCategory", "channelGroup")
Metrics(q) <- c("users", "sessions", "pageviews")

GetGaData(q)

```

	userGender	deviceCategory	channelGrouping	users	sessions	pageviews
1	female	desktop	(Other)	61	195	924
2	female	desktop	Direct	14469	22653	142853
3	female	desktop	Email	8099	14754	59855
4	female	desktop	Organic Search	98360	148007	923598
5	female	desktop	Paid Search	8397	10551	54822
6	female	desktop	Referral	9125	14864	99001
7	female	desktop	Social	421	637	3347
8	female	mobile	(Other)	43	49	141
9	female	mobile	Direct	8821	12078	58734
10	female	mobile	Email	7375	12344	29087



# Conditions for Table Filters and Segments

Conditions are expressed as:

~ **<variable>** **<comparator>** **<operand>**

Examples:

~ **UserGender** **==** **'female'**

~ **Bounces** **>** **0**

~ **Hour** **in** **0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11**



# Using conditions to form expressions

Combine one or more conditions with **AND, OR**

Define **User** and **Session** level segments

Define **Sequences** used for segmentation

**First, Then, Later**

**Scope** conditions to **hit, session** or **user** level



# Sequence segments

```
played_video <-  
  Expr(~EventCategory == "Video") &  
  Expr(~EventAction == "Play")  
  
purchased_ticket <-  
  PerHit(  
    Expr(~goal1completions > 0)  
  )  
  
journey <- PerUser(Sequence(  
  played_video,  
  Then(purchased_ticket)  
))
```

The screenshot shows a sequence builder interface with the following components:

- Filter:** Includes dropdowns for "Include", "Users", "Sequence start", and "Any user interaction".
- STEP 1:**
  - Event Category: exactly matches Video
  - AND
  - Event Action: exactly matches Play
  - is followed by...
- STEP 2:**
  - [Macro] Buy Movie Tickets (General) (Goal 1 Completions)
  - per hit
  - >
  - 0
- Add Step:** A button at the bottom to add a new step.

```
users::sequence::ga:eventCategory==Video;ga:  
eventAction==Play;->perHit::ga:  
goal1Completions>0
```

ganalytics for R

# Goals of the 'ganalytics' R package

- Access to all Google Analytics features
- Easy to use, familiar to R users and in GA lingo
- Abstract away from API technicalities
- Interactively build and manipulate queries
- Detect and correct query definition errors
- Integrate with R classes, functions and other packages



# Getting started

Google Developers Console

API Project ▼

[Sign up for a free trial.](#)

Overview

Permissions

**APIs & auth**

APIs

[Credentials](#)

Consent screen

Push

Monitoring

Source Code

Compute

## OAuth

OAuth 2.0 allows users to share specific data with you (for example, contact lists) while keeping their usernames, passwords, and other information private.

[Learn more](#)

[Create new Client ID](#)

## Client ID for native application

Client ID	[REDACTED]
Client secret	[REDACTED]
Redirect URIs	urn:ietf:wg:oauth:2.0:oob http://localhost

[Reset secret](#) [Download JSON](#) [Delete](#)

# Installation and “Hello World!”

```
devtools::install_github("jdeboer/ganalytics")
```

```
library(ganalytics)
```

```
creds <- GoogleApiCreds(  
  "you@domain.com", # Your Google username  
  "client_secret.json" # From Google APIs console  
)
```

```
# Default view of first account and property  
view <- NA
```

```
query <- GaQuery(view, creds)
```

```
GetGaData(query)
```



# “Hello World!” of ganalytics

```
|=====| 100%  
Completed after 1 s  
  date sessions  
1 2015-05-03    3566  
2 2015-05-04    2378  
3 2015-05-05    2028  
4 2015-05-06    1970  
5 2015-05-07    1962  
6 2015-05-08    1727  
7 2015-05-09    1381  
8 2015-05-10    1510  
> |
```



Defining your query and  
getting data into R with  
ganalytics

## Example query

```
query <- GaQuery(  
  view = 94605596,  
  startDate = "2015-01-01", endDate = "2015-01-31",  
  metrics = c("users", "sessions", "pageviews"),  
  dimensions = c("deviceCategory", "dayOfWeekName"),  
  filter = Expr(~deviceCategory %matches% "mobile|tablet"),  
  segment = Expr(~country == "Australia")  
)
```

