

# SPARK+AI SUMMIT 2020

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# How R Developers Can Build and Share Data and AI Applications that Scale with Databricks and Rstudio Connect

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# Agenda

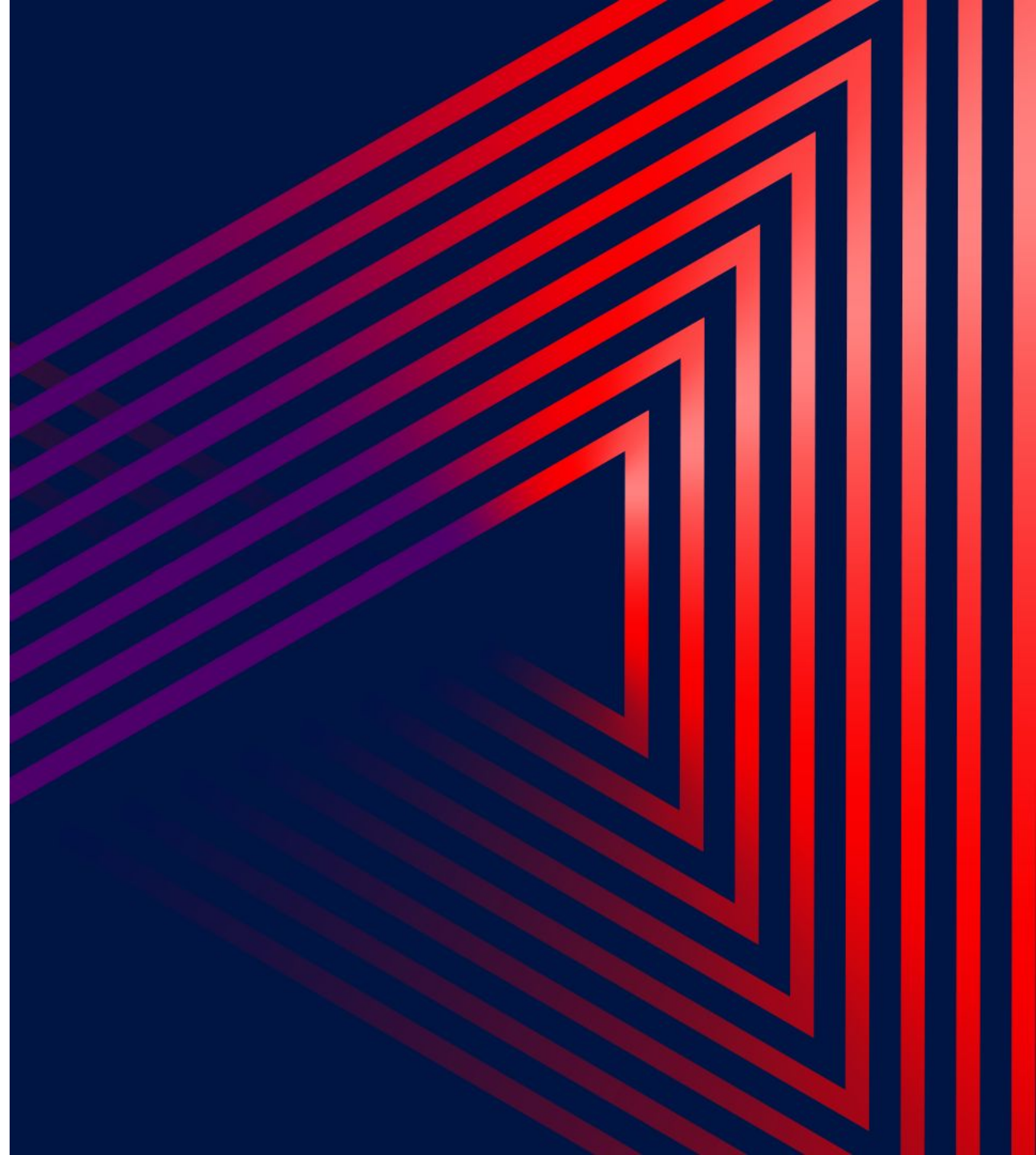
## Rafi Kurlansik, Databricks

Building Scalable R and Shiny apps with RStudio  
and Databricks

## James Blair, RStudio PBC

Deploying Scalable Shiny apps with RStudio  
Connect and Databricks

Benchmarking performance of Shiny connections  
to Spark





# How to scale R and Shiny with RStudio and Databricks

# How can we open up the data lake to R users?

Imagine trying to do so with traditional R development...

