

FP&A with Spreadsheets and Spark

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#SAISEco2

About

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- Research Interests:
 - Program Transformation,
 - Model-driven Data Product Design & Development

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 - ▶ Design and Development of
 - ▶ Large-scale Data Products
 - ▶ with Spreadsheets as Models

Prototyping ...

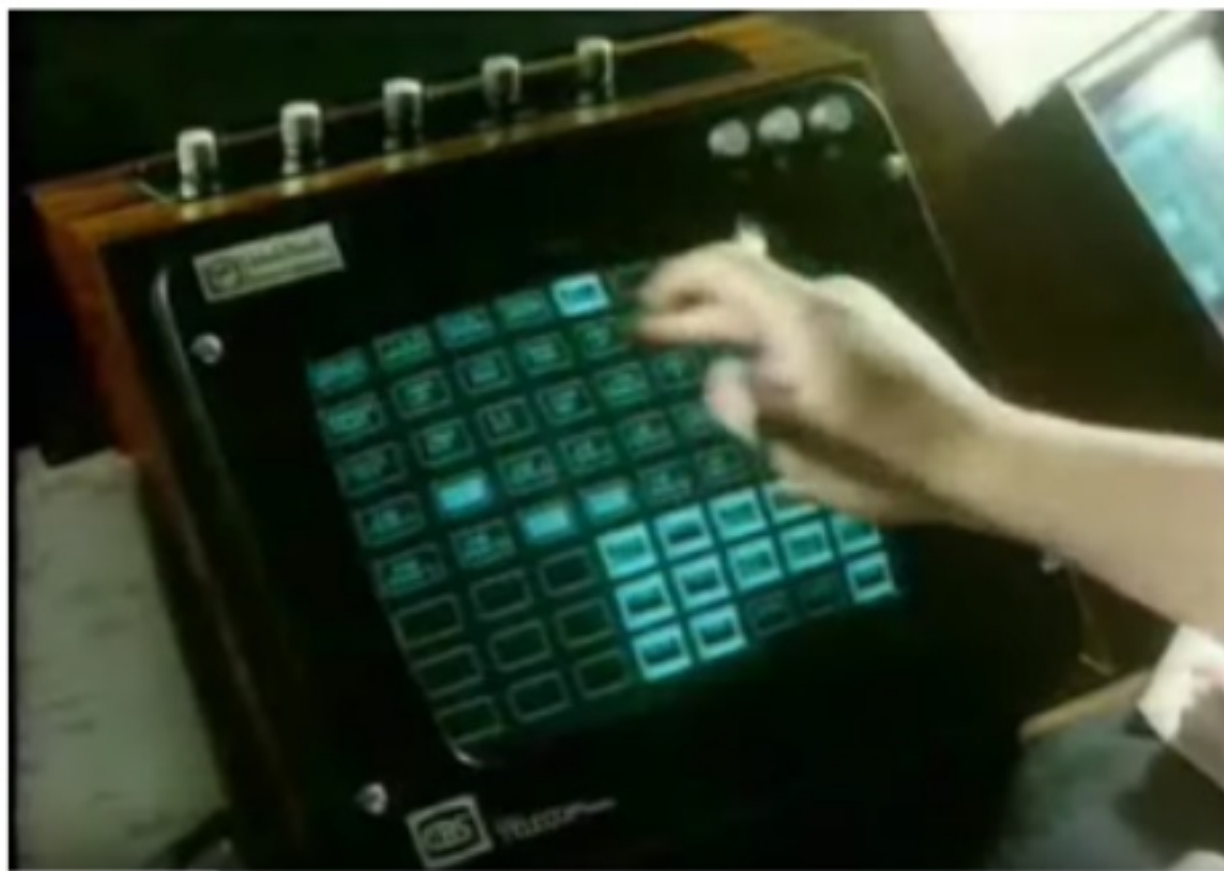


<http://bit.ly/2e5GmyY>

Prototyping Spark programs with ...



<http://bit.ly/2e5GmyY>



<http://bit.ly/2edYfMs>

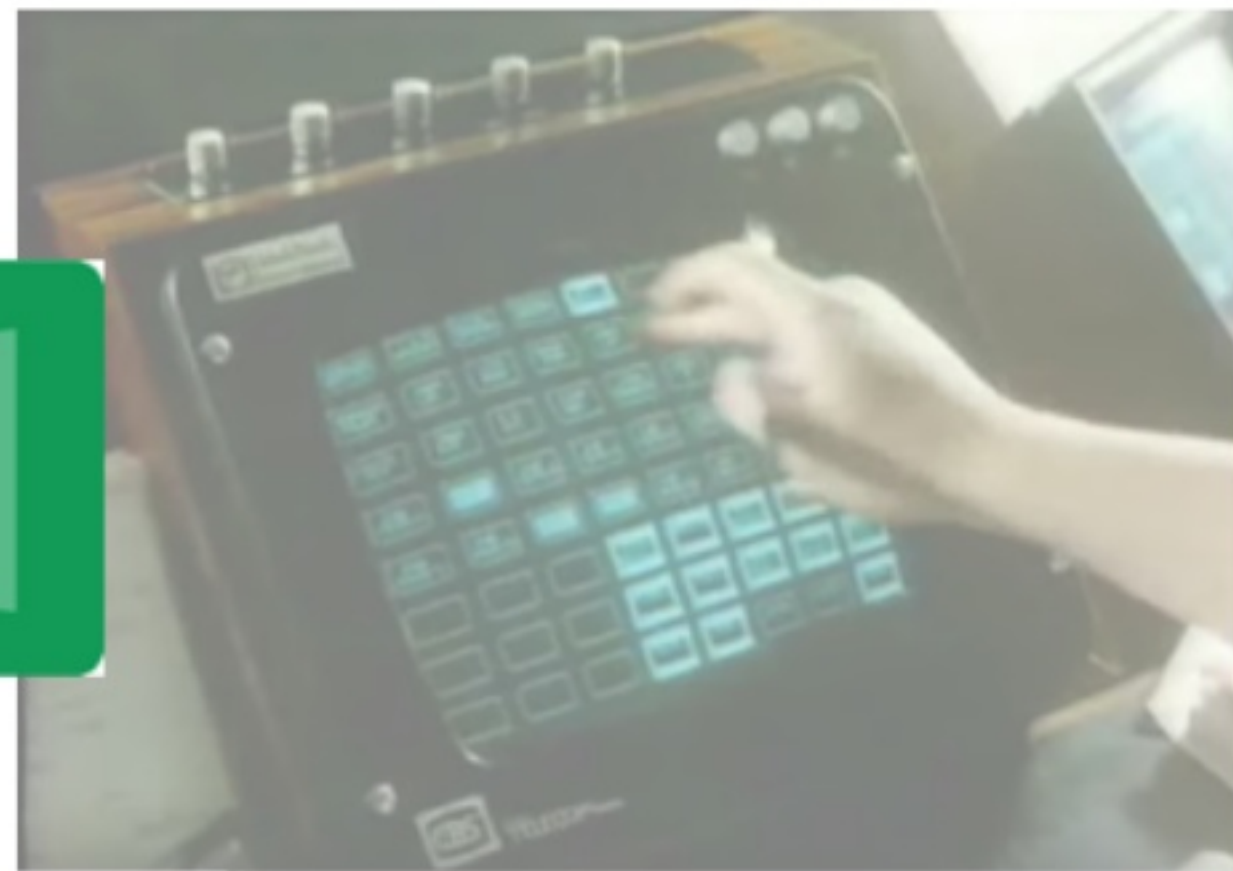
Spreadsheets!



