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Oh yeah, there's also the massive shift in industry to data-oriented decision making too!

...and many applications are "data science in the large".

By using a language like Scala, it's easier to scale your small problem to the large with Spark, whose API is almost 1-to-1 with Scala's collections.

That is, by working in Scala, in a functional style, you can quickly scale your problem from one node to tens, hundreds, or even thousands by leveraging Spark, successful and performant large-scale data processing framework which looks a and feels a lot like Scala Collections!



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- Good for data science. Not just because of performance, but because it enables iteration, which is required by most algorithms in a data scientist's toolbox.



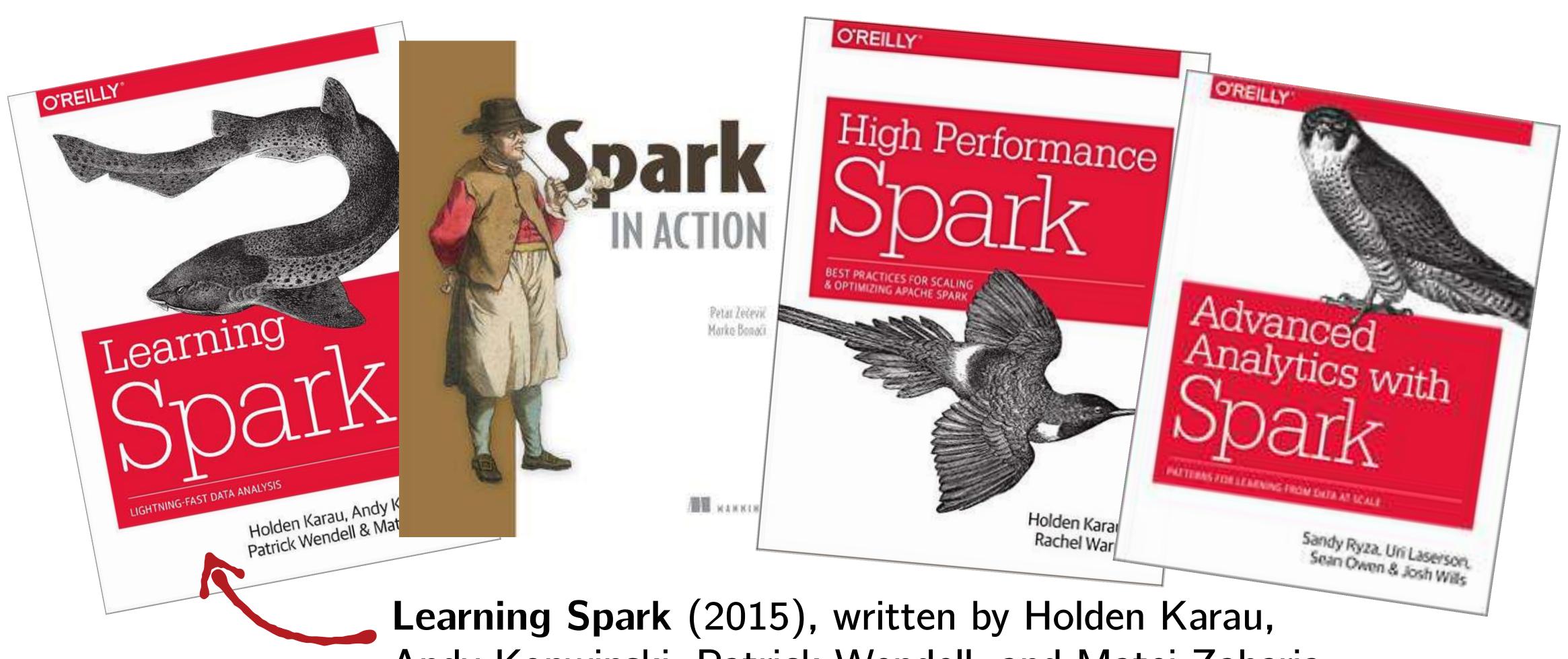
# Also good to know...

Spark and Scala skills are in extremely high demand!

