

# DATA PRODUCT SUCCESS

A scenic view of snow-capped mountains and a body of water with icebergs. The mountains are covered in white snow, with some dark rock faces visible. The water is dark blue, and several icebergs of various sizes are floating in the foreground. The sky is a clear, pale blue with some light clouds.

WITH RICK SAPORTA

@ricksaporta

Aligning with data's core purpose to  
build data products that customers love

# What does it take to create data products that customers love?

1. THE RIGHT FOCUS
2. WELL-EXECUTED HANDOFFS



# GOAL OF THIS TALK

## BY THE END OF THIS (ABBREVIATED) TALK

- Framework for identifying what to build / prioritize
- Tips to help you achieve customer love with your data products

## THIS TALK IS FOR YOU IF

- You or your company build data products
- You or future-you work extensively with data
- If this is not you, *I'm curious about your curiosity. Drop me a line!*

# AGENDA

1. Intro
2. Clarifying of terms
3. Three core problems addressed by data products
4. A framework for deciding what to build
5. How to ensure the success of our data products

# GOAL OF THIS TALK

## BY THE END OF THIS (ABBREVIATED) TALK

- Clear model for building a data product
- Framework for how to build / prioritize
- Tips to help with the process

**full version of this framework:**  
**[RickSaporta.com/DataProduct](https://RickSaporta.com/DataProduct)**

## THIS TALK IS FOR YOU IF

- You or your company build data products
- You work extensively with data
- You're data curious
- If this is not you, *Hi! Thank you for being here!*

# Hi, I am Rick 🖐️

Also go by: **Ricky**  
Pronouns: **He/Him**

Currently: **SVP, Data** at [Entera.AI](#)

**Working in data for nearly two decades**  
**Building & leading data teams for the past 10 years**

- Head of Data Strategy & Insights @ The Farmer's Dog
- VP of Data @ Vydia (acquired by gamma.)
- The Orchard (acquired by Sony)
- NBS (acquired by Pandora)
- Phocas BI
- AIG
- Other startups

# CLARIFYING TERMS

Words get thrown around. Let's make sure we all mean the same thing

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# CONFLATABLE



turtle

java

product management

data product management



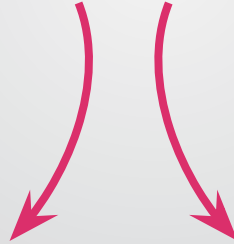
tortoise

javascript

project management

managing data as a product

Very different  
concepts





**Who are the "customers"  
of data products?**

What does it mean to  
*love* a data product?

# What is data's purpose?

# How does data impact outcomes?

data →



→ outcomes

# How does data impact outcomes?

by enabling decisive action

data → decisions → actions → outcomes

# Data impacts outcomes through:

1. better decision-quality
2. faster time to decisive action

Data does not *produce* outcomes

Someone (or thing) must take  
that decisive action

*Stating the obvious.  
Will become important later.*

# DATA PRODUCTS

Data products generally address one of **three core problems**

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## Problem #3

**OUR DATA NEEDS TO BE  
ACCESSED AND MANAGED**

## Problem #2

**DATA WE NEED  
DOES NOT EXIST**

# Problem #1

**DECISIONS ARE  
HARD TO MAKE**

## Data Products solve one of three problems:

1. **Decisions** are hard to make
2. Data we need **does not exist**
3. Our data needs to be **accessed and managed**

# A FRAMEWORK FOR DECIDING WHAT TO BUILD

Infinite swiss army knives provide  
infinite possibilities that lead to dead ends.

# 4

The most important consideration

**Focus on the decisions  
your customers make**

The most important consideration

**Focus on the decisions  
your customers make**

**Yes, even Data Engineering!**

## Summary

# A Framework for deciding which data products to build

1. **Identify the decisions**
2. Understand the decision's frequency
3. Understand the influence of the data relative to the decision
4. Filter the decisions to ONLY the sweet spot
5. **Identify opportunities for improving decision making**
6. **Identify the hand-offs.** Focus on the actions
7. Scope the work to determine effort
8. Quantify the value of the improvement
9. Prioritize the decisions (value x effort)

BONUS POINTS:

**What are your customers' customer's goals?**



# Summary

## A Framework for deciding which data products to build

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BONUS POINTS:

**What are your customers' customer's goals?**

# Identifying the decisions your customers make

1. Ask
2. Observe
3. Inbound

# What do decision points look like?

what are we looking for?

