



For Business

Solving a **\$15M/year** Employee Attrition Problem with the **tidyverse**
rstudio::conf(2019)



Matt Dancho

Business Science University
university.business-science.io



About Your Host

- Founder, Business Science
- Lover of
- Educator of Data Science



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Matt Dancho

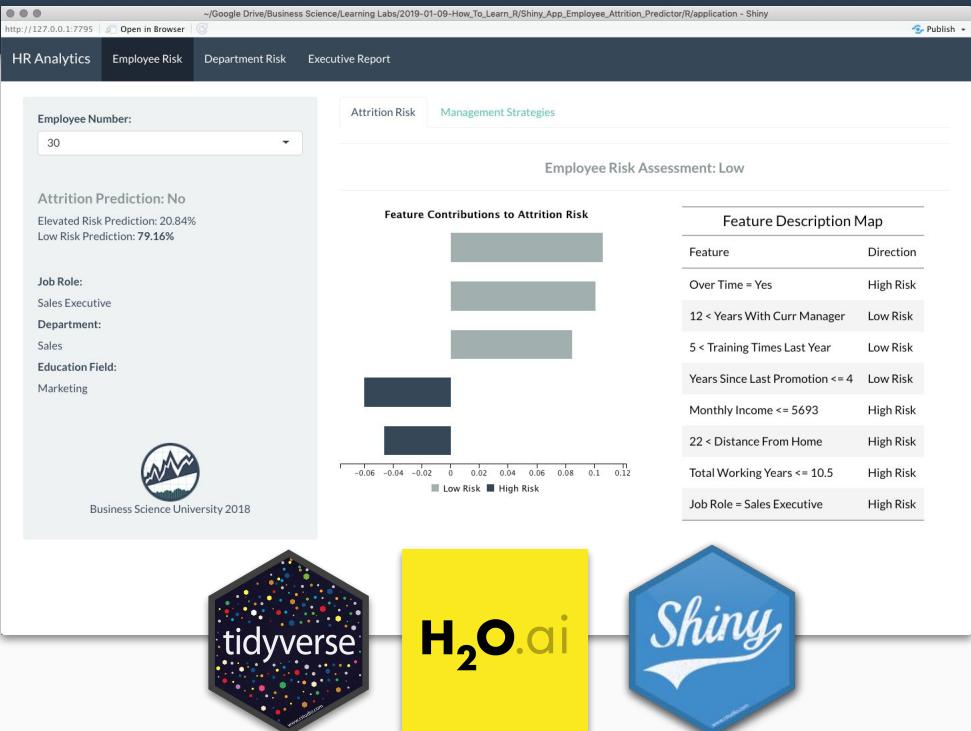
Founder of Business Science, Matt designs and executes educational courses and workshops that deliver immediate value to organizations. His passion is up-leveling future data scientists coming from untraditional backgrounds.

*“I like converting **Business People** to **Data Scientists**”*

Agenda: R For Business



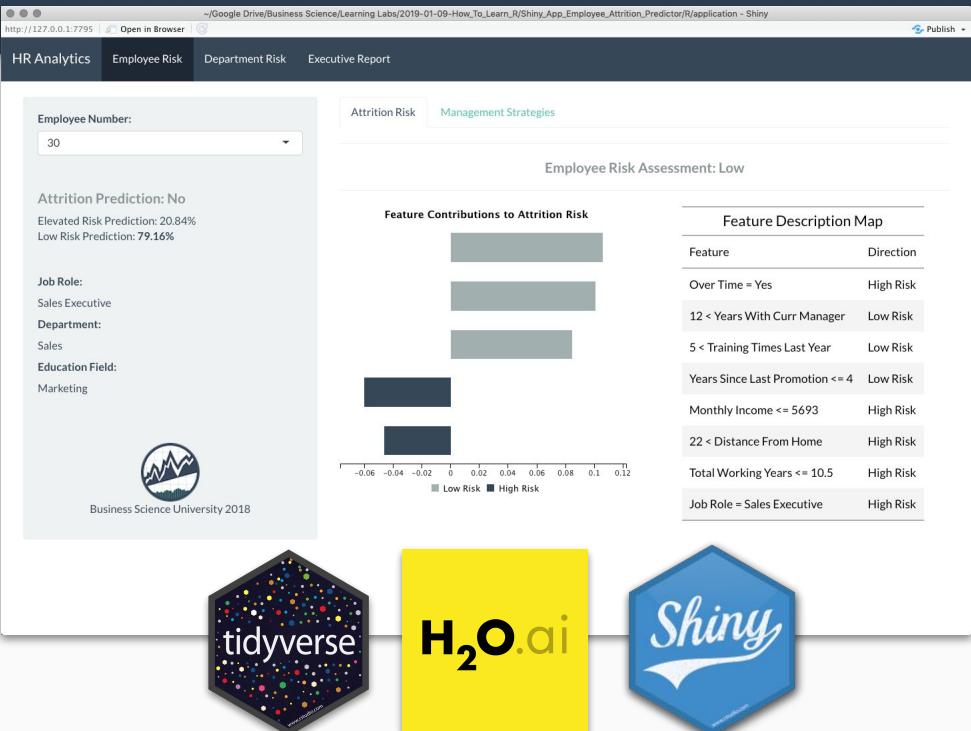
- \$15M/Year Problem
- Shiny Web App Example
- Data Science Workflow
 - Tidyverse
 - H2O & LIME
- Learning R for Business



Agenda: R For Business



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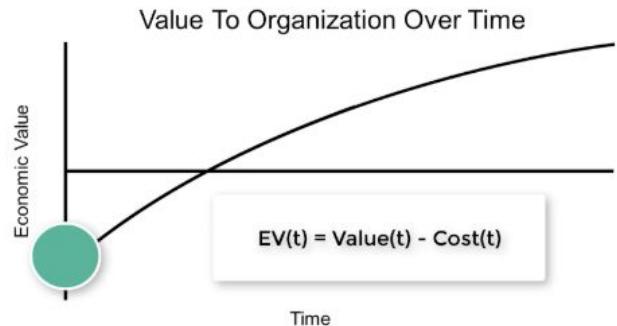
A close-up photograph of Bill Gates, an elderly man with grey hair and glasses, wearing a dark suit and red tie, smiling broadly. He is positioned on the left side of the frame, with a plain wall behind him.