### In this course you'll learn...

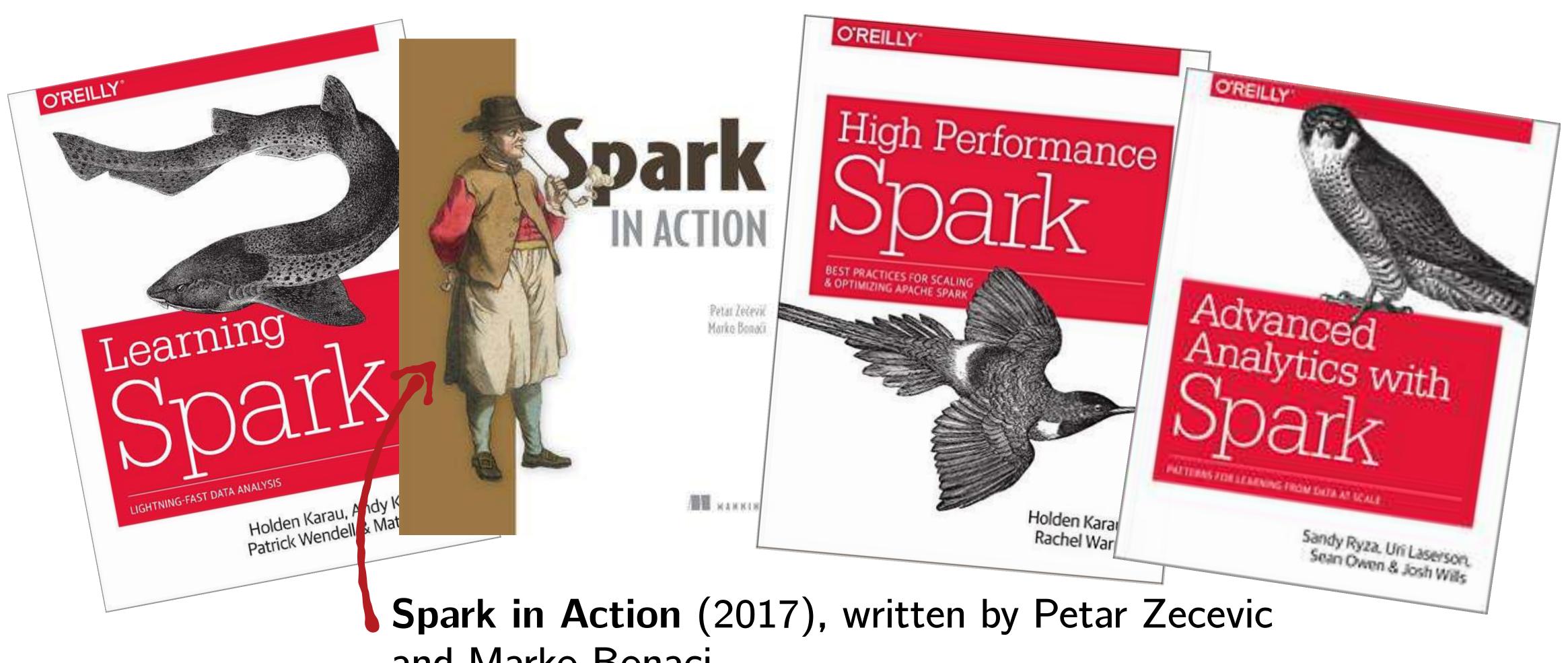
- Extending data parallel paradigm to the distributed case, using Spark.
- Spark's programming model
- Distributing computation, and cluster topology in Spark
- How to improve performance; data locality, how to avoid recomputation and shuffles in Spark.
- Relational operations with DataFrames and Datasets