```
response <- GetGaData(query)
```



# Example query response

	deviceCategory	dayOfWeekName	users	sessions	pageviews
1	mobile	Friday	429	500	816
2	mobile	Monday	273	329	466
3	mobile	Saturday	339	399	640
4	mobile	Sunday	256	301	487
5	mobile	Thursday	420	475	689
6	mobile	Tuesday	314	358	542
7	mobile	Wednesday	304	357	571
8	tablet	Friday	209	235	424
9	tablet	Monday	160	190	355
10	tablet	Saturday	195	225	435
11	tablet	Sunday	128	151	334
12	tablet	Thursday	217	266	455
13	tablet	Tuesday	169	194	374
14	tablet	Wednesday	161	178	356

# Conditional expressions

```
expr1 <- Expr(~variable comparator value)  
expr2...
```

```
expr3 <- expr1 | !expr2
```

```
expr5 <- expr3 & expr4
```

```
expr6 <- xor(expr1, expr2)
```



# From R to an API call

```
e1 <- Expr(~keyword %starts_with% 'buy')  
e2 <- Expr(~city %matches% '^(Sydney|Melbourne)$')  
e3 <- Expr(~deviceCategory == 'tablet')
```

```
e4 <- e1 & (e2 | e3)
```

```
as(e4, 'character')
```

```
# ga:keyword=@buy;ga:city=~^(sydney|melbourne)$,ga:deviceCategory==tablet
```



How does traffic from desktop, mobile and tablet users change throughout the day and over the week?

Average number of visits per hour and day – split by desktop, mobile and tablet

## Step 1: Define the query

```
query <- GaQuery(view = '98765432')
```

```
DateRange(query) <- c("2014-01-01", "2014-12-31")
```

```
Dimensions(query) <- c(  
  "deviceCategory", "dayOfWeekName", "hour", "date"  
)
```

```
Metrics(query) <- "sessions"
```

```
MaxResults(query) <- 30000
```



## Step 2: Get data and summarise

```
data <- GetGaData(query)
```

```
library(dplyr)
```

```
weekly_data <- tbl_df(data) %>%  
  group_by(deviceCategory, dayOfWeekName, hour) %>%  
  summarise(avg_sessions_per_day = mean(sessions))
```

#	deviceCategory	dayOfWeekName	hour	avg_sessions_per_day
#1	desktop	Friday	0	83.83673
#2	desktop	Friday	1	81.79167
#3	desktop	Friday	2	77.29167
#4	desktop	Friday	3	80.35417
#5	desktop	Friday	4	91.60417
#..	...	...	...	...



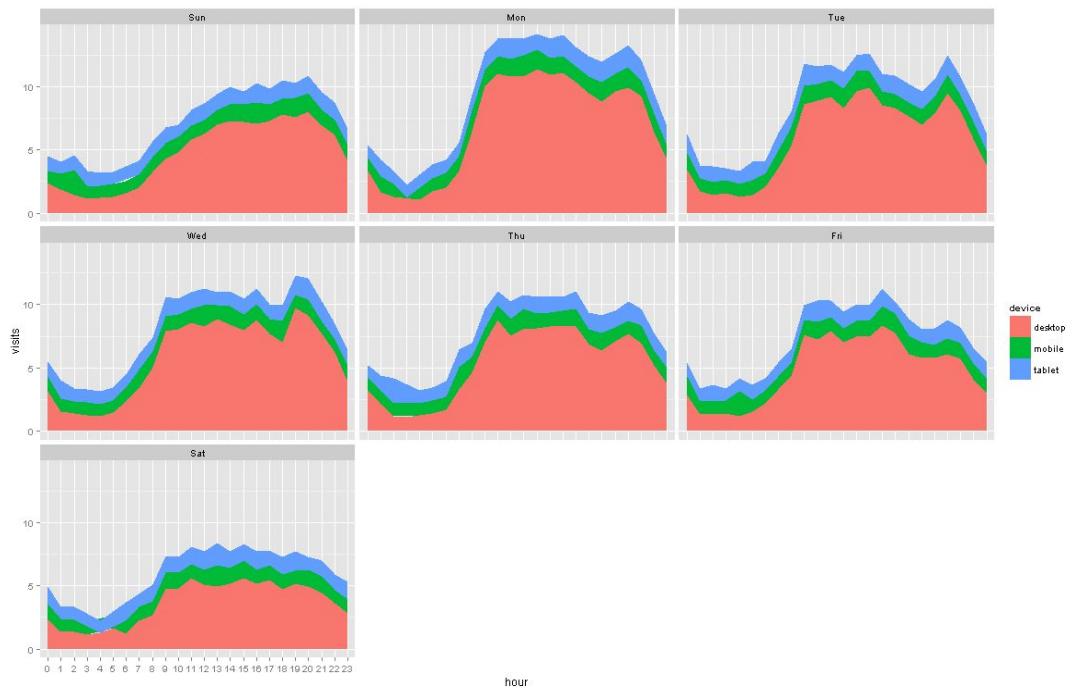
## Step 3: Plot the summary