<http://bit.ly/2e5GmyY>



<http://bit.ly/2e0TZA8>

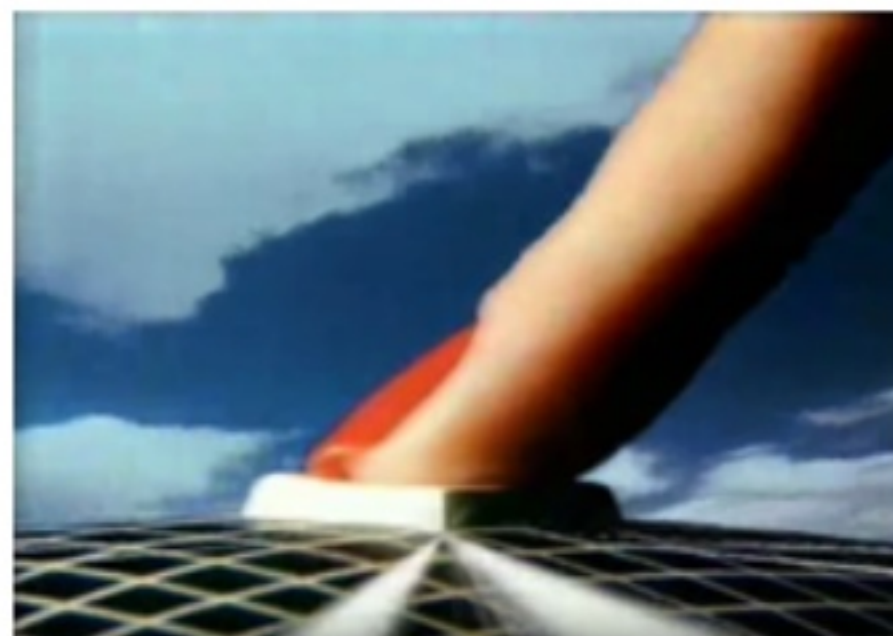


<http://bit.ly/2edYfMs>



Agenda

- Problem Statement and Motivation
 - Architecture
- Program Transformation
 - Pipeline
 - Code-to-Code Transformation
- Code Generation
 - Abstract Tree
 - Parse Tree
- Spreadsheets as a DSL
 - Generating Code
- Demo
- Q&A



Disclaimer(s)

- Ongoing research ...
- FP&A is one use case, but
Spreadsheets are much broader!



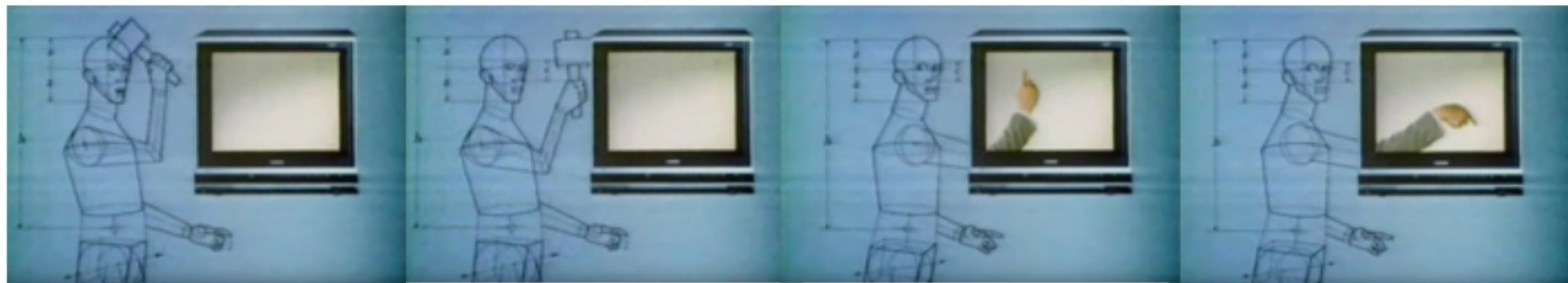
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- Ongoing research ...
- FP&A is one use case, but Spreadsheets are much broader!
 - E.g. People have even modeled Turing machines with Spreadsheets! [1]



Problem Statement

Prototype FP&A programs using Spreadsheet formulas and automatically translate to Scala / Spark.



Problem Statement

Prototype Any program using Spreadsheet formulas and automatically translate to Scala / Spark.



Motivation

- At Spark Summit Europe 2016 I presented the **Sparksheet** code generator for Spreadsheet formulas.
- Initially **Sparksheet** supported only 5 Spreadsheet formulas, now it supports 150+ Spreadsheet formulas!
- Motivation is finding use cases.