- Typical development patterns

- Local
- Cloud / On Prem VM

- Challenges with big data

- Server memory - can only process so much data in the app itself before crashing R
- Performance - even on a powerful VM, eventually see our app get less responsive as we reach 100+ GBs
- Managing big data infrastructure - app value must be higher to justify the energy investment

If only there was a technology with a familiar API in R that let our app scale to process 100s of GBs...

# Scale R Apps with Databricks and RStudio

Databricks Spark, RStudio IDE

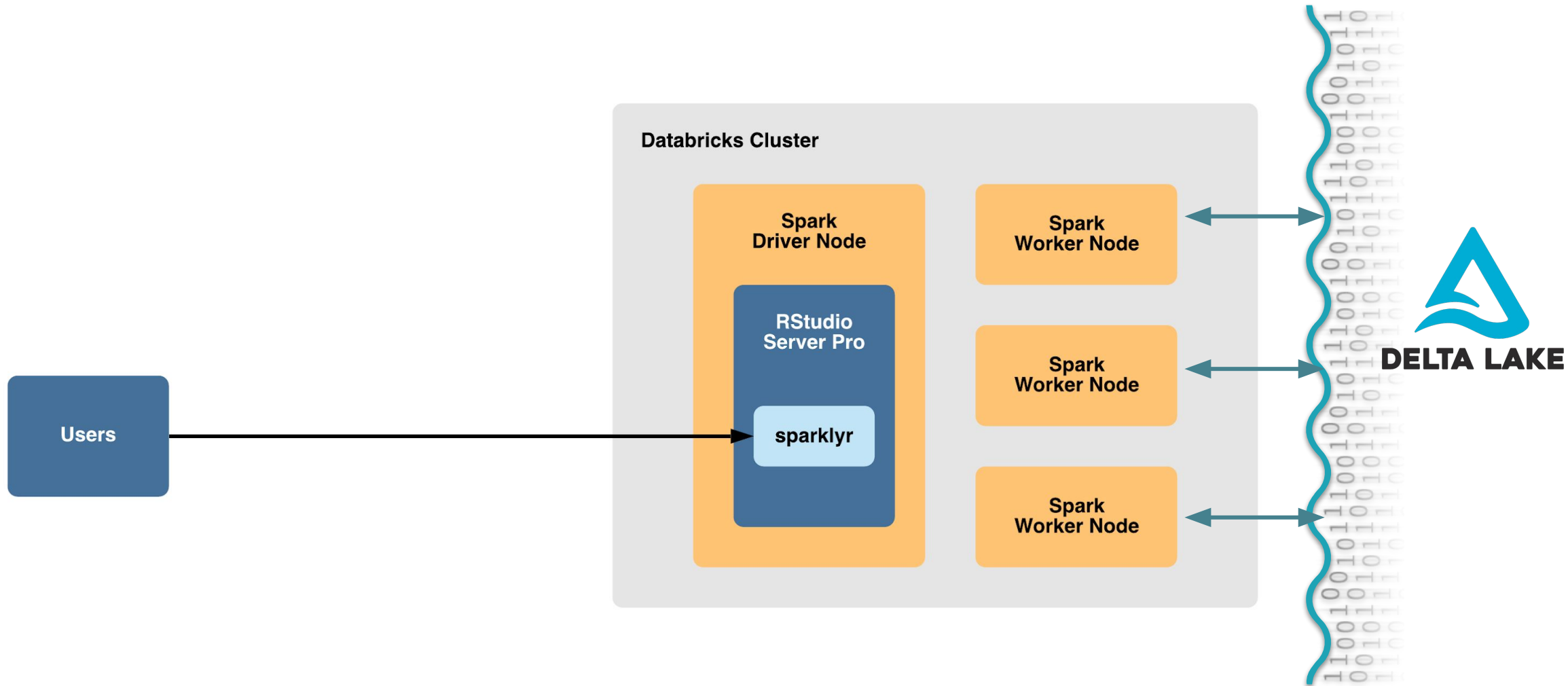
## ■ Development Patterns

- Hosted RStudio Server (Pro) on Databricks Cluster
- RStudio with remote Spark access using Databricks Connect