YOU TAKE AWAY OUR
TOP 20 EMPLOYEES
AND WE BECOME A
MEDIocre COMPANY

- Bill Gates



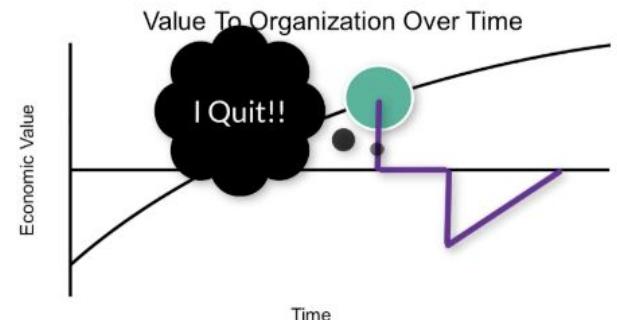
Economic Value Of Employee Over Time



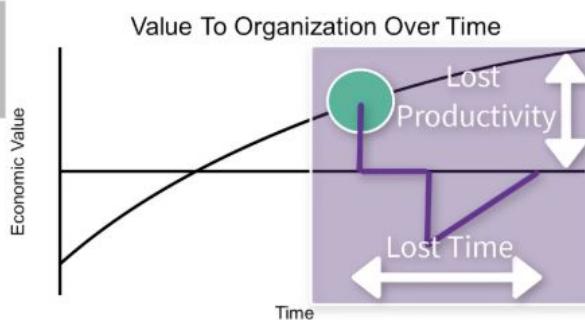
1



2



3

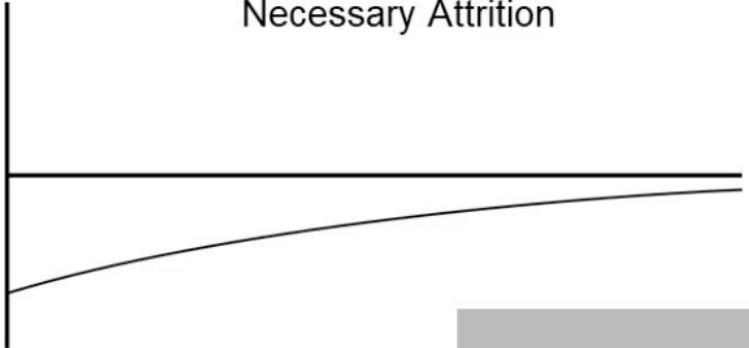


4

Good Attrition Vs Bad Attrition

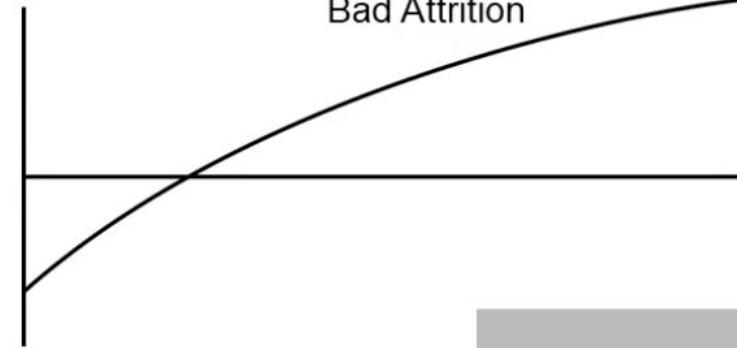


Necessary Attrition



1. POOR PERFORMERS

Bad Attrition



2. HIGH PERFORMERS

OK to lose these employees

Save these employees if possible



Cost of Attrition

```
calculate_attrition_cost <- function(  
  # Employee  
  n           = 1,  
  salary      = 80000,  
  
  # Direct Costs  
  separation_cost = 500,  
  vacancy_cost   = 10000,  
  acquisition_cost = 4900,  
  placement_cost  = 3500,  
  
  # Productivity Costs  
  net_revenue_per_employee = 250000,  
  workdays_per_year        = 240,  
  workdays_position_open   = 40,  
  workdays_onboarding      = 60,  
  onboarding_efficiency    = 0.50  
  
) {  
  
  # Direct Costs  
  direct_cost <- sum(separation_cost, vacancy_cost, acquisition_cost, placement_cost)  
  
  # Lost Productivity Costs  
  productivity_cost <- net_revenue_per_employee / workdays_per_year *  
    (workdays_position_open + workdays_onboarding * onboarding_efficiency)  
  
  # Savings of Salary & Benefits (Cost Reduction)  
  salary_benefit_reduction <- salary / workdays_per_year * workdays_position_open  
  
  # Estimated Turnover Per Employee  
  cost_per_employee <- direct_cost + productivity_cost - salary_benefit_reduction  
  
  # Total Cost of Employee Turnover  
  total_cost <- n * cost_per_employee  
  
  return(total_cost)  
}
```

• SIMPLE CALCULATION

Direct costs

Lost Productivity

Savings (Salary & Benefits)

• \$78K COST / EMPLOYEE

• IF ORGANIZATION LOSES 200 HIGH PERFORMERS EACH YEAR...

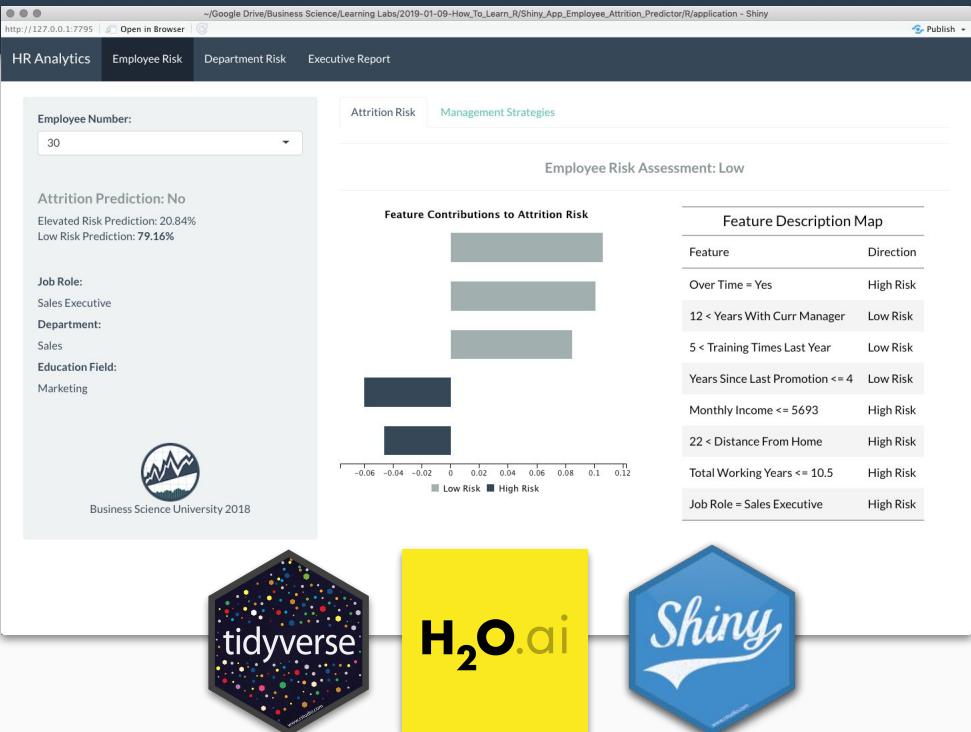
A young boy with blonde hair, wearing a grey herringbone vest over a pink long-sleeved shirt, stands against a dark background. He is holding an open book in front of him with both hands and has a wide-eyed, shocked expression on his face.

**\$15M / YEAR
PROBLEM**

Agenda: R For Business



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Demo Time!



File Edit Code View Plots Session Build Debug Profile Tools Window Help

~Google Drive/Business Science/Learning Labs/2019-01-09-How To Learn R_Fast/Shiny App_Employee Attrition Predictor - RStudio

```
1 # HR 301 Shiny Application
2 #
3 #
4 # library(shiny)
5 # library(shinythemes)
6 # library(DT)
7 # library(tidyverse)
8 # library(glue)
9 # library(lime)
10 # library(hilloorder)
11 # library(sheets)
12 # library(gt)
13 #
14 # source("employee_risk/rec_strategies.R")
15 # source("employee_risk/sheets_interactions.R")
16 # source("department_risk/department_summary.R")
17 pred <- readRDS("data/lime_prediction_results.RDS")
18 #
19 mgmt_note_fields <- c("obs", "mgmt.notes")
20 ui <- fluidPage(
21   headerPanel(
22     theme = shinytheme("flatly"),
23     "HR Analytics",
24     "Employee Risk",
25     "NoveltyLog"
26   ),
27   mainPanel(
28     sidebarPanel(
29       selectInput("obs", "Employee Number:", choices = application_data$EmployeeNumber),
30       hr(),
31       htmlOutput("model_pred"),
32       hr(),
33       htmlOutput("role"),
34       hr(),
35       htmlOutput("dept"),
36     ),
37     hr(),
38     uiOutput(session$selectedInputs)
39   )
40 )
41 
```

Console Terminal Job ~Google Drive/Business Science/Learning Labs/2019-01-09-How To Learn R_Fast/Shiny App_Employee Attrition Predictor

label: unnamed-chunk-2 (with options)

List of 1

\$ echo: logi FALSE

Quitting from lines 17-22 (dept_report.Rmd)

Warning: Error in filter_(impl): Evaluation error: object 'Attrition' not found.