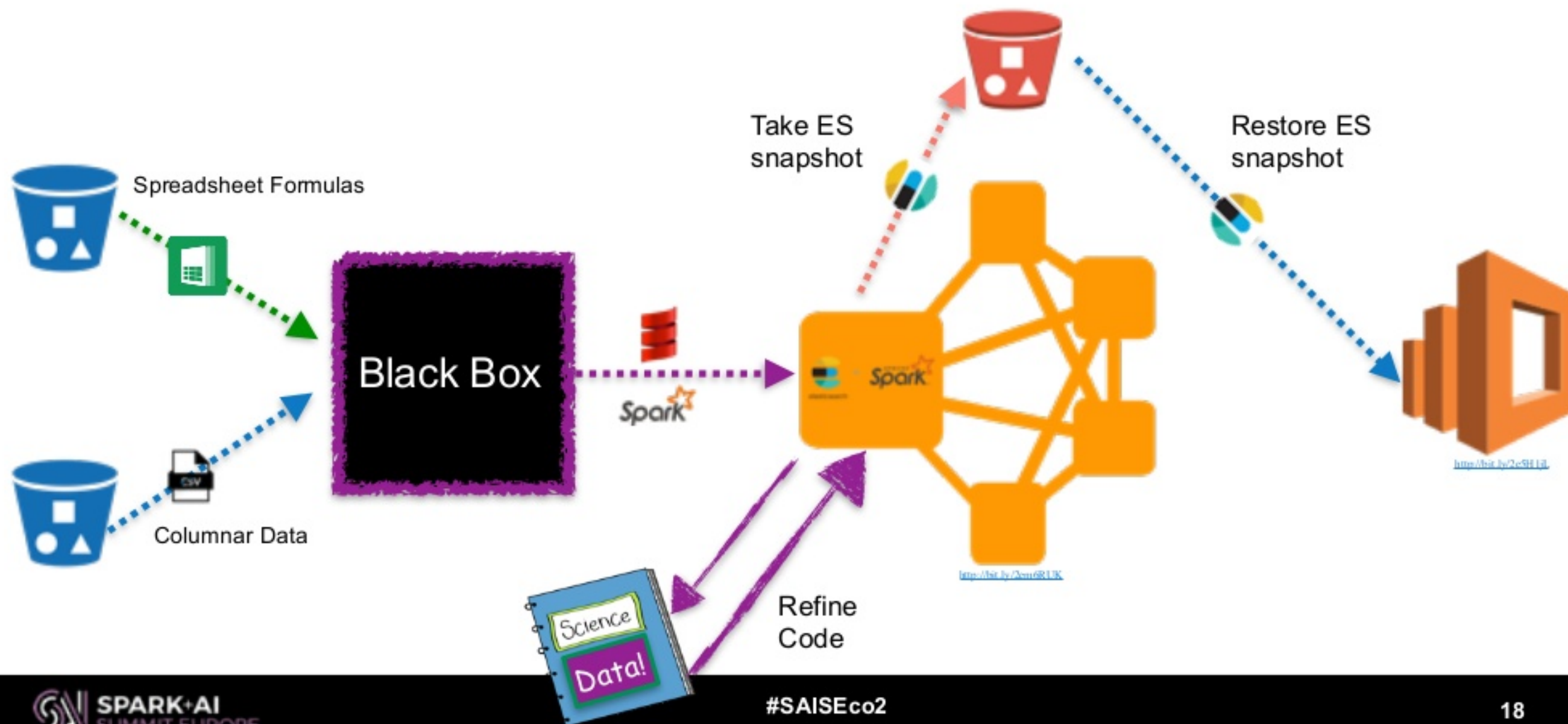
Motivation

**Automatically translate
Spreadsheet datasets* to
Spark data pipelines on
Scala/Spark**

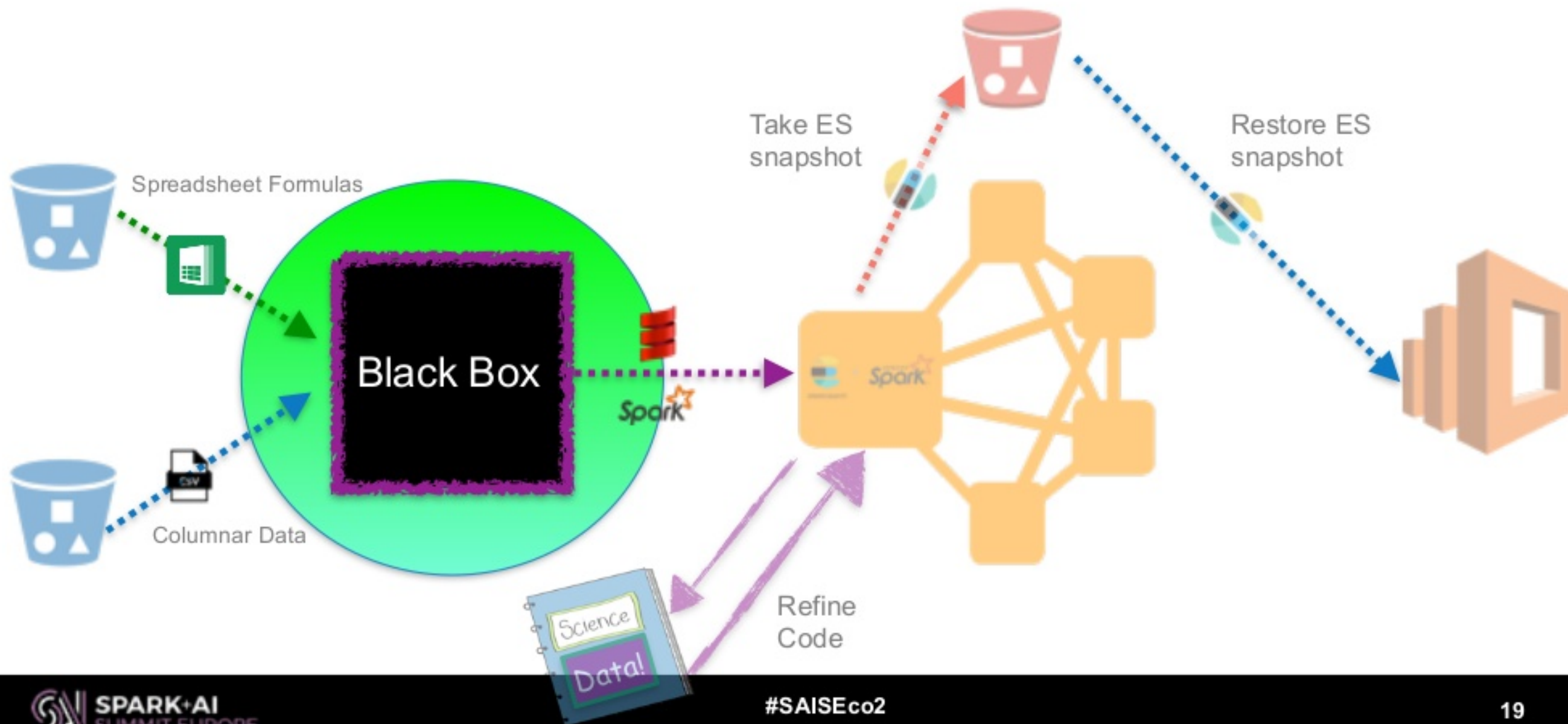
*Spreadsheet dataset = Structured data + Spreadsheet formulas