```
library(ggplot2)

ggplot(weekly_data) +
  aes(
    x = hour,
    y = avg_sessions_per_day,
    fill = deviceCategory,
    group = deviceCategory
  ) +
  geom_area(position = "stack") +
  facet_wrap(~dayOfWeekName)
```



# ggplot2 + dplyr + ganalytics =



Which permutation of two different device types are signed-in users most likely to transition between?

Define sequences to segment users by permutations of device type and compare against total users of each type of device.

## Step 1: Setup query (needs UserID)

```
Dimensions(query) <- NULL
```

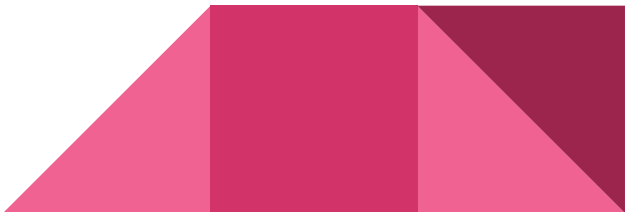
```
Metrics(query) <- c("users")
```

```
DateRange(query) <- c("2015-01-01", "2015-03-31")  
# Maximum of 90 days for user-based segmentation
```



## Step 2: Define sequence segments


```
devices <- list(  
  desktop = Expr(~deviceCategory == "desktop"),  
  mobile = Expr(~deviceCategory == "mobile"),  
  tablet = Expr(~deviceCategory == "tablet")  
)  
  
device_sequences <- lapply(devices, function(from) {  
  lapply(devices, function(to) {  
    SegmentFilters(  
      Sequence(First(from), Then(to)),  
      scope = "users"  
    )  
  })  
})
```



## Step 3: Get data for each segment

```
data <- lapply(seq_along(device_sequences), function(from_index){  
  from_name <- names(device_sequences)[from_index]  
  from_seq <- device_sequences[[from_index]]  
  lapply(seq_along(from_seq), function(to_index){  
    to_name <- names(from_seq)[to_index]  
    Segment(query) <- from_seq[[to_index]]  
    df <- GetGaData(query)  
    df <- cbind(from = from_name, to = to_name, df)  
  })  
})
```

```
data <- unlist(data, recursive = FALSE)  
data <- do.call(rbind, data)
```



# Step 4: Summarise the data

```
library(dplyr)
```

```
benchmarks <- data %>%  
  subset(from == to) %>%  
  select(from, benchmark = users)
```