### Many excellent books released in the past year or two!



Andy Konwinski, Patrick Wendell, and Matei Zaharia

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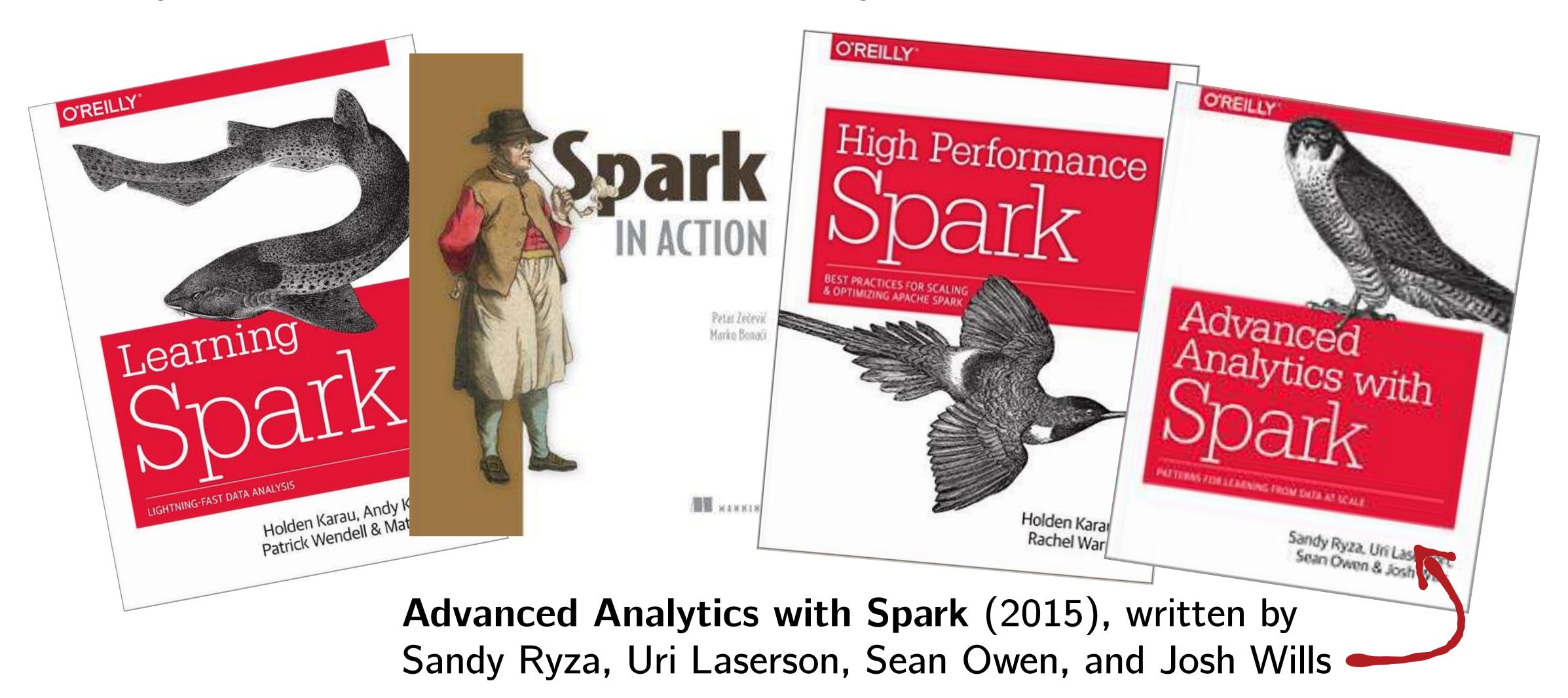
and Marko Bonaci

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High Performance Spark (2017), written by Holden Karau and Rachel Warren

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### **Mastering Apache Spark 2**

Welcome to Mastering Apache Spark 2 (aka #SparkLikePro)!

I'm Jacek Laskowski, an **independent consultant** who is passionate about **Apache Spark**, Apache Kafka, Scala, sbt (with some flavour of Apache Mesos, Hadoop YARN, and DC/OS). I lead Warsaw Scala Enthusiasts and Warsaw Spark meetups in Warsaw, Poland.

#### Tools

- IDE of your choice
- sbt
- Databricks Community Edition (optional)

Free hosted in-browser Spark notebook. Spark "cluster" managed by Databricks so you don't have to worry about it. 6GB of memory for you to experiment with.