

Are you ready for AI?

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Does AI have a massive future? Sure! Please insert another coin.

Do we (the builders) have a **clear** idea how to get there? Hmmmm.

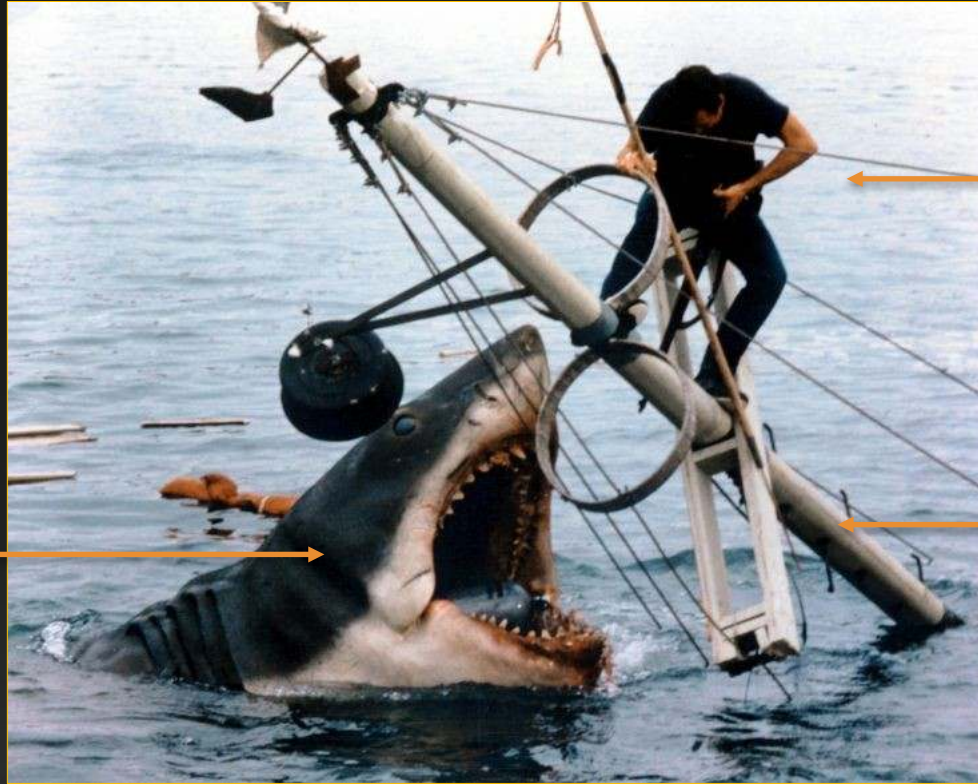
« If you want to know the future,
look at the past »»

Albert Einstein

What's our collective track record on understanding
and implementing disruptive technologies?