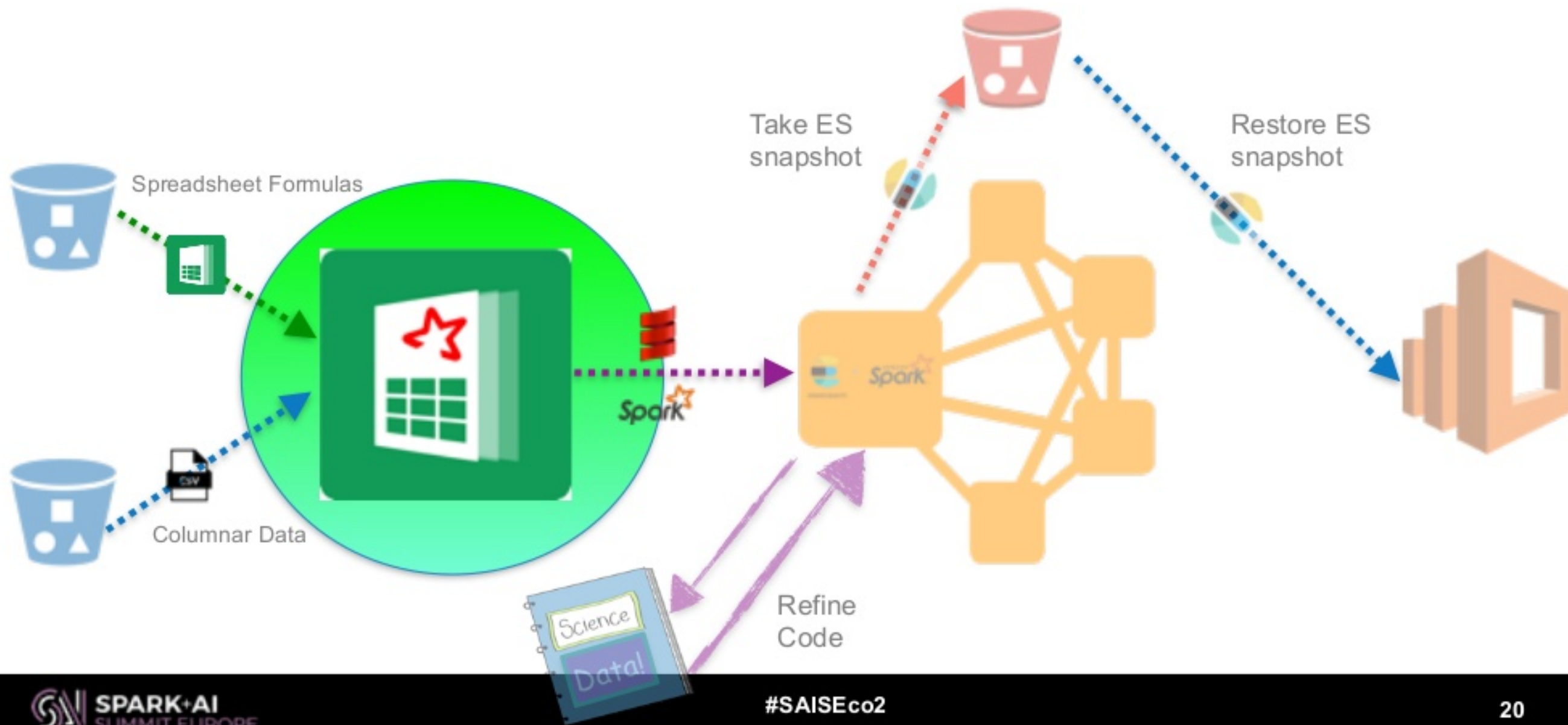
Architecture



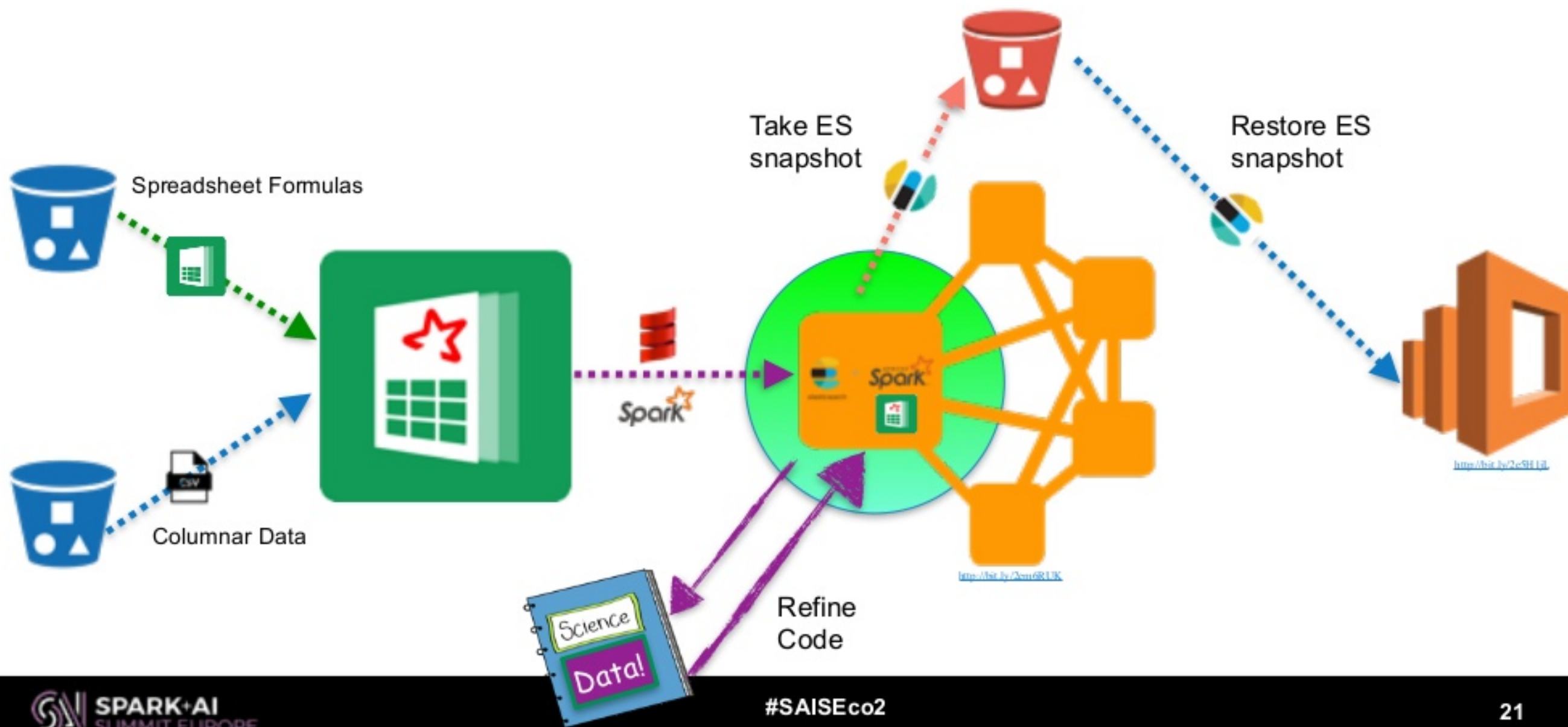
Architecture



Architecture



Architecture



Program Transformation

“A **program transformation** is any operation that takes a **computer program** and generates another program.”

https://en.wikipedia.org/wiki/Program_transformation

Program Transformation Pipeline



<http://bit.ly/2e0TZA8>



<http://bit.ly/2e0TZA8>



<http://bit.ly/2d0cFg>

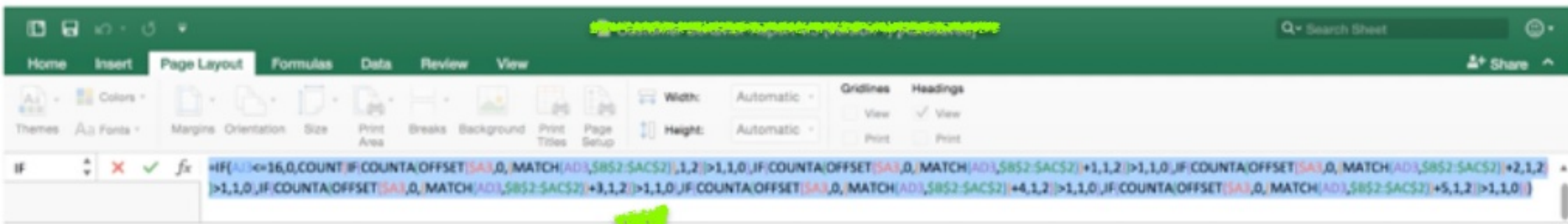
Demo #1



<http://bit.ly/2e0TZA8>

1. Show Spreadsheet model
2. Show Complex Spreadsheet Formula

Program Transformation

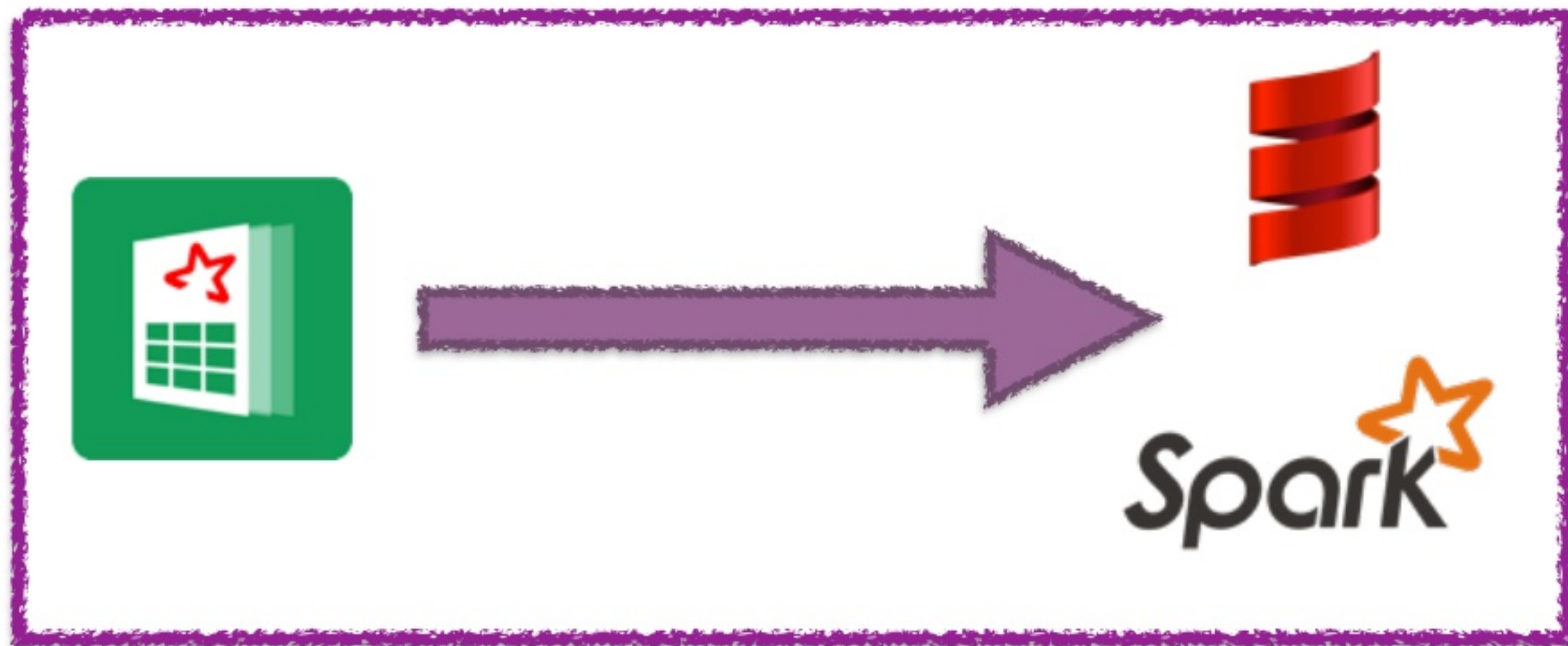


Source program in Excel formula language.