[No stack trace available]

Tue Jan 16 3:27 PM Matthew Dancho

Shiny_App_Employee_Attrition_Predictor

Environment History Connections QR

Staged Status Path

Icons R/icon R/application/icon R/application/data/icon R/application/department_risk/icon R/application/employee_risk/icon

http://127.0.0.1:6570 Open in Browser

HR Analytics Employee Risk Department Risk Executive Report

Employee Number: 891

Attrition Prediction: Yes

Elevated Risk Prediction: 47.08%
Low Risk Prediction: 52.92%

Job Role: Manager

Department: Human Resources

Education Field: Human Resources

Shiny Business Science University 2018

Employee Risk Assessment: Low

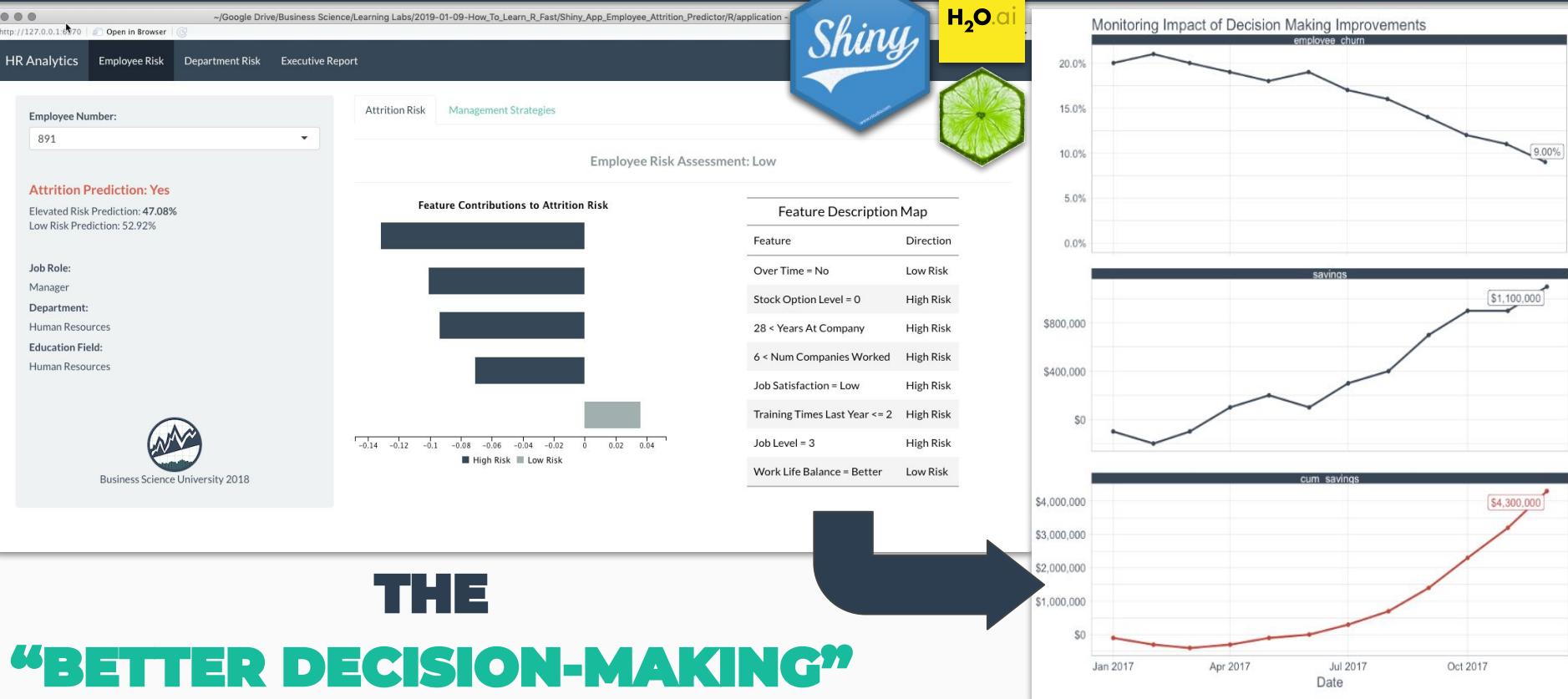
Attrition Risk Management Strategies

Feature Contributions to Attrition Risk

Feature	Direction
Over Time = No	Low Risk
Stock Option Level = 0	High Risk
28 < Years At Company	High Risk
6 < Num Companies Worked	High Risk
Job Satisfaction = Low	High Risk
Training Times Last Year <= 2	High Risk
Job Level = 3	High Risk
Work Life Balance = Better	Low Risk