2000

Your
user
s



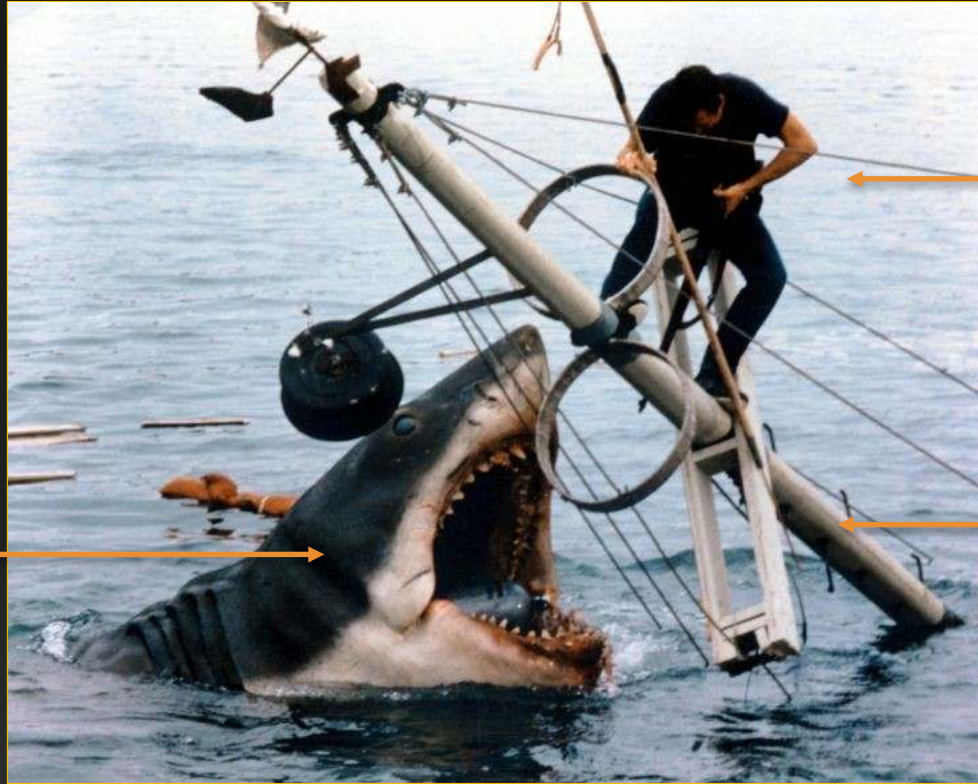
You

Your
Web
project

Universal
Pictures

2005

Your
user
s



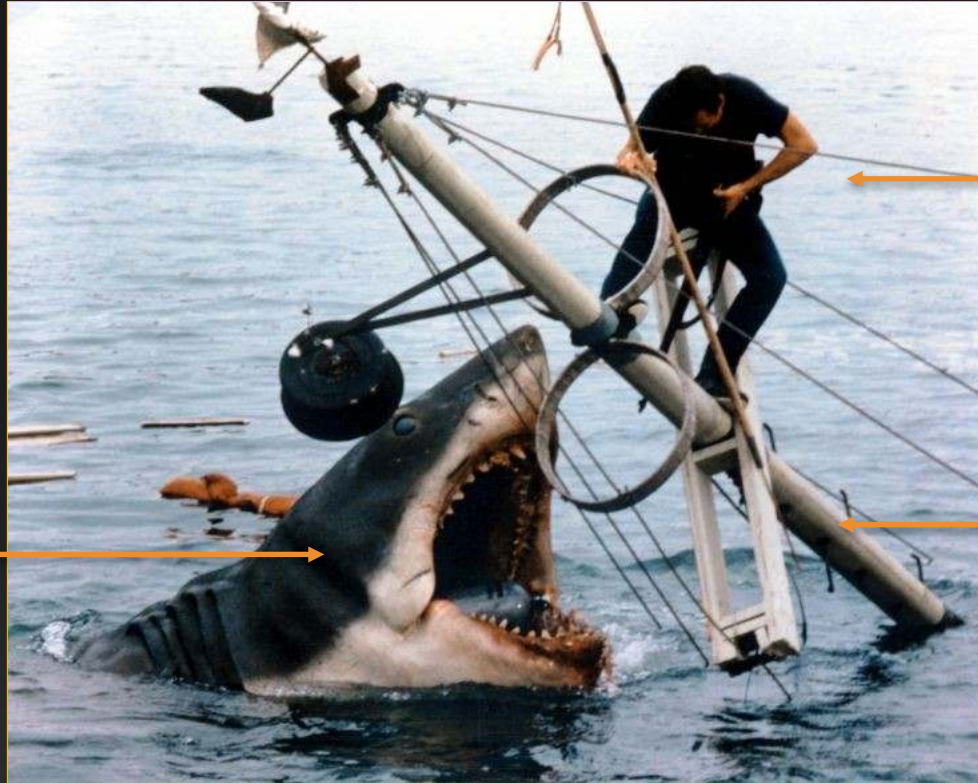
You

Your
E-commerce
project

Universal
Pictures

2010

Your
user
s



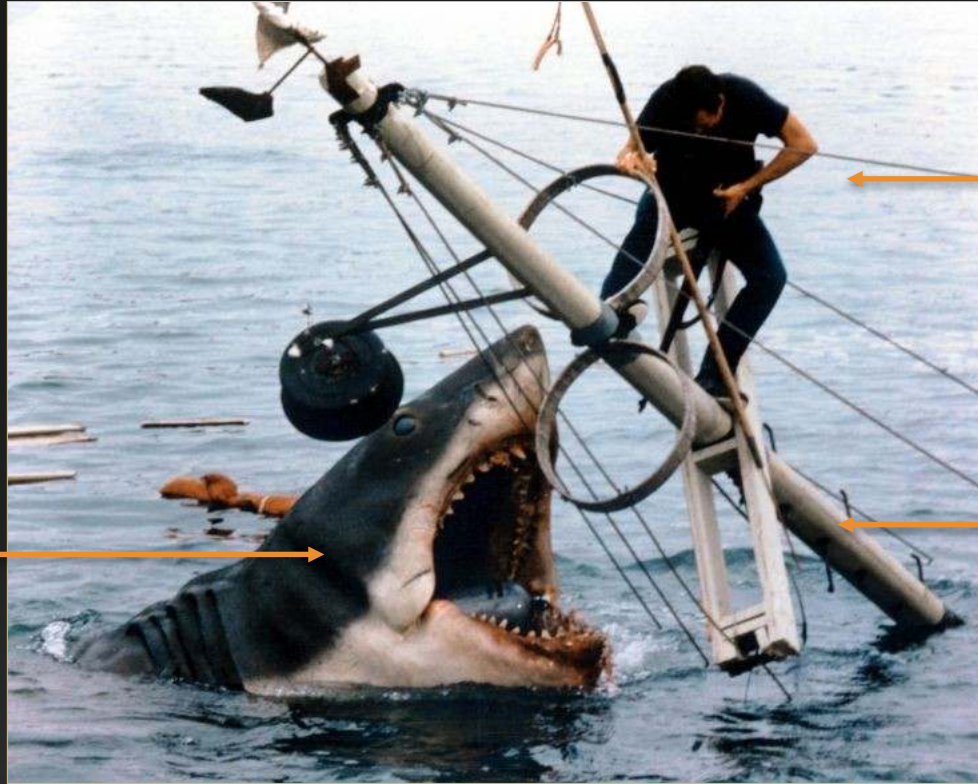
You

Your
Mobile
project

Universal
Pictures

2015

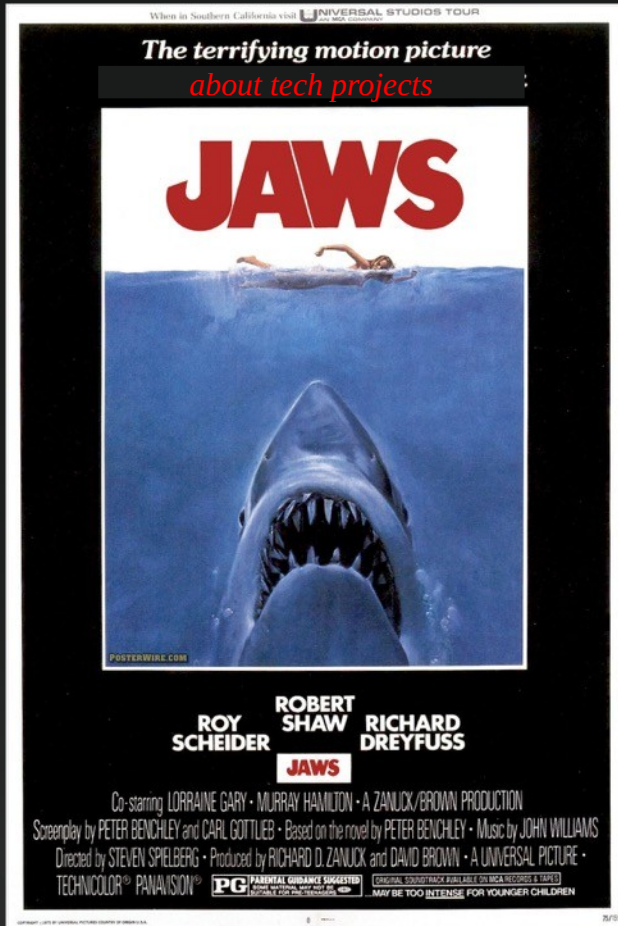
Your
user
s



You

Your
Big
Data
project

Universal
Pictures



The terrifying truth about tech projects

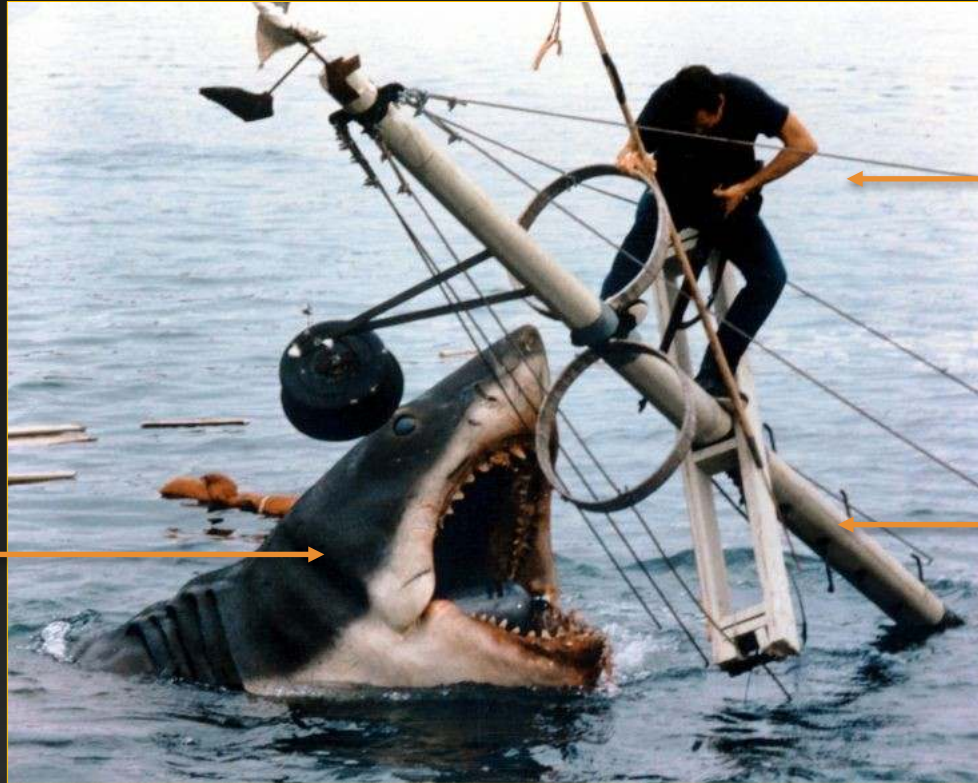
Delusional stakeholders
Business pressure
Unprepared team
Inadequate tools
Improvised tactics
Random acts of bravery

« It's different this time!
The AI revolution is here!
Blah blah blah »

You know who

2020

Your
user
s



You

Your
AI / ML
project

Universal
Pictures

« Insanity is doing the same thing
over and over again and
expecting different results »