Demo #2



Demo #2



Demo #2

Sparksheet



Complex

Spreadsheet Formula



Spark

SparkSheet Demo

SparkSheet - Databricks

localhost:39901

SparkSheet

Input Excel Formula

```
IF(A23<=16,0,COUNT(IF(COUNTA(OFFSET($A3,0,(MATCH(AD3,$B$2:$AC$2),1,2))>1,1,0),IF(COUNTA(OFFSET($A3,0,(MATCH(AD3,$B$2:$AC$2))+1,1,2))>1,1
```

Convert to Scala

1.2.0

Demo

Output Spark App in Scala

```
import org.apache.spark.sql._
import org.apache.spark.sql.functions._
import scala.util.Try

//*****
//The following code was generated for this spreadsheet formula:
//*****

val ifColumn22DF = ifColumn1DFname.coalesce(1), "1", "lit(A23<=16)", "lit(0)",
"if_22", "COLUMN", "input_ifColumn22", "output_ifColumn22")
val if_22 = countColumn4DFname.coalesce(1), "1", "count_4", "count_5", "count_6",
"count_7", "count_8", "count_9", "input_countColumn4", "output_countColumn4")
val count_4 = ifColumn1DFname.coalesce(1), "1", "if_23>1", "lit(1)", "lit(0)",
"COLUMN", "input_ifColumn23", "output_ifColumn23")
val if_23 = countColumn19(COUNTA.coalesce(1), "1", "counta_19",
"input_countaColumn19", "output_countaColumn19")
val counta_19 = offsetColumn19(Dfname.coalesce(1), "1", A3, "lit(0)", "offset_19",
"lit(1)", "lit(2)", "input_offsetColumn19", "output_offsetColumn19")
val offset_19 = matchColumn19(Dfname.coalesce(1), "1", "AD3", "B2", "AC2", "COLUMN",
"input_matchColumn19", "output_matchColumn19")

val count_4 = ifColumn1DFname.coalesce(1), "1", "if_24>1", "lit(1)", "lit(0)",
"COLUMN", "input_ifColumn24", "output_ifColumn24")
val if_24 = countColumn20(COUNTA.coalesce(1), "1", "counta_20",
"input_countaColumn20", "output_countaColumn20")
val counta_20 = offsetColumn20(Dfname.coalesce(1), "1", A3, "lit(0)", "offset_20",
"lit(1)", "lit(2)", "input_offsetColumn20", "output_offsetColumn20")
val offset_20 = matchColumn20(Dfname.coalesce(1), "1", "AD3", "B2", "AC2", "COLUMN",
"input_matchColumn20", "output_matchColumn20")

val count_4 = ifColumn1DFname.coalesce(1), "1", "if_25>1", "lit(1)", "lit(0)",
"COLUMN", "input_ifColumn25", "output_ifColumn25")
val if_25 = countColumn21(COUNTA.coalesce(1), "1", "counta_21",
"input_countaColumn21", "output_countaColumn21")
```

Code-to-Code Transformation



Code-to-Code Transformation

“The input to the code generator typically consists of a parse tree or an abstract syntax tree.”



[https://en.wikipedia.org/wiki/Code_generation_\(compiler\)](https://en.wikipedia.org/wiki/Code_generation_(compiler))

Generating Code

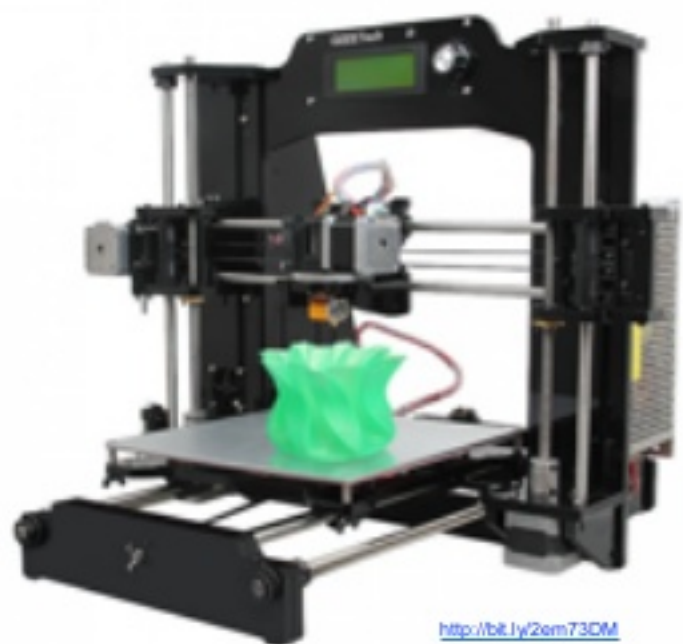
*“An elegant way to **generate code from an AST** is to write a class for each non-terminal node in the tree, and then each node in the tree simply generates the piece of code that it is responsible for.”*



<http://www.codeproject.com/Articles/26975/Writing-Your-First-Domain-Specific-Language-Part>

Generating Code

A **practical** way to **generate code** is to take a Parse Tree and write a **pretty printer** for the target language.



<http://bit.ly/2em73DM>

Generating Code (Example)

SUM(A,C)



Generating Code (Example)



SUM(A,C)

```
> import org.apache.sql._
import org.apache.spark.sql.functions._

def SUM(A:Int, C:Int): Int = {
  return A + C
}

val applySUM = udf(SUM _)
val sumDF = baseDF.withColumn("",
  applySUM(
    col("A"),
    col("C")
  ))
```



Generating Code (Example)