## ■ Overcoming challenges with big data

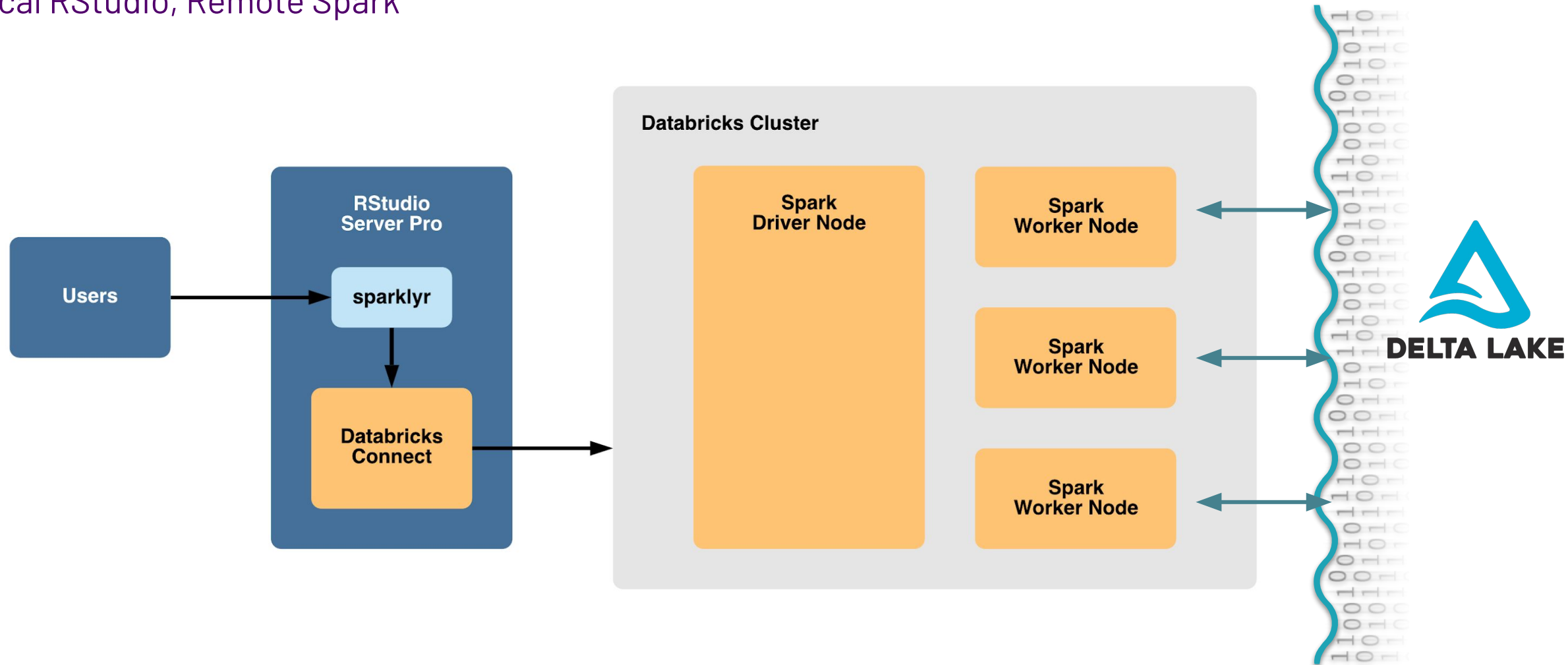
- Auto-scaling Databricks Spark Clusters - dynamically respond to accommodate larger data processing tasks
- Consistently fast performance with Delta Lake and Databricks Runtime
- Managed service allows data teams to focus on building data products, not maintaining infrastructure