Find Inaction and you  
find decision points

# Problems present themselves as solutions

follow the thread back

# Identifying Opportunities for data-products

1. improve the **decision quality**
2. improve **time to decisive** action

# Questions that uncover opportunities for improving decision quality

*Weave the decisions to the outcomes*

1. **How do you measure the success of the decision?**
2. **What would a better decision look like?**
3. **What does an ideal decision look like?**

 *repeat multiple times*

4. **What is preventing a better decision from happening today?**
5. **What else?**

*they'll say they need more data. dig deeper.*

6. **What would having that data allow you to do that you cannot do today?**
7. **How, specifically, would that happen?**

*Explore for other paths to reach the outcome.*

8. **If non or your known options were available, what would you do?**

# Questions that uncover opportunities for improving decision quality

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*Explore for other paths to reach the outcome.*

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# Identify...

- ... the **decisions**



- ... the **opportunities**



- ... the **handoffs**



**The handoff is where  
data products go to die**

# Identify the hand-offs

Data does not produce outcomes

Someone (or thing) must take  
that decisive action

*Stating the obvious.  
Will become important later.*

# hand-offs

# Identify the hand-offs

How will your work get converted into action?

- ◇ Who **specifically** is going to act
- ◇ What do they **know**...
  - about this handoff
  - about how to use the data
- ◇ What do they **expect**
- ◇ What is **their timeline**

# Identify the hand-offs

How will your work get converted into action?

- ◇ Who **specifically** is going to do it
- ◇ What do they know...
  - about this handoff
  - about how to use it
- ◇ What do they expect
- ◇ What is **their timeline**

The more detail you gather here prior to building, the greater the likelihood of your data product succeeding

With data and a long enough timeline, we can accomplish nearly anything

# Now we can apply the infinite swiss army

**The challenge is to apply it thoughtfully and with purpose**

# Initiatives that can lead to improved decision quality

1. **gather data**
2. **create data (forecasting; segmentation)**
3. **integrate existing data sources**
4. **improve access**
5. **lower latency**
6. **derive insights**
7. **find new opportunities**
8. **improve interactions with the data**
9. **improve trust in the system**
10. **catalyze the action**



# ENSURING THE SUCCESS OF OUR DATA PRODUCTS

BUILDING DATA PRODUCTS THAT  
CUSTOMERS LOVE

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# Frontend data products have two design processes

not just processes, but skill sets

# Two distinct design processes

**Data Design** is to **Product Design**

as **Writing a Book** is to **Publishing a Book**

# Two distinct design processes

## Data Design

Telling the story  
Making a decision

## Product Design\*

Interacting with the story  
Taking an action

\* in the context of data products

**We are responsible for  
our work getting used**

**The biggest obstacle to  
successful data science  
is the sea of possibilities**

# Use “the twoday rule”

can we do it? → **what can we do in 2 days**

Do not allow more than 2 days for the first iteration of any new data science project

**Data is full of bias**  
**Use it responsibly**



# Why do data products fail?

1. The work was not sufficiently connected to our customers' decisions
2. The handoffs were not sufficiently understood and established

**Success**  $\propto$  **How specific the answers are:**

Decision

**What decisions**

Handoffs

**Who is going to take the handoff**

# Success $\infty$ How specific the answers are:

Decision

**What decisions, specifically, does our work enable? How do those decision(s) tie to our customers' goal(s)?**

Handoffs

**Who is going to take the handoff, specifically? Who will take a valuable, decisive action with it? What expectations, knowledge, skills do they have?**

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Thank You



[RickSaporta.com/DataProduct](https://RickSaporta.com/DataProduct)

🐦: @RickSaporta

in: /in/RickSaporta

Thank You