SUM(A,C)

```
> import org.apache.sql._
import org.apache.spark.sql.functions._

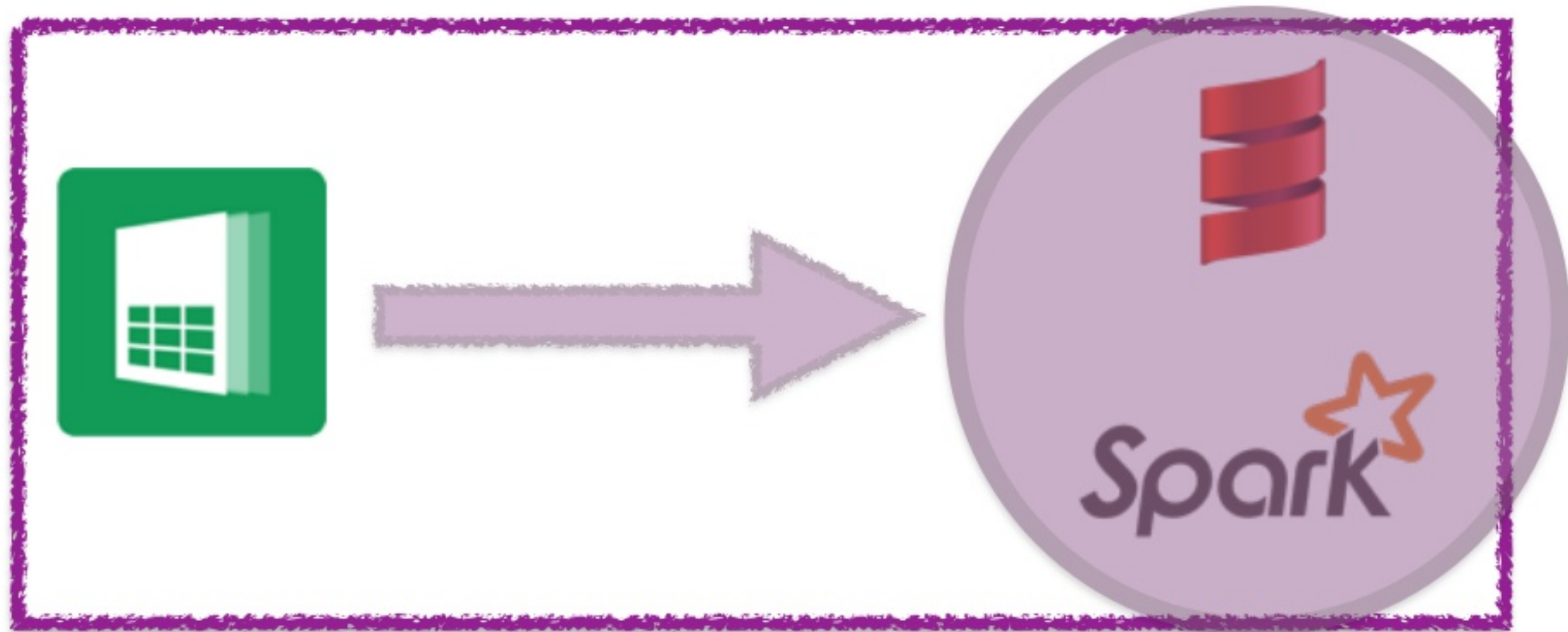
def SUM A: Int, C: Int Int = {
  return A + C
}

val applySUM = useDF.writes { df => {
  applySUM(df, Column("A"), Column("C"))
}}
```

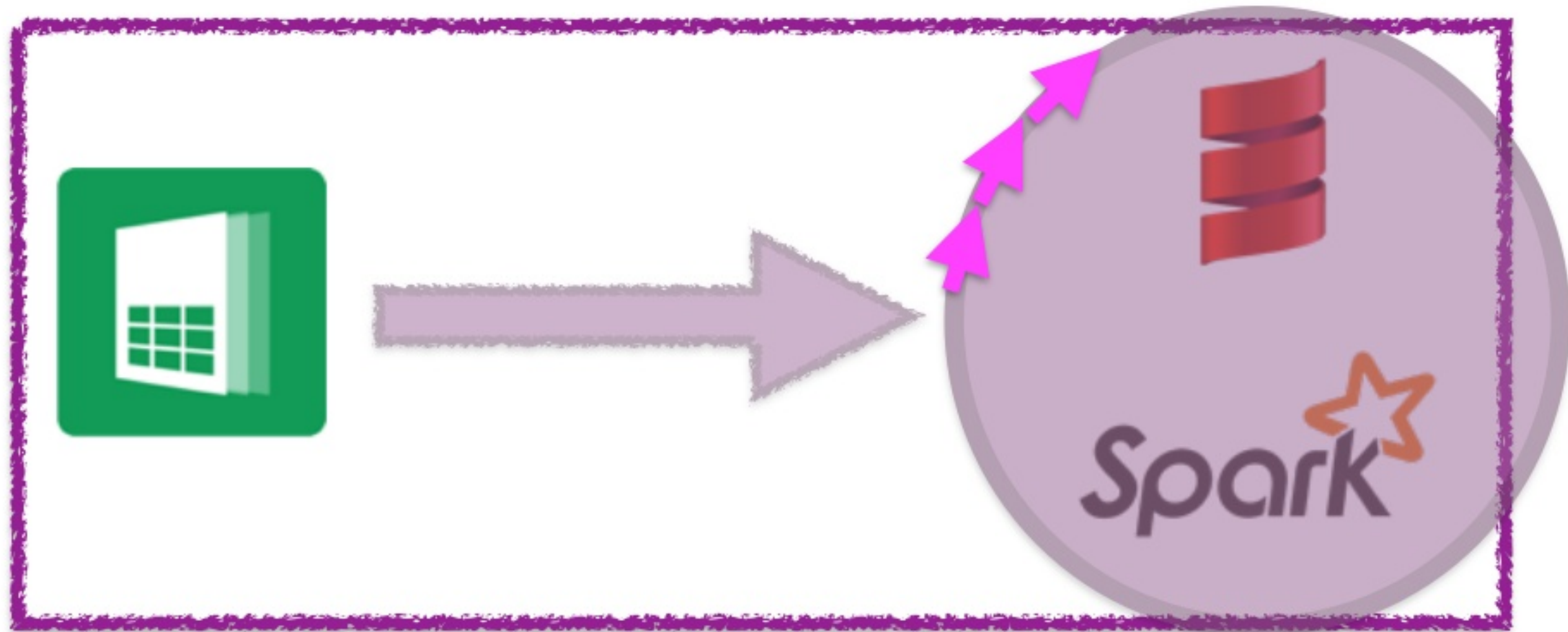
The diagram shows a code editor window with Scala code. The code defines a function `SUM` that takes two integers `A` and `C` and returns their sum. The function is then used to create a `UDF` (User Defined Function) named `applySUM`. The `applySUM` function is used to apply the `SUM` function to the columns `A` and `C` of a DataFrame. The code is written in Scala and uses the `org.apache.sql` and `org.apache.spark.sql.functions` packages.