Whoever said it first

Tired of being shark food?

Delusional stakeholders
Business pressure
Unprepared team
Inadequate tools
Improvised tactics
Random acts of bravery



Set expectations
Define clear metrics
Assess your skills
Pick the best tool for the job
Use best practices
Iterate, iterate , iterate

1 - Set expectations

- What is the **business question** you're trying to answer?
 - One sentence on the whiteboard
 - Must be **quantifiable**
- What **data** would a **human** look at to answer it?
 - Is that data available?
 - How much do you have?
 - How hard is it to collect more?
- Involve everyone and come to a **common** understanding
 - Execs, Business, IT, Data Engineering, Data Science, etc.

« We want to see what this technology can do for us »

« We have tons of relational data, surely we can do something with it »

« I read this cool article about FooBar ML, we ought to try it »



2 - Define clear metrics

- What is the **business metric** showing success?
- What's the **baseline** (human and IT)?
- What would be a **significant** and **reasonable** improvement?
- What would be **reasonable** further improvements?

« The confusion matrix for our IT ticket classifier looks much better ». **Huh?**

« P90 time-to-resolution is now under 24 hours ». **Err....**

« Misclassified emails have gone down 5.3% using the latest model ». **So?**

« The latest survey shows that 'very happy' **customers** are up 9.2% ». **Woohoo!**

3 - Assess your skills

- Can you **build a data set** describing the problem?
- Do you know how to **clean** and **curate** it?
- Can you write and tweak ML **algorithms**?
- Can you manage ML **infrastructure**?
- ... Or do you only want to call an **API** and get the job done?