Cause & Effect

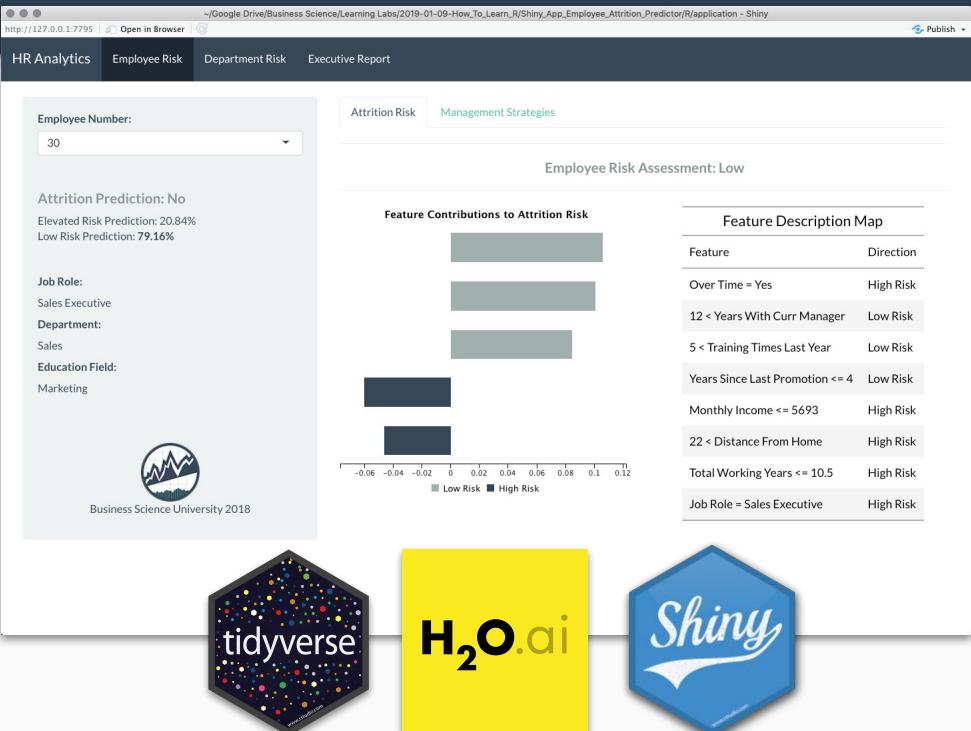


THE
“BETTER DECISION-MAKING”
EFFECT

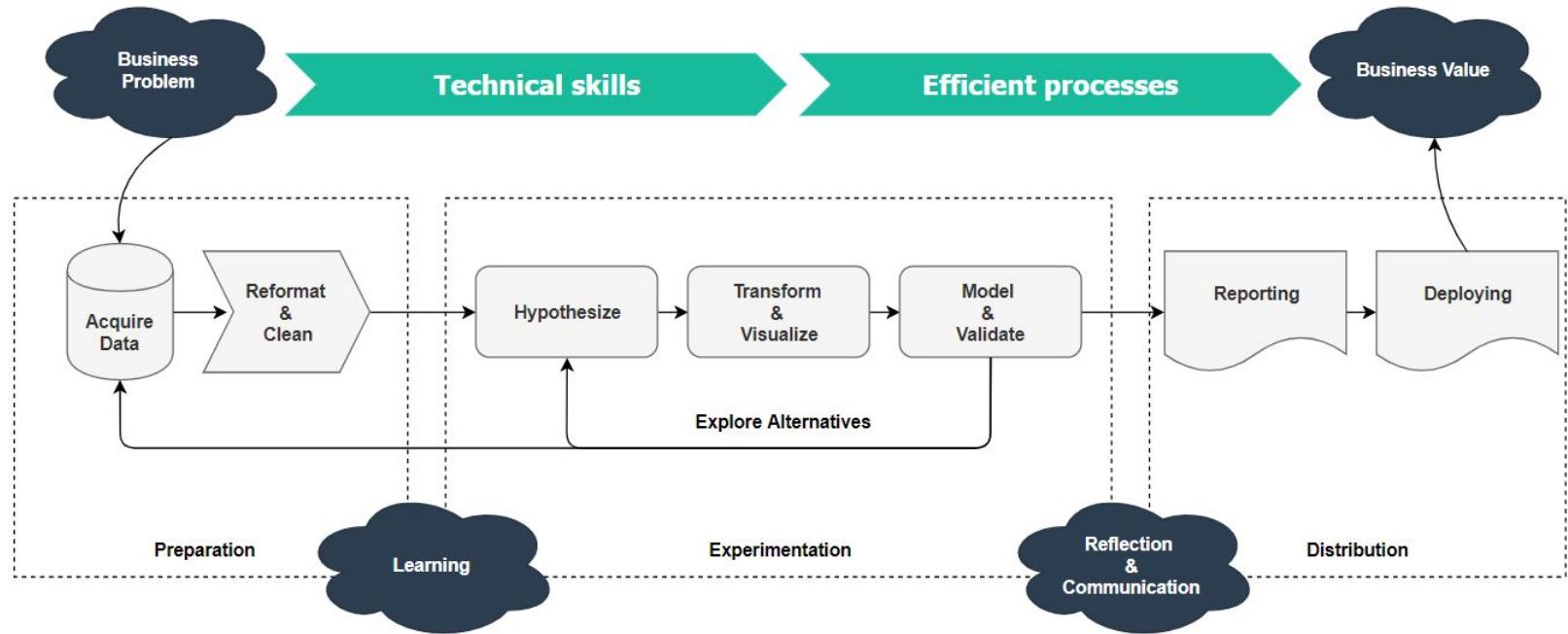
Agenda: R For Business



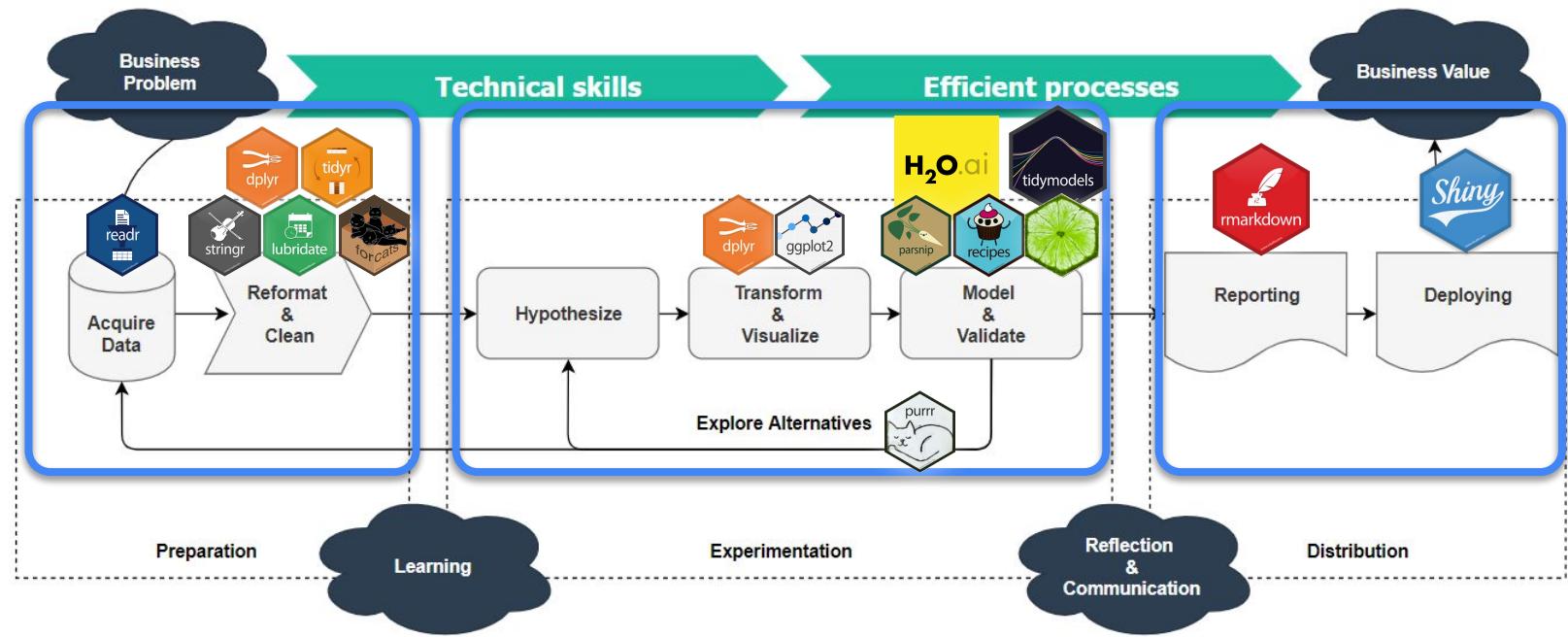
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Data Science Workflow



Data Science Workflow



R Just Fits!



Data Science Workflow

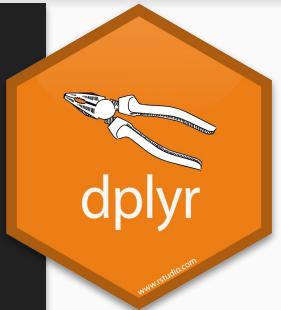


The  Toolchain

Data Science Workflow



```
# 1C. Measure The Drivers ----  
  
# Collect Information on Employee Attrition: On going  
  
# Develop KPI's: Industry KPIs: 8.8%  
  
dept_job_role_tbl %>%  
  
  count(Department, JobRole, Attrition) %>%  
  
  count_to_pct(Department, JobRole) %>%  
  
  assess_attrition(Attrition, attrition_value = "Yes", baseline_pct = 0.088) %>%  
  
  mutate(  
    cost_of_attrition = calculate_attrition_cost(n = n, salary = 80000)  
  )  
  ...
```

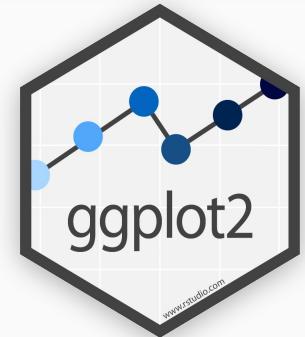
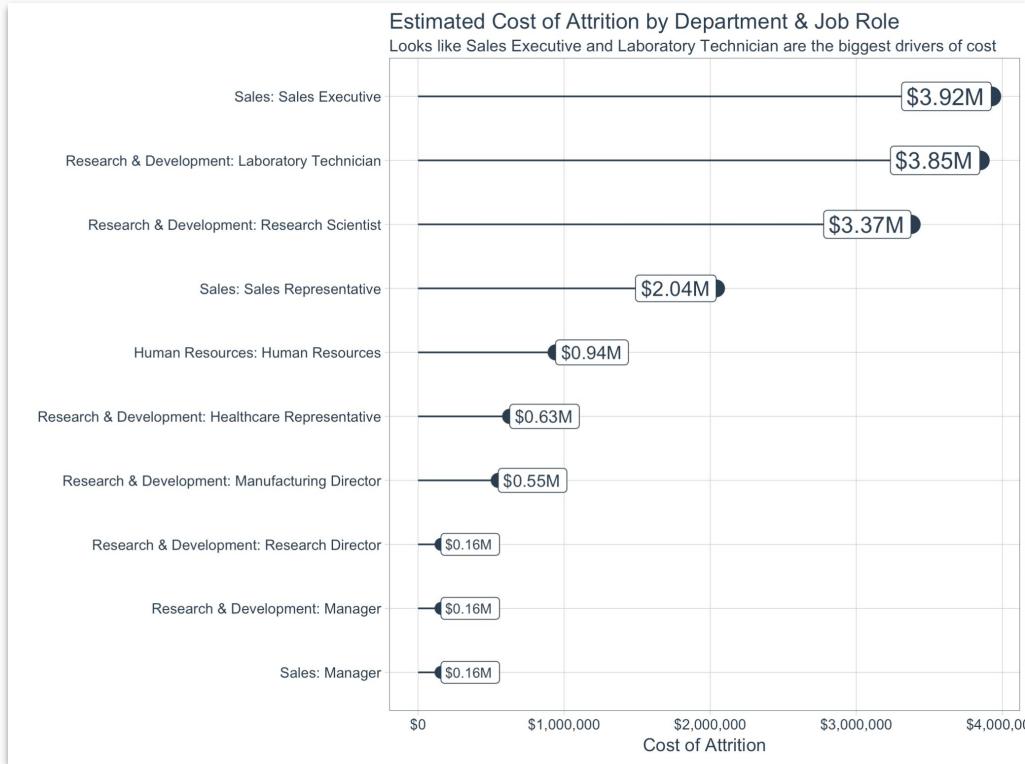