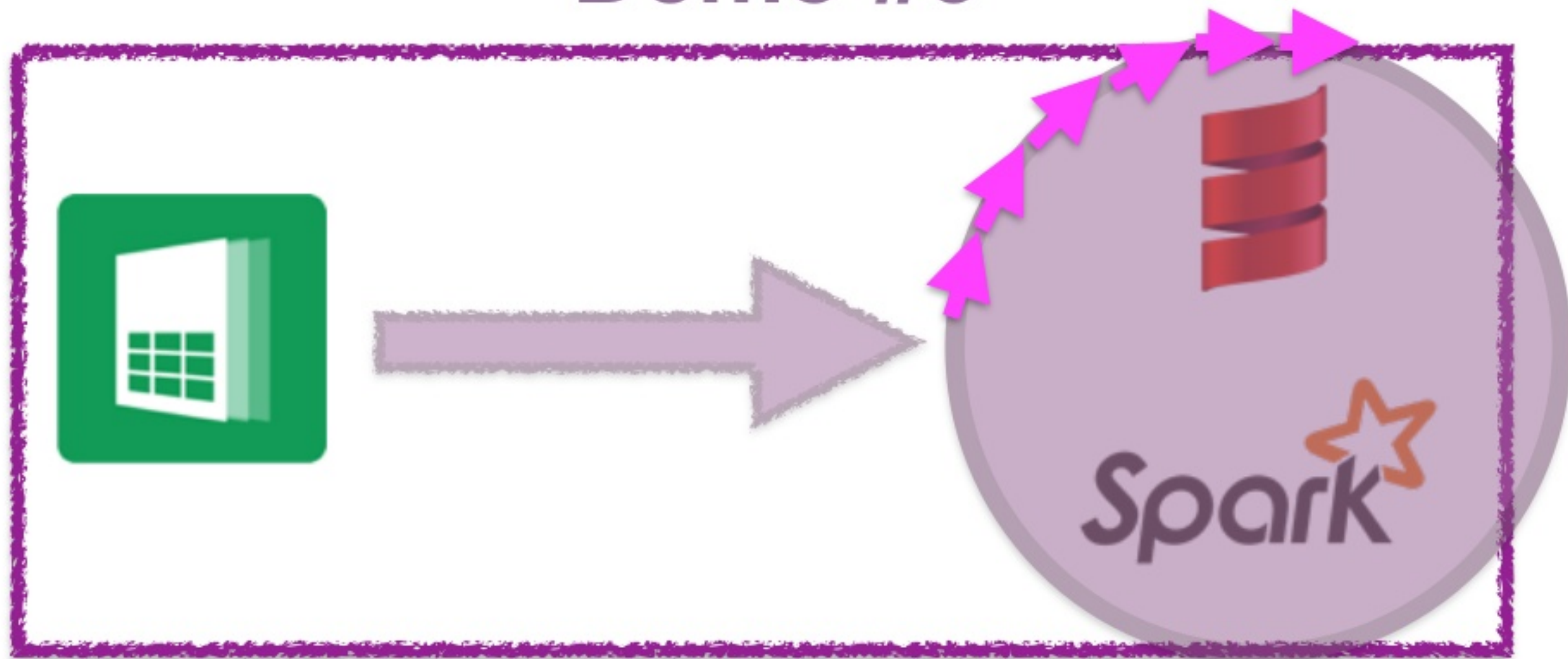
Demo #3



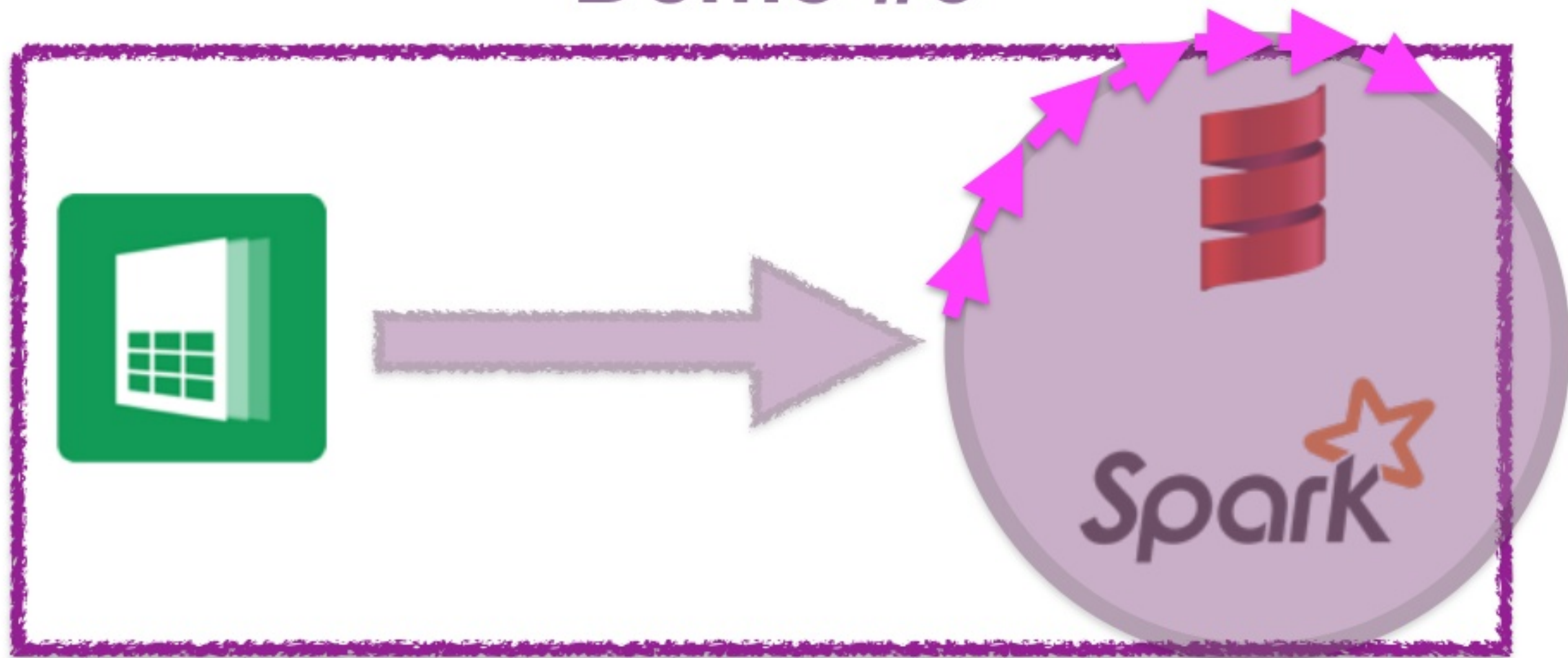
Demo #3



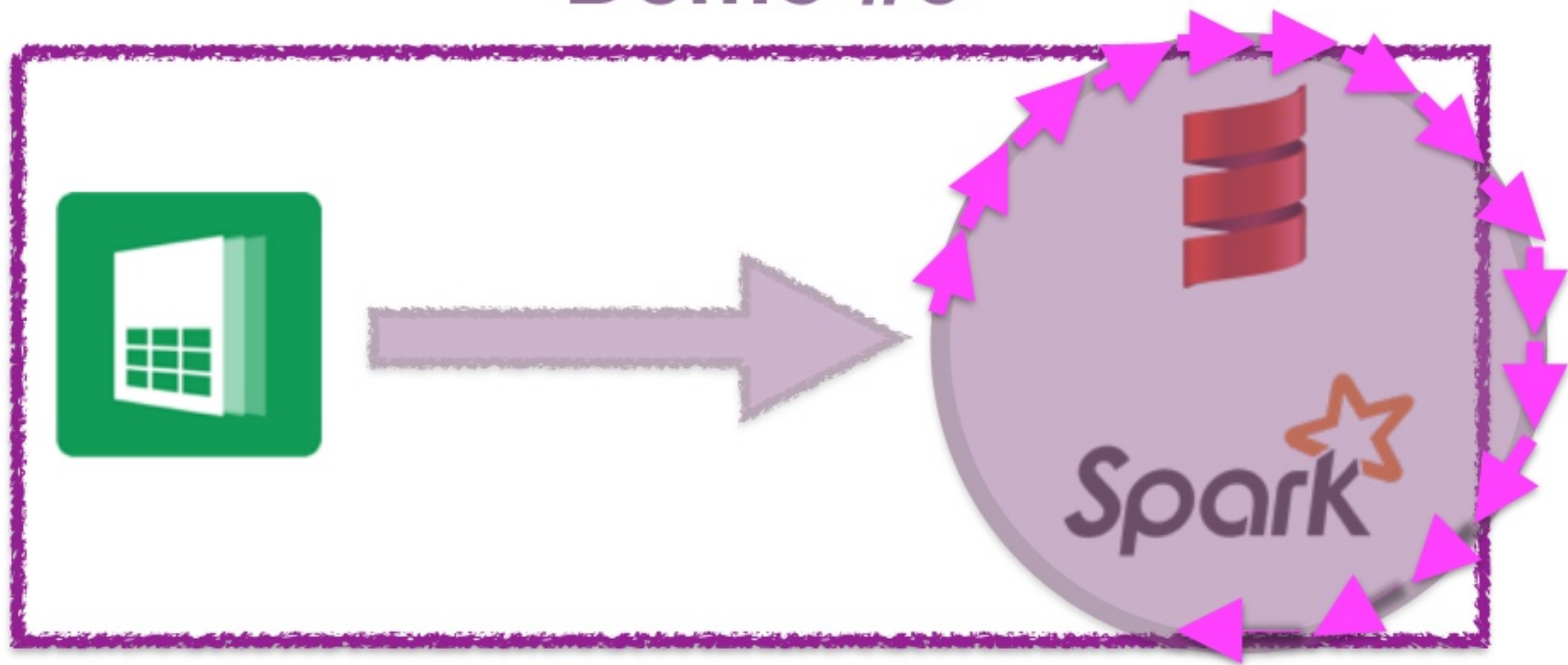
Demo #3



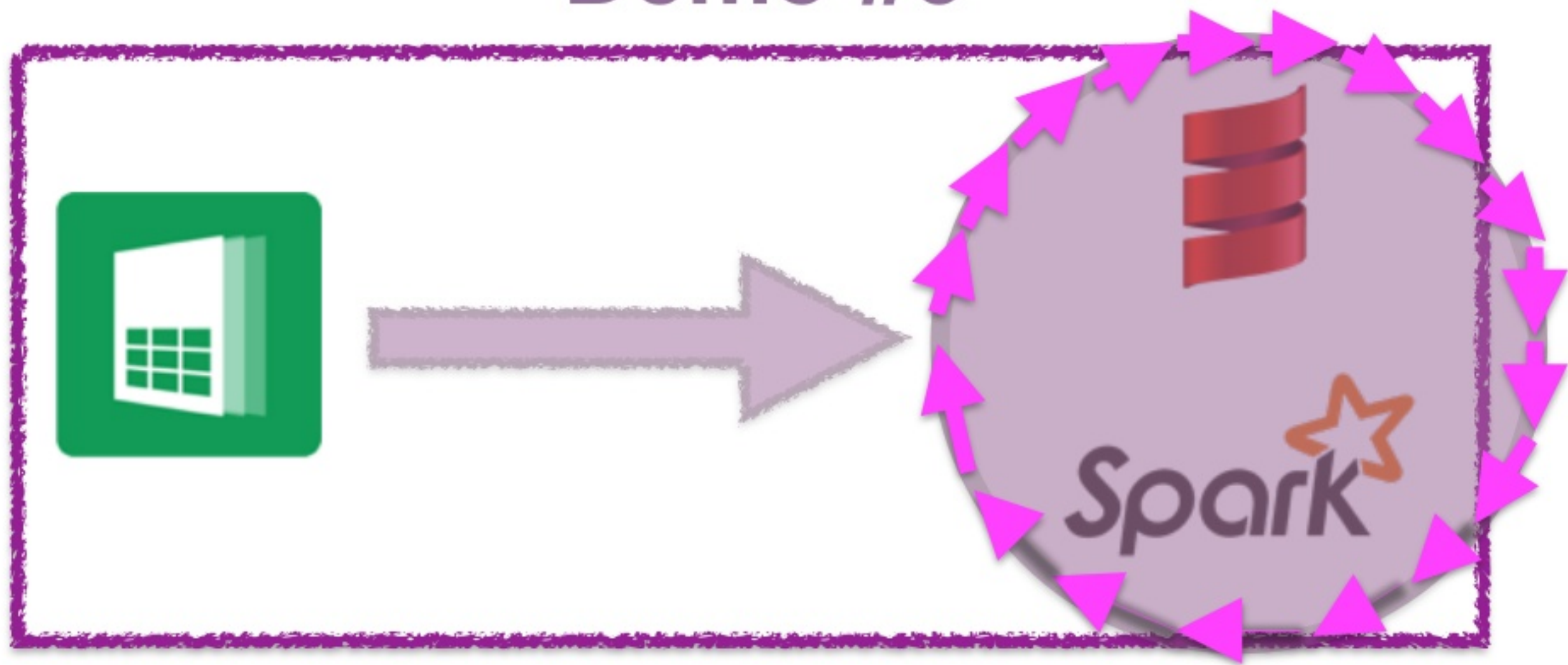
Demo #3



Demo #3



Demo #3



Demo #3

Executors

Show Additional Metrics

Summary

| | RDD Blocks | Storage Memory | Disk Used | Cores | Active Tasks | Failed Tasks | Complete Tasks | Total Tasks | Task Time (GC Time) | Input | Shuffle Read | Shuffle Write | Blacklisted |
|------------|---------------|-------------------------|--------------|-------|-----------------|-----------------|-------------------|----------------|---------------------------|-----------|-----------------|------------------|-------------|
| Active(18) | 125 | 2.3 MB / 146.1 GB | 0.0 B | 48 | 48 | 0 | 33975 | 34023 | 2.9 h (0.2 min) | 4.9 GB | 0.0 B | 0.0 B | 0 |
| Dead(0) | 0 | 0.0 B / 0.0 B | 0.0 B | 0 | 0 | 0 | 0 | 0 | 0 ms (0 ms) | 0.0 B | 0.0 B | 0.0 B | 0 |
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| Active(18) | 125 | 2.3 MB / 146.1 GB | 0.0 B | 48 | 31 | 0 | 33996 | 34027 | 3.1 h (0.3 min) | 5 GB | 0.0 B | 0.0 B | 0 |
| Dead(0) | 0 | 0.0 B / 0.0 B | 0.0 B | 0 | 0 | 0 | 0 | 0 | 0 ms (0 ms) | 0.0 B | 0.0 B | 0.0 B | 0 |