```
data <- data %>%  
  subset(from != to) %>%  
  inner_join(benchmarks, by = "from") %>%  
  group_by(from, to) %>%  
  summarise(transitioned = users / benchmark)
```

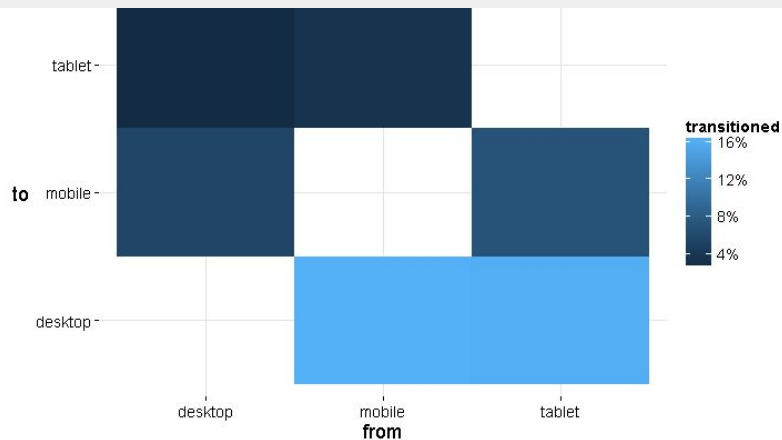
#	from	to	transitioned
# 1	desktop	mobile	0.05565673
# 2	desktop	tablet	0.02547988
# 3	mobile	desktop	0.16147748
# 4	mobile	tablet	0.03634899
# 5	tablet	desktop	0.15945559
# 6	tablet	mobile	0.07034457



# Step 5: Plot the results