Department <chr>	JobRole <chr>	Attrition <chr>	n <int>	pct <dbl>	above_industry_avg <chr>	cost_of_attrition <dbl>
Sales	Sales Representative	Yes	26	0.40000000	Yes	2040566.7
Human Resources	Human Resources	Yes	12	0.30769231	Yes	941800.0
Research & Development	Laboratory Technician	Yes	49	0.21875000	Yes	3845683.3
Sales	Sales Executive	Yes	50	0.18315018	Yes	3924166.7
Research & Development	Research Scientist	Yes	43	0.16602317	Yes	3374783.3
Research & Development	Healthcare Representative	Yes	8	0.07619048	No	627866.7
Sales	Manager	Yes	2	0.06451613	No	156966.7
Research & Development	Manufacturing Director	Yes	7	0.05691057	No	549383.3
Research & Development	Manager	Yes	2	0.04166667	No	156966.7
Research & Development	Research Director	Yes	2	0.02739726	No	156966.7

1-10 of 10 rows

Size the Problem

Data Science Workflow



Visualize Attrition Cost

Data Science Workflow



```
# 2. Modeling ----  
  
h2o.init()  
  
split_h2o <- h2o.splitFrame(as.h2o(train_tbl), ratios = c(0.85), seed = 1234)  
  
train_h2o <- split_h2o[[1]]  
valid_h2o <- split_h2o[[2]]  
test_h2o <- as.h2o(test_tbl)  
  
y <- "Attrition"  
x <- setdiff(names(train_h2o), y)  
  
automl_models_h2o <- h2o.automl(  
  x = x,  
  y = y,  
  training_frame = train_h2o,  
  validation_frame = valid_h2o,  
  leaderboard_frame = test_h2o,  
  max_runtime_secs = 30,  
  nfolds = 5  
)
```

H₂O.ai



Predict Attrition Risk

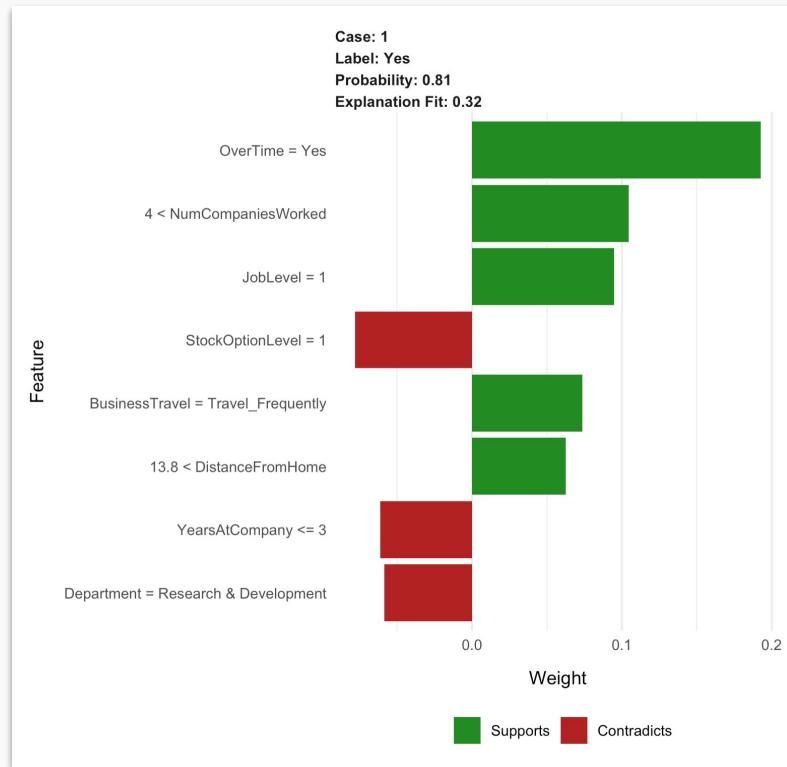


Data Science Workflow

```
explainer <- train_tbl %>%
  select(-Attrition) %>%
  lime(
    model              = automl_leader,
    bin_continuous     = TRUE,
    n_bins             = 4,
    quantile_bins      = TRUE
  )

explainer

explanation <- test_tbl %>%
  slice(5) %>%
  select(-Attrition) %>%
  lime::explain(
    explainer = explainer,
    n_labels   = 1,
    n_features = 8,
    n_permutations = 5000,
    kernel_width   = 1
  )
```



Explain Features Attrition Risk

Data Science Workflow



~/Google Drive/Business Science/Learning Labs/2019-01-09-How_To_Learn_R_Fast/Shiny_App_Employee_Attrition_Predictor/R/application - Shiny
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H₂O.ai

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Business Science University 2018