# Hosted RStudio Server Pro on Databricks



# RStudio with Databricks Connect

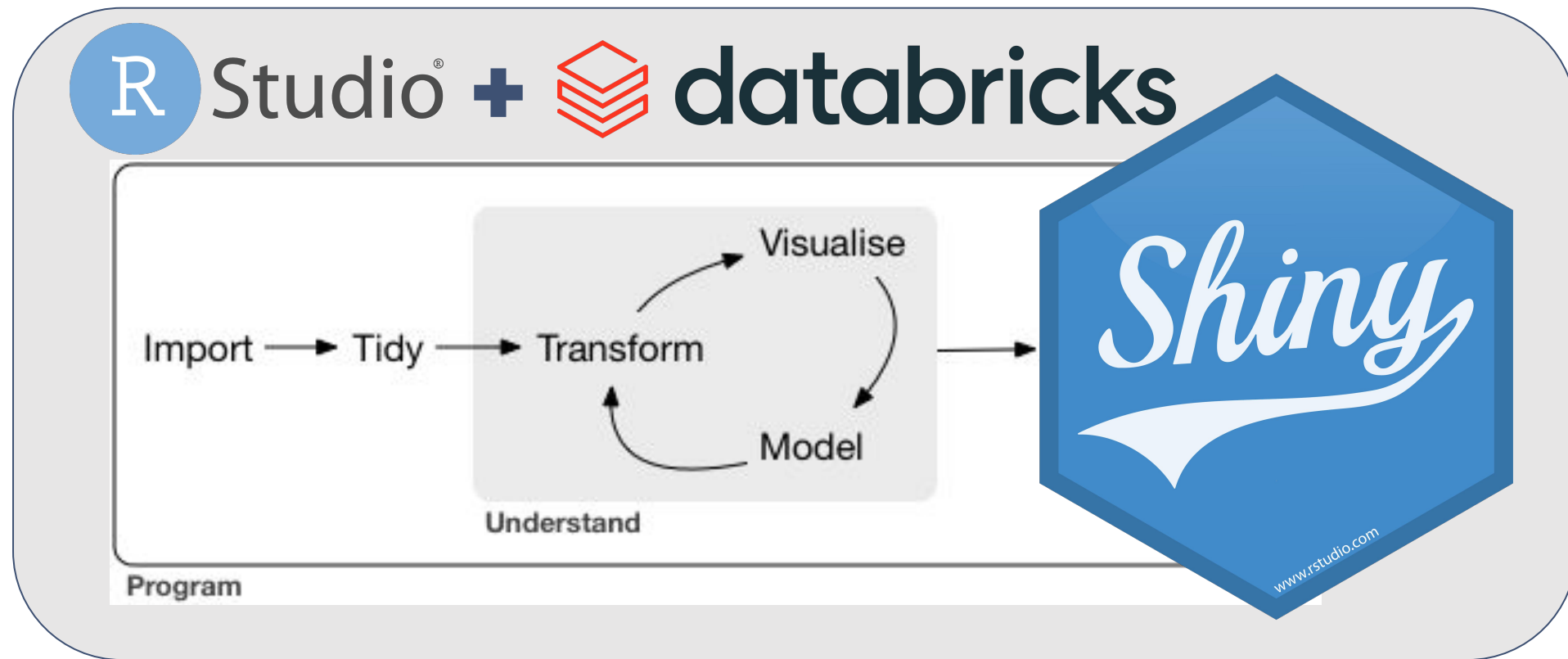
Local RStudio, Remote Spark





# Sharing Scalable Shiny Apps

# The Data Science Process



# Shiny and Spark: A cautionary tale

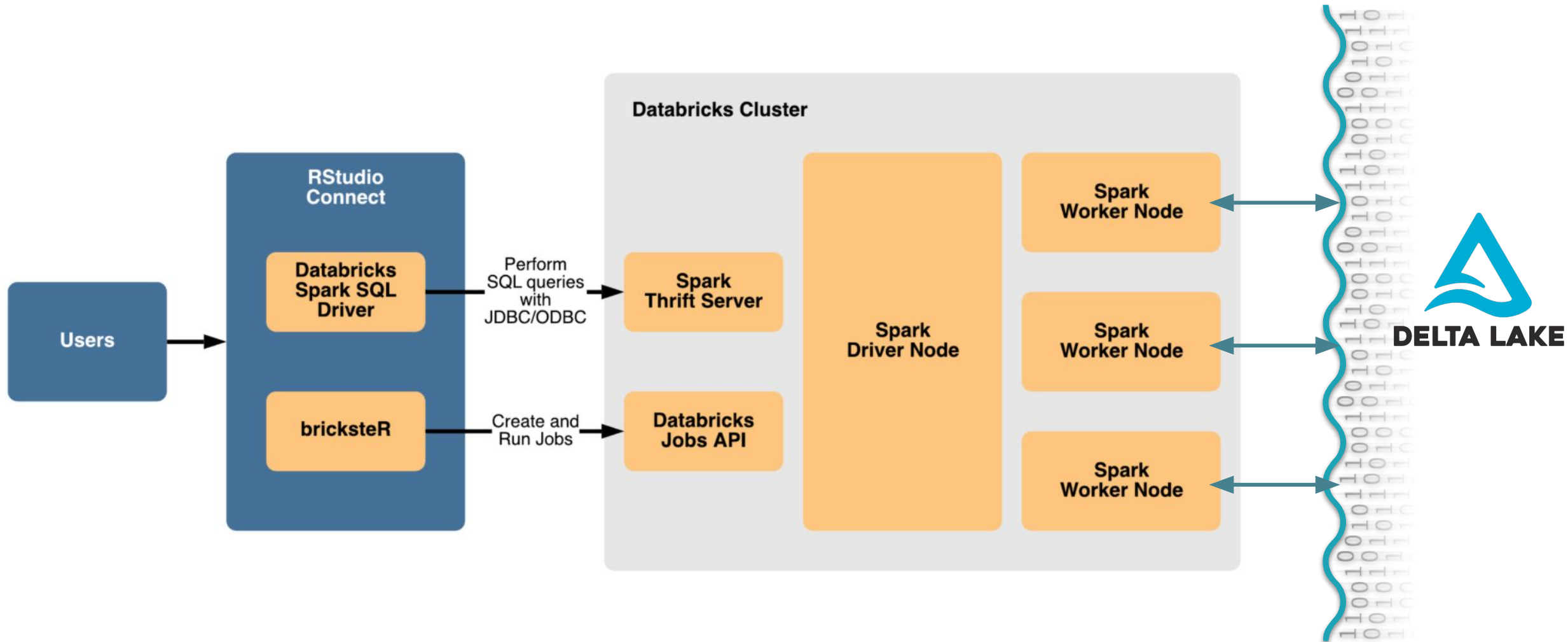


# ODBC to the Rescue

- The R + ODBC toolchain is robust and stable
- As performant as a native Spark connection
- Easy to migrate code from sparklyr to ODBC
- Spark still does all of the computation
- Databricks provides an optimized Spark ODBC driver