```
library(ggplot2, scales)
```

```
ggplot(data) +  
  aes(x = from, y = to, fill = transitioned) +  
  geom_tile() +  
  scale_fill_continuous(labels = percent) +  
  theme_minimal(base_size = 18) +  
  theme(axis.title.y = element_text(angle = 0))
```



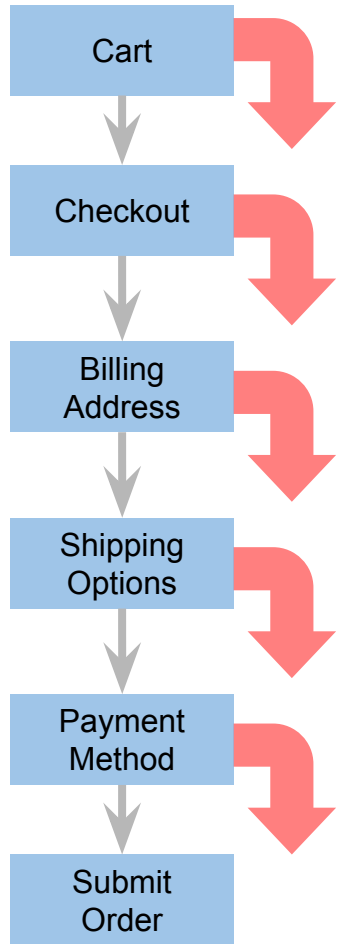


# Custom Funnels

Where are shoppers likely to drop out of your ecommerce website?

# Custom Funnels: GA Premium

- A linear sequence of stages like a goal funnel:
  - Restricted to only five stages
- Apply retrospectively
- Beyond just pageviews, e.g. campaigns, events
- Session-based like goals or user-based
- Like the enhanced ecommerce funnel reports:
  - Create segments on-the-go



# Custom funnels for everyone else

1. Create segments for:
  - a. each stage in the funnel
  - b. the sequence of steps required to reach that stage
2. Get the number of users for each segment
3. Calculate:
  - a. Entrances and exits for each stage
  - b. Abandonment and conversion rates

# Count traffic at each funnel stage

- Cart

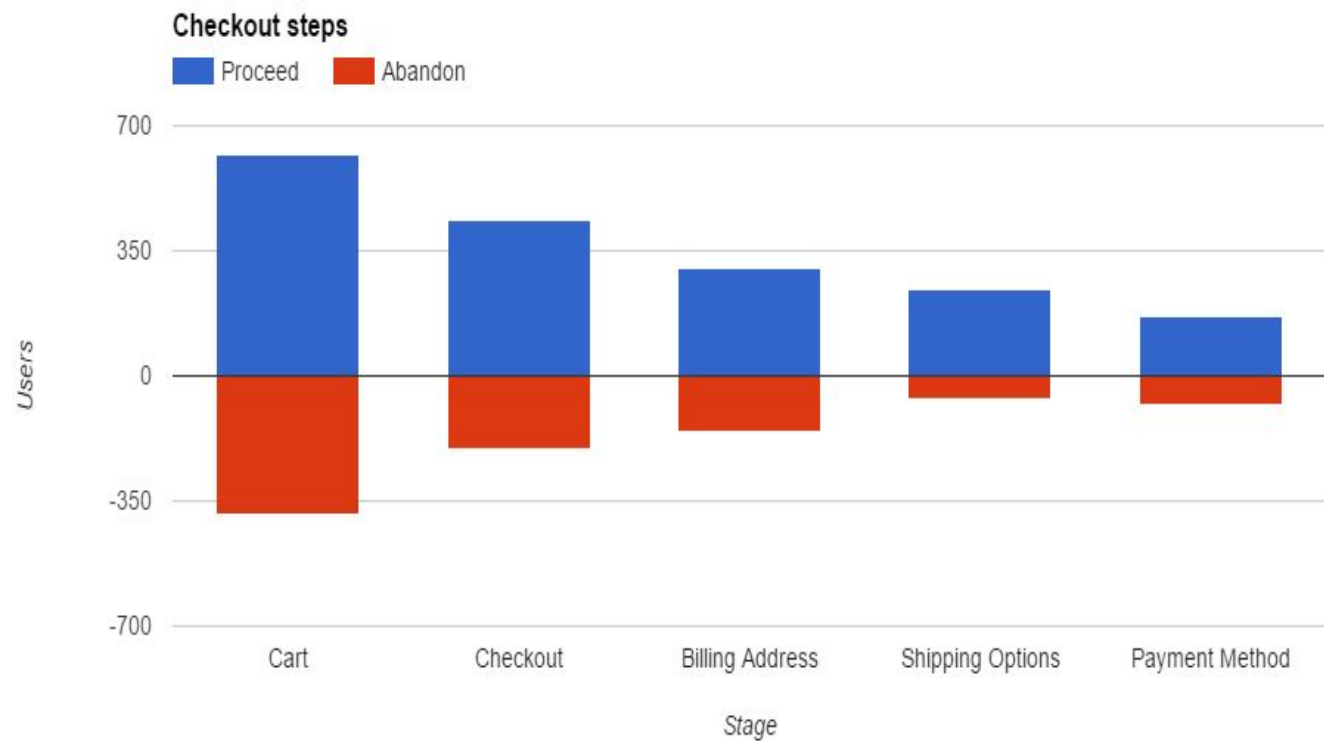
**users::condition::**pagePath=~cart\$

# Count traffic at each funnel stage

- Cart =  
`users::condition::pagePath=~cart$`
- Checkout =  
`users::condition::pagePath=~checkout$`
- Billing Address =  
`users::condition::pagePath=~billingaddress$`
- Shipping Options =  