Attrition Risk Management Strategies

Employee Risk Assessment: Low

Feature Contributions to Attrition Risk

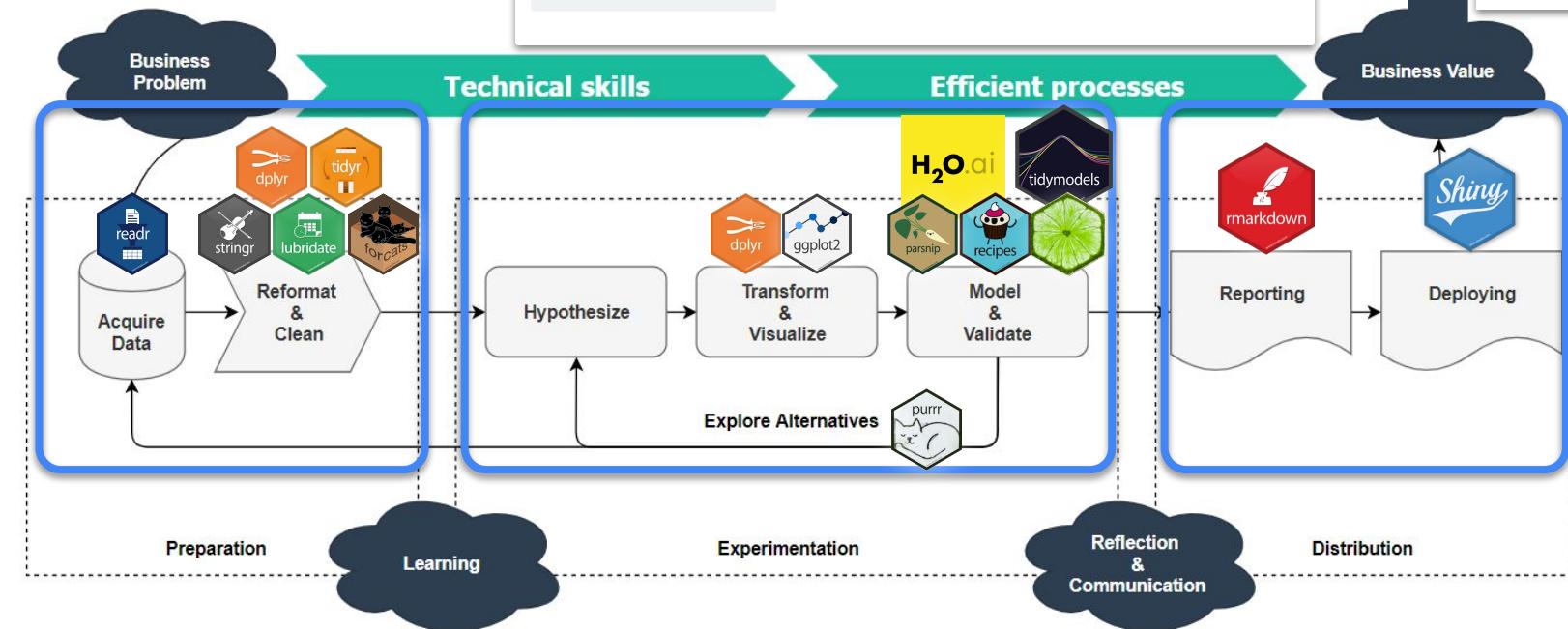
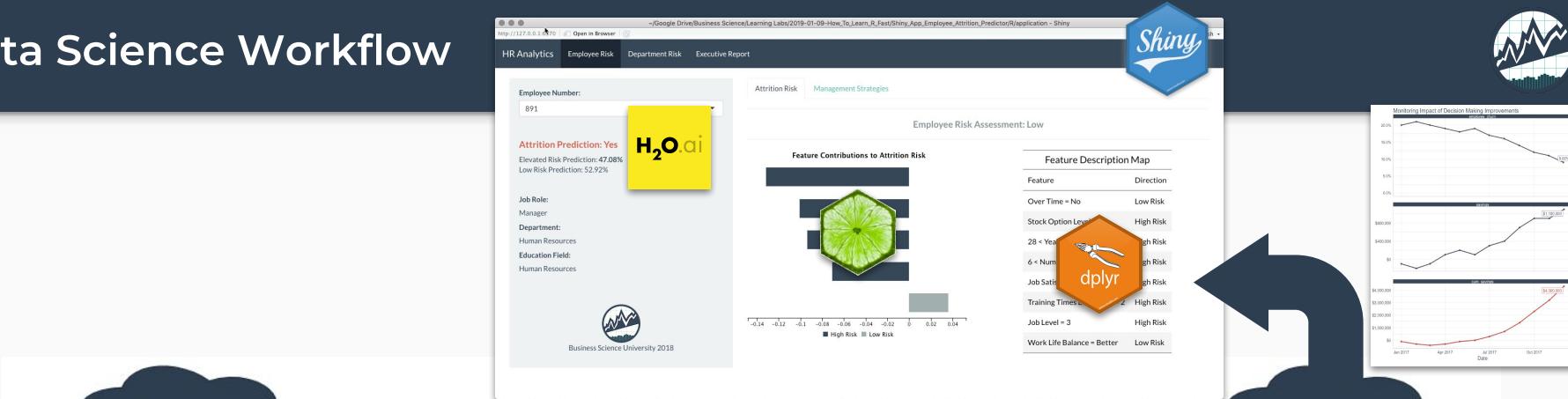
Feature Description Map

Feature Direction

Over Time = No	Low Risk
Stock Option Level = 1	High Risk
28 < Years = No	High Risk
6 < Number of Children = No	High Risk
Job Satisfaction = High	High Risk
Training Times Last Year > 2	High Risk
Job Level = 3	High Risk
Work Life Balance = Better	Low Risk