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| Active(18) | 125 | 2.3 MB / 146.1 GB | 0.0 B | 48 | 4 | 0 | 34023 | 34027 | 3.4 h (1.4 min) | 5.1 GB | 0.0 B | 0.0 B | 0 |
| Dead(0) | 0 | 0.0 B / 0.0 B | 0.0 B | 0 | 0 | 0 | 0 | 0 | 0 ms (0 ms) | 0.0 B | 0.0 B | 0.0 B | 0 |
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Spreadsheets as a DSL

- Spreadsheet is a powerful data modeling tool.
- Start simple and evolve into a complex ML pipeline.
- Spreadsheets are suitable to many domains (FP&A is one such domain).



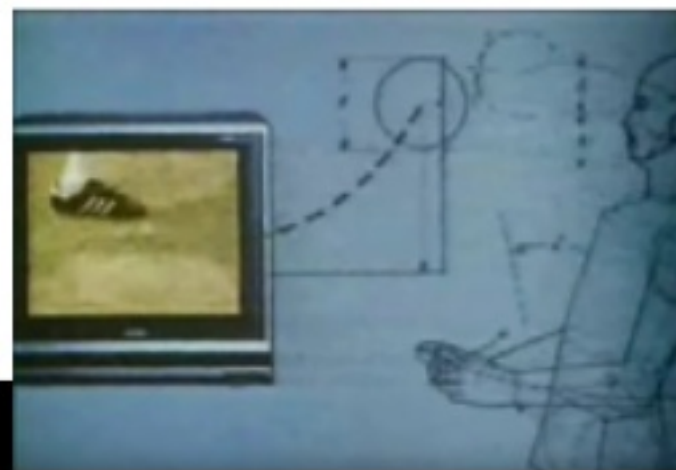
What have we seen?

- Spreadsheet applications as Prototypes for Spark programs
- Program Transformation
 - How to model as Pipeline
 - Why considered Code-to-Code Transformation
- How to Generate Code
 - AST (elegant)
 - Parse Tree (practical)
- Spreadsheets as a DSL
 - Generating Code
- Next Steps



Next Steps

- Use cases!
- Modeling Machine Learning in a Spreadsheet
- Prototype D|'s and ML|'s in a Spreadsheet



References

- A Grammar for Spreadsheet Formulas Evaluated on Two Large Datasets – Efthimia Aivaloglou, David Hoepelman & Felienne Hermans, Proceedings of SCAM '15
- <http://www.felienne.com/archives/2974>
- Pictures in presentation from Boards of Canada video “roygbiv”
<https://youtu.be/yT0gRc2c2wQ>

Q&A

THANK YOU.

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