`users::condition::pagePath=~shippingoptions$`
- Payment Method = `users::condition::`  
`pagePath=~paymentmethod$`
- Submit Order =  
`users::condition::pagePath=~submitorder$`

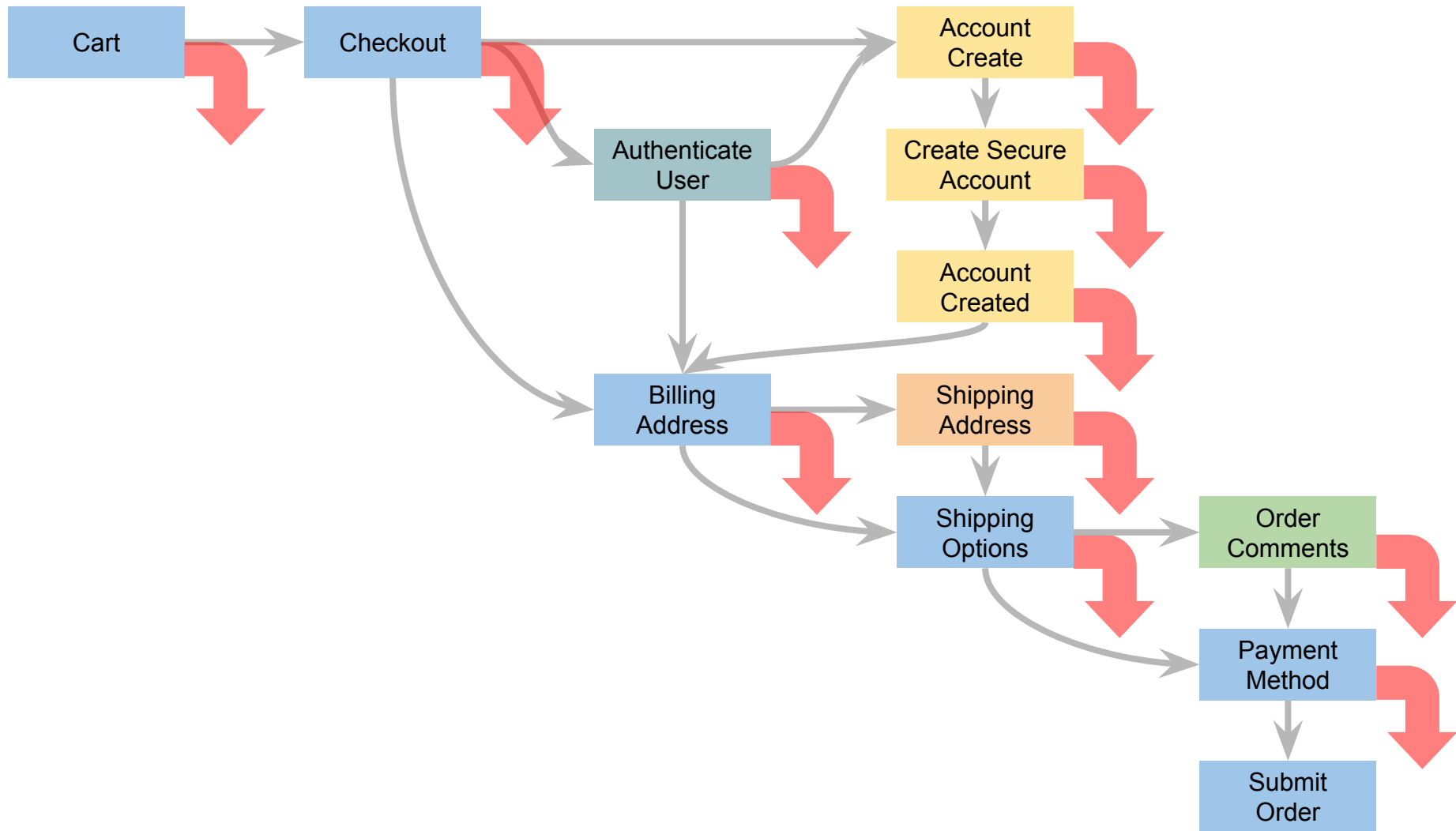
# Count path completions

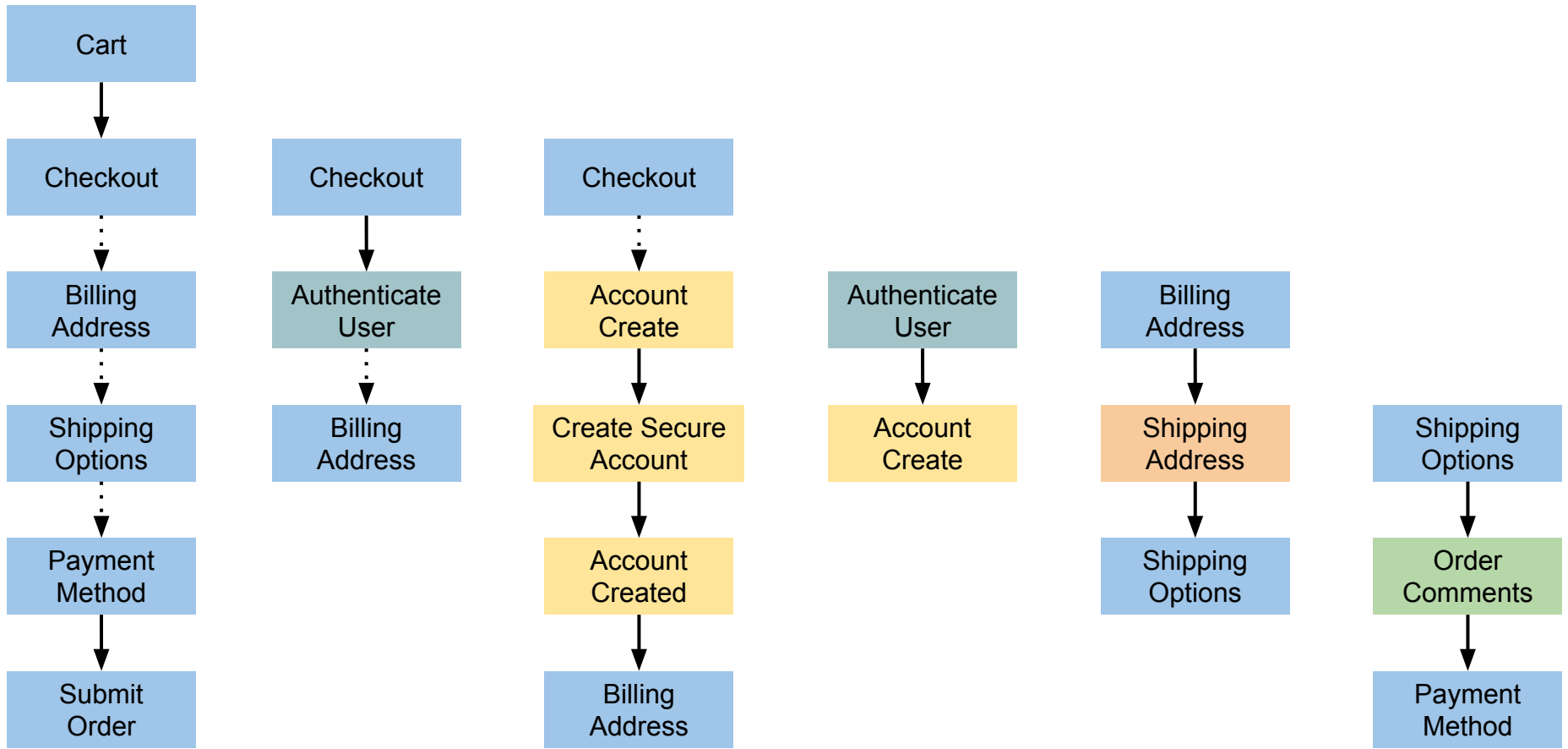
- `users::sequence::`
- `pagePath=~cart$`
- `; ->>pagePath=~checkout$`
- `; ->>pagePath=~billingaddress$`
- `; ->>pagePath=~shippingoptions$`
- `; ->>pagePath=~paymentmethod$`
- `; ->>pagePath=~submitorder$`



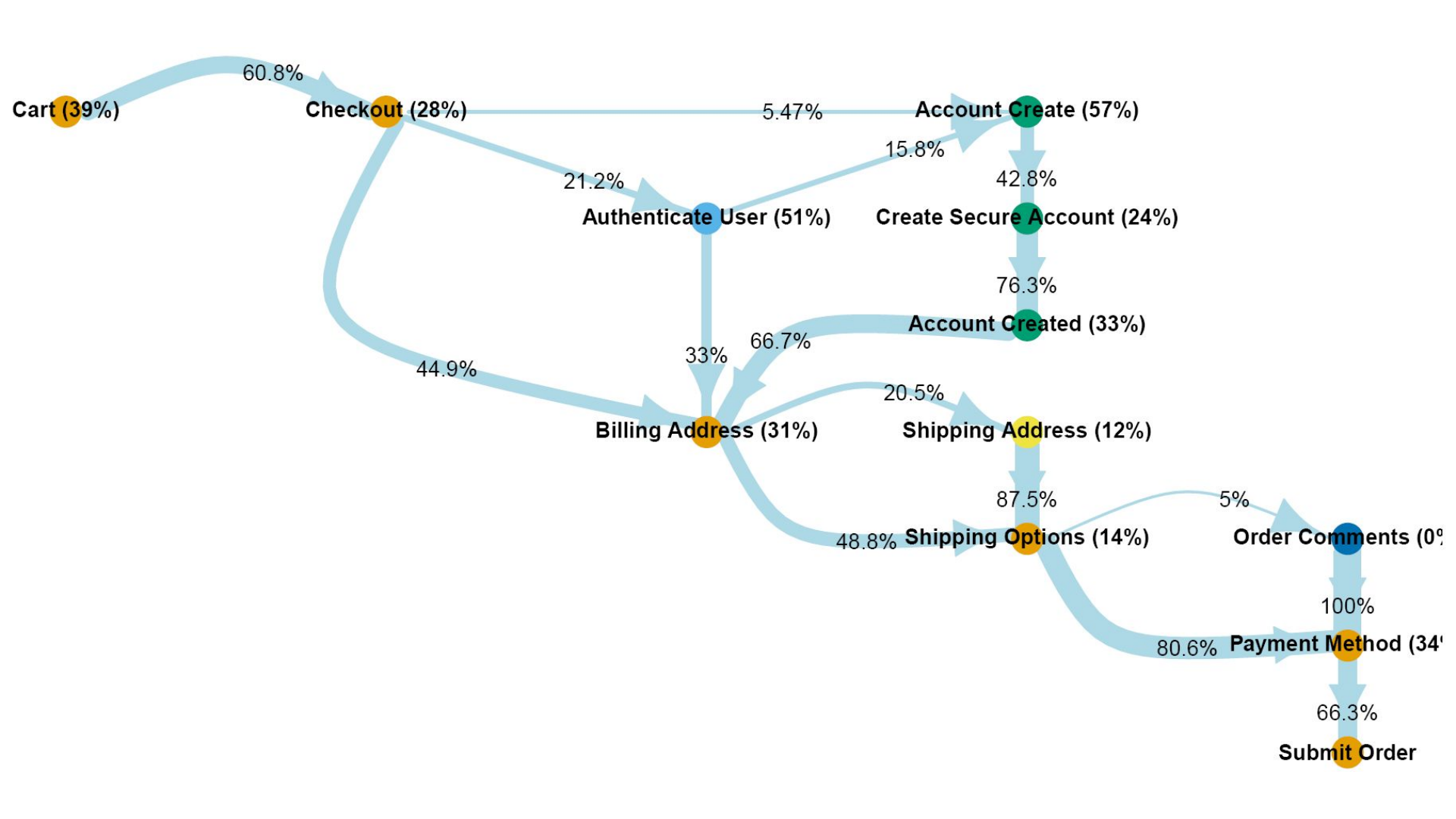
# Conversion funnel analysis with R







16 transitions (5 are shortcuts), and 16 routes from start to end.



# Using R