Shiny

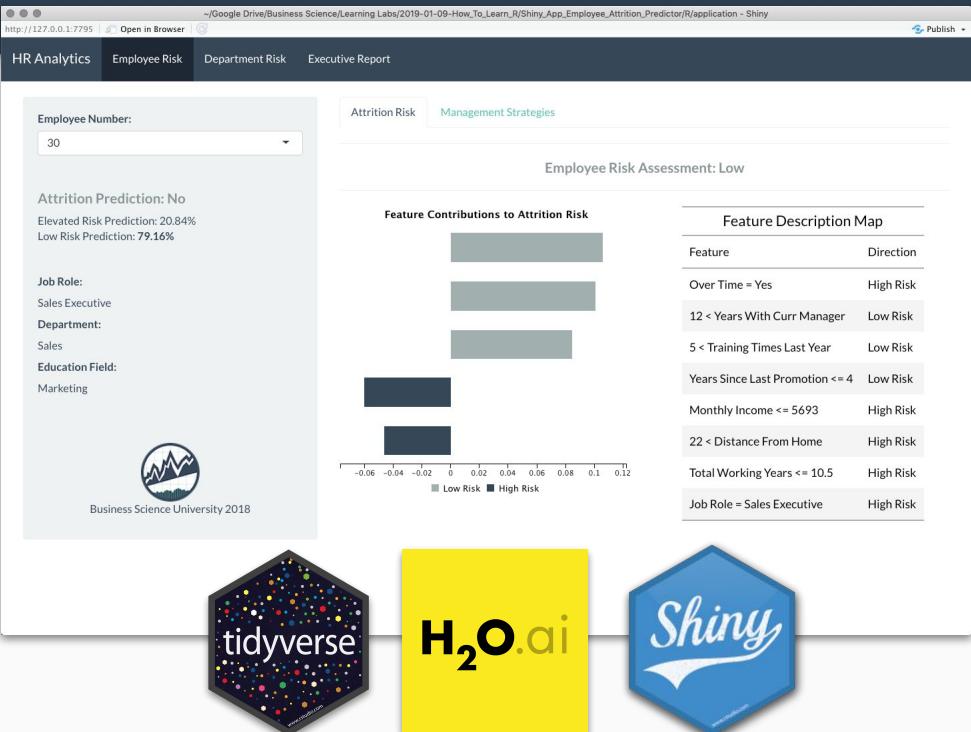
Data Science Workflow



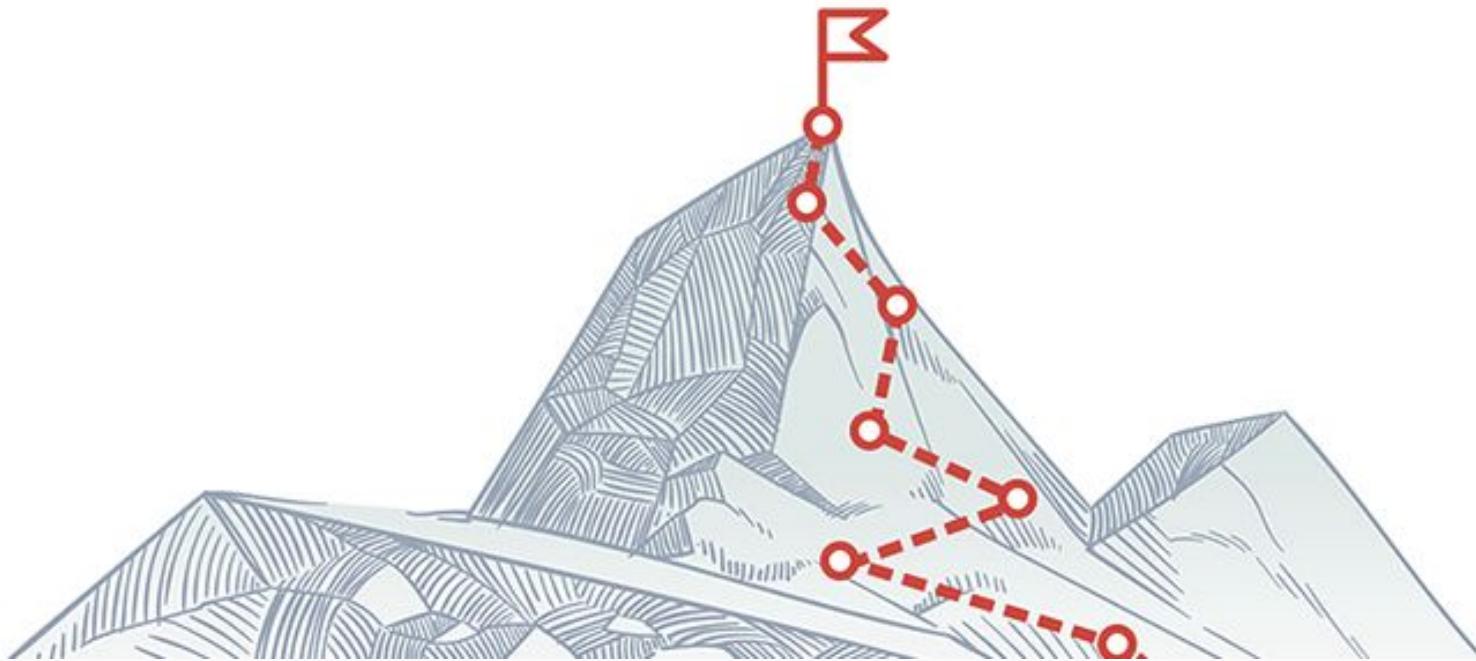
Agenda: R For Business



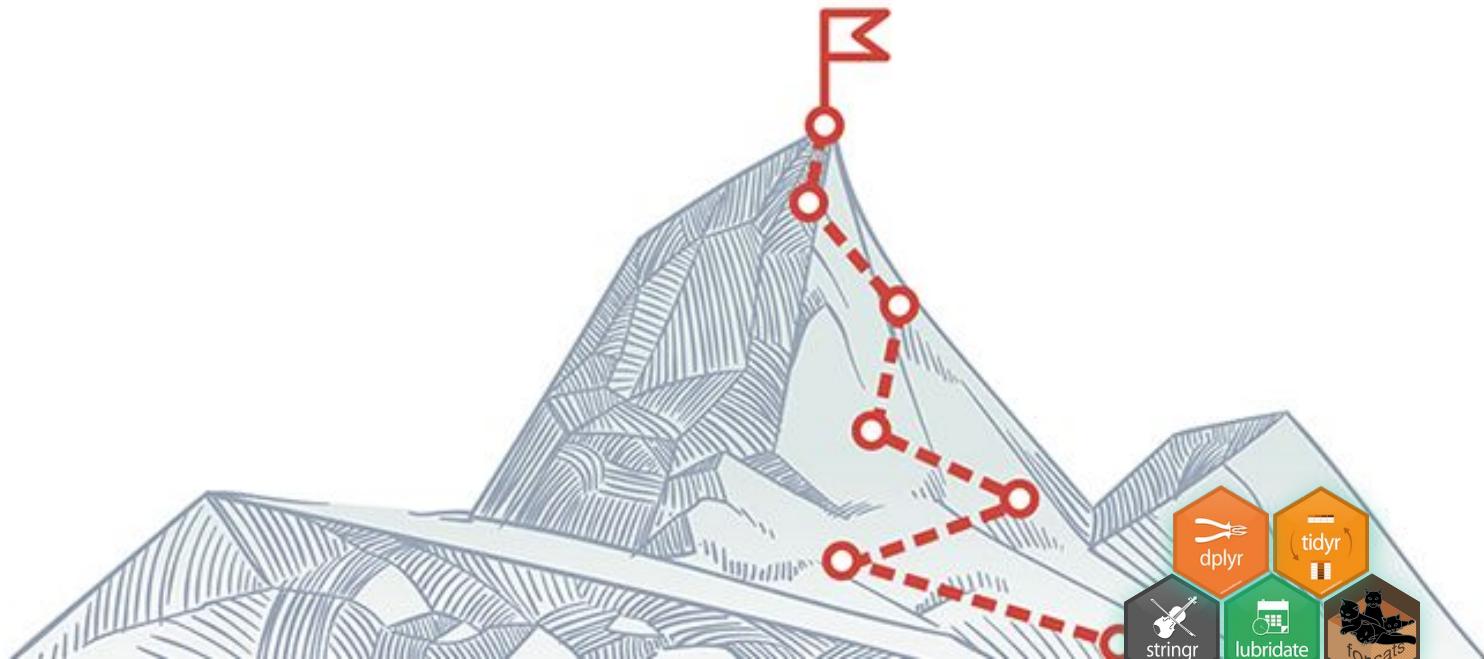
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Learning R is a Hill Climb

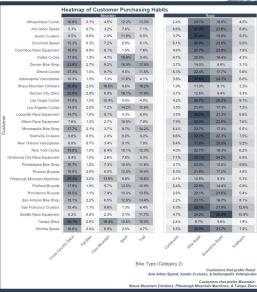
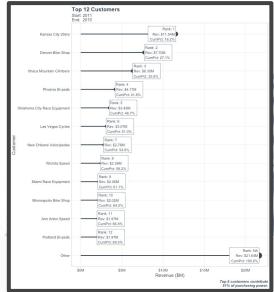


Learning R is a Hill Climb



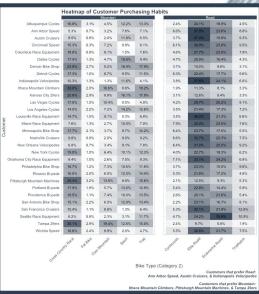
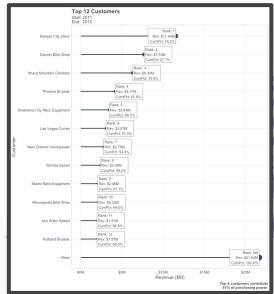
**Data Cleaning
& Manipulation**