# ODBC with RStudio Connect

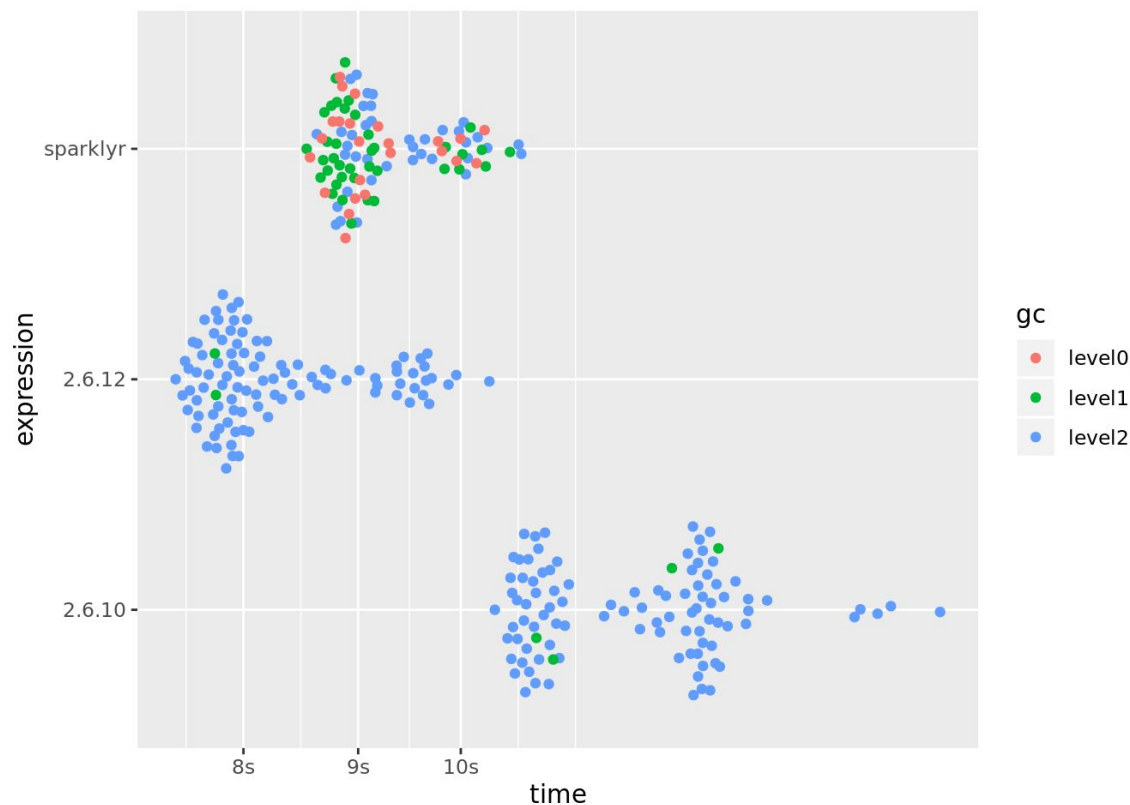




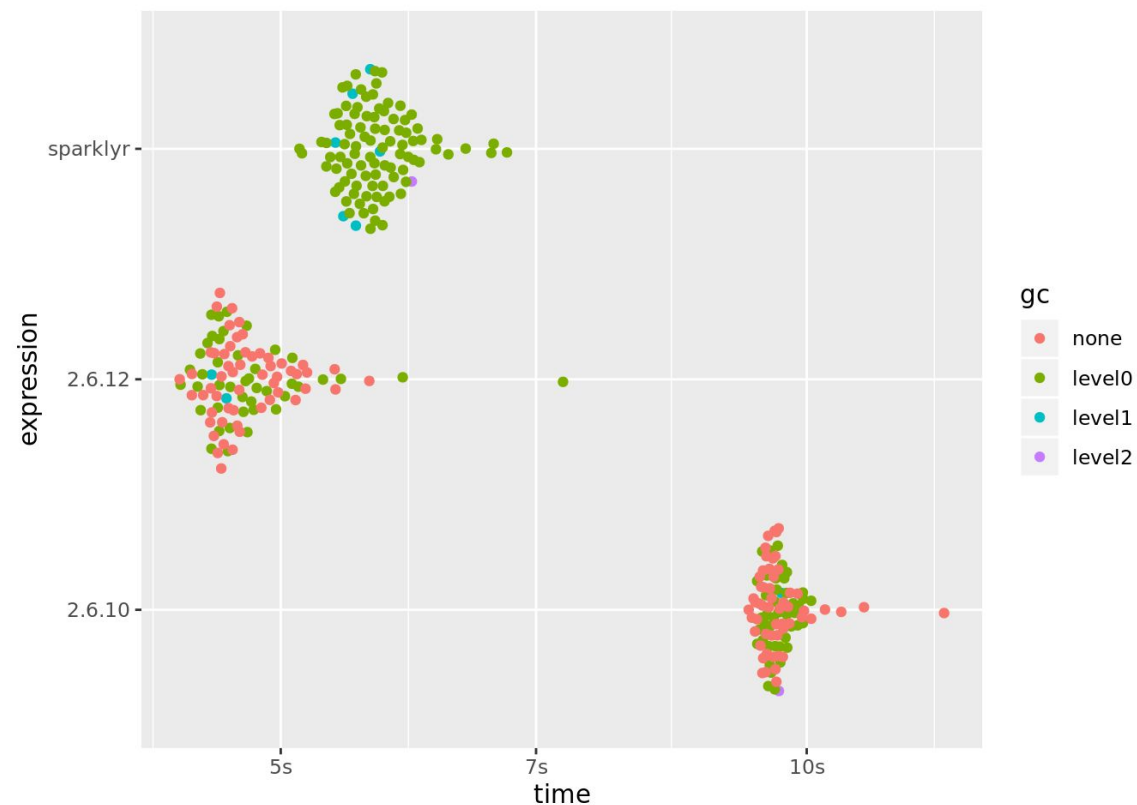
# ODBC Performance

Comparing sparklyr against two versions of the Databricks ODBC/JDBC Driver

## Collecting



## Joins



# Sparklyr to ODBC

```
``{r sparklyr}
library(tidyverse)
library(sparklyr)

spark_home <- system("databricks-connect get-spark-home", intern = TRUE)
sc <- spark_connect(method = "databricks", spark_home = spark_home)

all_flights <- tbl(sc, "all_flights")

all_flights %>%
  count(Year) %>%
  arrange(Year)

spark_disconnect(sc)
```

The screenshot shows a terminal window with a table of data. The table has two columns: 'Year' and 'x'. The data rows are as follows:

Year	x
1980	1311826
1988	1202096
1989	1641208
1999	3279853
1995	1078325
1993	1092117
1992	1477841
1994	
1997	
1991	

Overlaid on the bottom right of the terminal is the text 'tbl\_sql' in a large, bold, black font, with '0 x 2' below it.

Year	n
<int>	<dbl>
1987	1311826
1988	5202096
1989	5041200
1990	5270893
1991	5076925
1992	5092157
1993	5070501
1994	5180048
1995	5327435
1996	5351983

1-10 of 22 rows

Previous **1** 2 3 Next

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```
```{r odbc}  
library(tidyverse)  
library(DBI)  
  
sc <- dbConnect(odbc::odbc(), "databricks")  
  
all_flights <- tbl(sc, "all_flights")  
  
all_flights %>%  
  count(Year) %>%  
  arrange(Year)  
  