```
query <- GaQuery(view = 98765432)
Segment(query) <- Sequence(
  Cart,
  Checkout,
  BillingAddress,
  ShippingOptions,
  PaymentMethod,
  SubmitOrder
)
GetData(query)
```

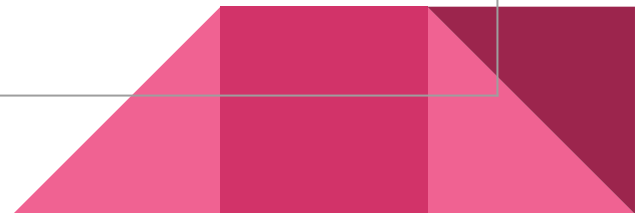
# Using analytics with other Google Analytics APIs

# Reporting APIs

GaQuery()	Core Reporting API	Standard reporting over a date range
McfQuery()	Multi-channel funnels reporting API	Requires conversion tracking to be set up (goals or ecommerce)
RtQuery()	Real-time reporting API	Request developer access first

# Other APIs

MetaData API		Internal package use only
Management API	GaAccounts() GaUserSegments()	Work in progress so please use with caution!
Google Tag Manager API	GtmAccounts()	





# Get involved!

Open source R package development is fun!



# Remaining work to complete

- Package documentation
- Adding examples and demos
- Further error handling and messages
- Testing and finding bugs to fix
- Submitting to CRAN
- Code optimisation and tidy up
- Feature enhancements and improvements
- Occasional updates for API changes
- Internationalisation



# Shiny demo

<https://mark.shinyapps.io/GA-dashboard-demo/>

# How to get involved ...

- Go to the Github site:  
[github.com/jdeboer/ganalytics](https://github.com/jdeboer/ganalytics)
- Submit feedback via the issues tracker
- Get in touch on twitter: @johannux