Learning R is a Hill Climb

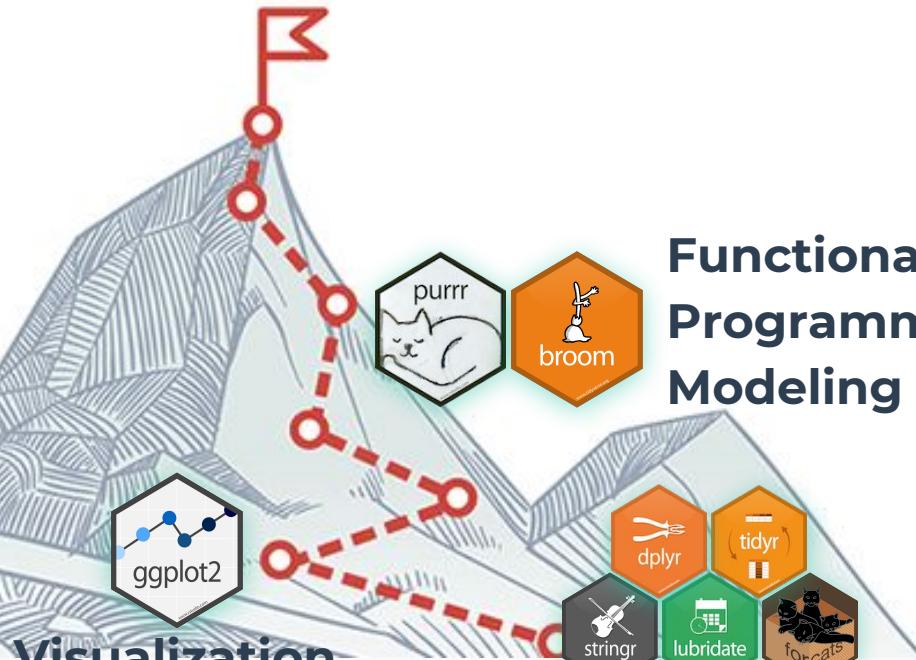


Data Cleaning
& Manipulation

Learning R is a Hill Climb



Visualization

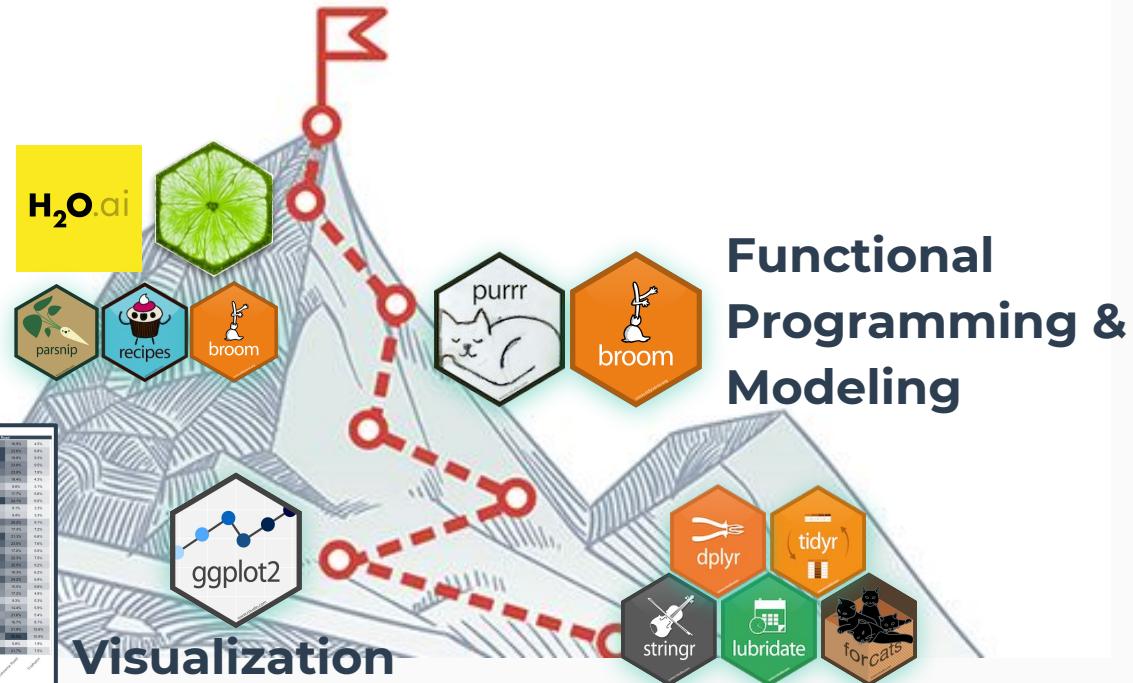
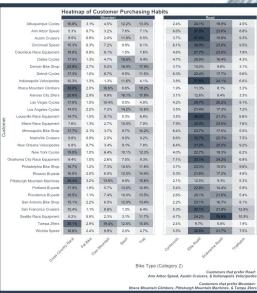
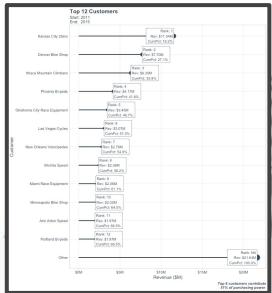


Data Cleaning
& Manipulation

Learning R is a Hill Climb



Advanced Data Science

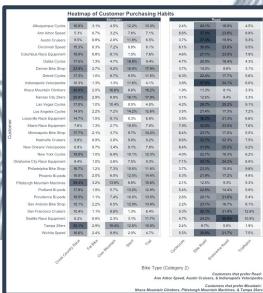
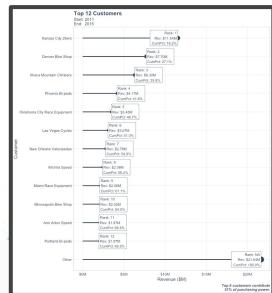


Data Cleaning & Manipulation

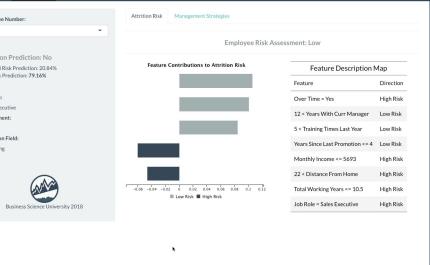
Learning R is a Hill Climb

THE GOAL

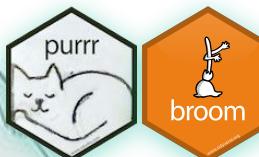
Advanced
Data Science



Visualization



Functional
Programming &
Modeling

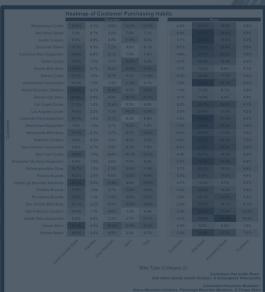
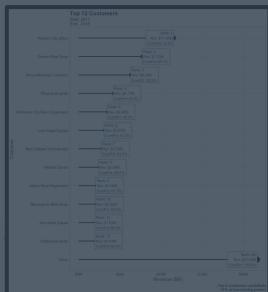


Data Cleaning
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Learning R is a Hill Climb

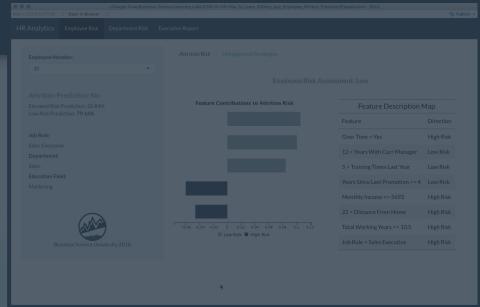
Path To Top Can Be Accomplished FAST

Advanced Data Science



Visualization

Data Cleaning & Manipulation



THE PLAN



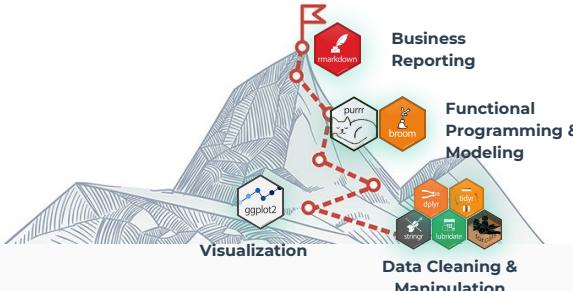
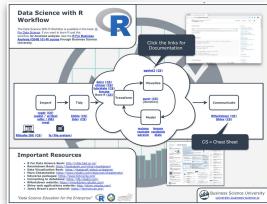
Business Analysis with R (DS4B 101-R)

Data Science For Business with R (DS4B 201-R)

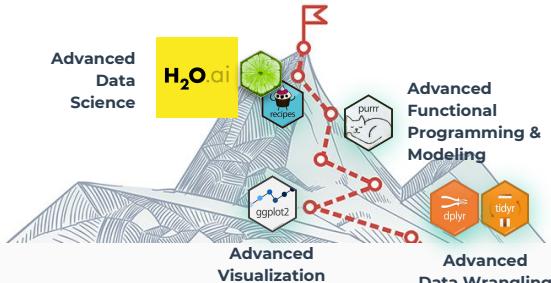
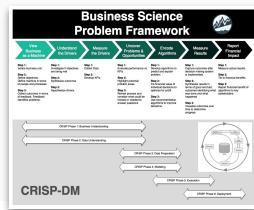
R Shiny Web Apps For Business (DS4B 301-R)

Project-Based Courses with Business Application

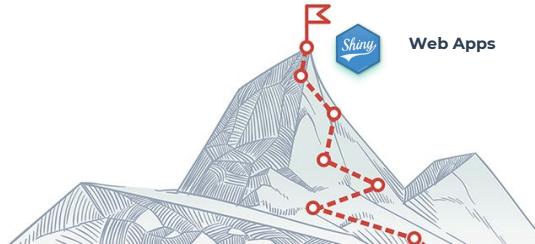
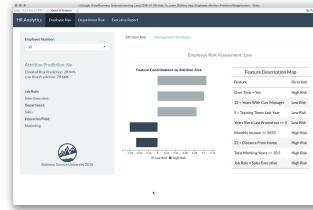
Data Science Foundations **7 Weeks**



Machine Learning & Business Consulting **10 Weeks**



Web Application Development **6 Weeks**





THE RESULT



Business Analysis with R

(DS4B 101-R)

7 WEEKS



Data Science For Business with R

(DS4B 201-R)

10 WEEKS



R Shiny Web Apps For Business

(DS4B 301-R)

6 WEEKS

23 WEEKS!
(OR LESS)



THE BONUS



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PROMO Code: `rstudio`**
university.business-science.io

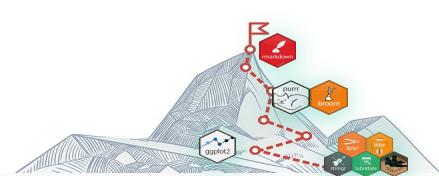
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(DS4B 201-R)**

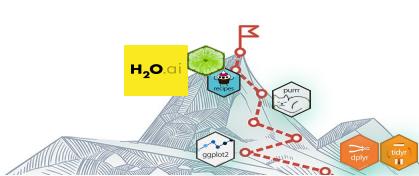
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