dbDisconnect(sc)  
```
```

| Year<br><int> | n<br><S3: integer64> |
|---------------|----------------------|
| 1987          | 1311826              |
| 1988          | 5202096              |
| 1989          | 5041200              |
| 1990          | 5270893              |
| 1991          | 5076925              |
| 1992          | 5092157              |
| 1993          | 5070501              |
| 1994          | 5180048              |
| 1995          | 5327435              |
| 1996          | 5351983              |

1-10 of 22 rows

Previous **1** 2 3 Next

#Datateams #SparkAISummit

# Conclusion



## Develop at scale

- Interactive data analysis with SparkSQL
  - sparklyr
  - ODBC
- Other Spark APIs
  - sparklyr

## Deploy at scale

- Interactive data analysis with SparkSQL
  - Shiny with ODBC
- Other Spark APIs
  - `\_(ツ)_/`
  - Deploy models with MLflow?
  - Submit individual commands with Databricks REST API 1.2?
  - Run sparklyr jobs from RStudio on Databricks with bricksteR?
  - Stay tuned....

# Additional Resources

## Documentation

- [Hosted RStudio on Databricks](#)
- [Databricks Connect](#)
- [ODBC](#)
- [ODBC Configuration](#)
- [RStudio Connect](#)
- [Sparklyr](#)

## Related Repos

- [blairj09-talks/spark-summit-2020](#)
- [RafiKurlansik/bricksteR](#)
- [delta-io/delta](#)
- [sparklyr/sparklyr](#)

# Feedback

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