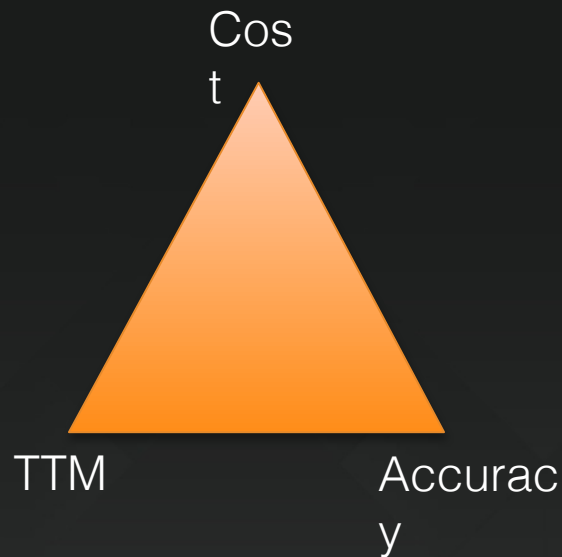
Fully
managed



100%
DIY

4 - Pick the best tool for the job

- Cost, time to market, accuracy: **pick two**
- The least expensive and fastest option won't probably be the most accurate.
 - Maybe enough to **get started**, and **learn** more about the problem.
- Improving accuracy will take **increasingly** more time and money.
 - Diminishing returns! Know when to stop.
- Keep an eye on **actionable** state of the art advances, ignore the rest
 - Transfer learning
 - AutoML



5 - Use best practices

- No, things are **not** different this time.
- AI / ML is **software engineering**
 - Dev, test, QA, documentation, Agile, versioning, etc.
 - Involve all teams
- Sandbox tests are nice, but truth is in **production**
 - Get there fast, as often as needed
 - CI / CD and automation are required
 - Devops for ML

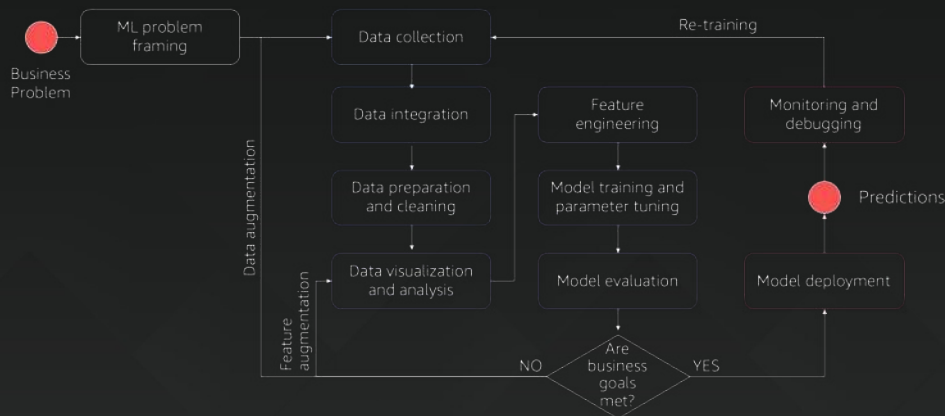


Universal
Pictures

6 - Iterate, iterate, iterate

aka Boyd's Law (1960)

- Start **small**
- Try the **simple** things first
- Go to production **quickly**
- Observe prediction **errors**
- Act: fix data set?
Add more data?
Tweak the algo?
Try another algo?
- Repeat until accuracy gains become **irrelevant**
- Move to the **next** project



Using Data in Education: Four Steps to Success

by AWS Public Sector Blog Team | on 21 FEB 2019 | in Education, Public Sector | Permalink | [Share](#)



Global education ministers gathered at the Amazon Web Service Institute's roundtable at Education World Forum to discuss using data to address major challenges in education.

<https://aws.amazon.com/blogs/publicsector/using-data-in-education-four-steps-to-success/>

« Does this work? »»

Everyone in this room

AI and Machine Learning on AWS

Tens of thousands of customers

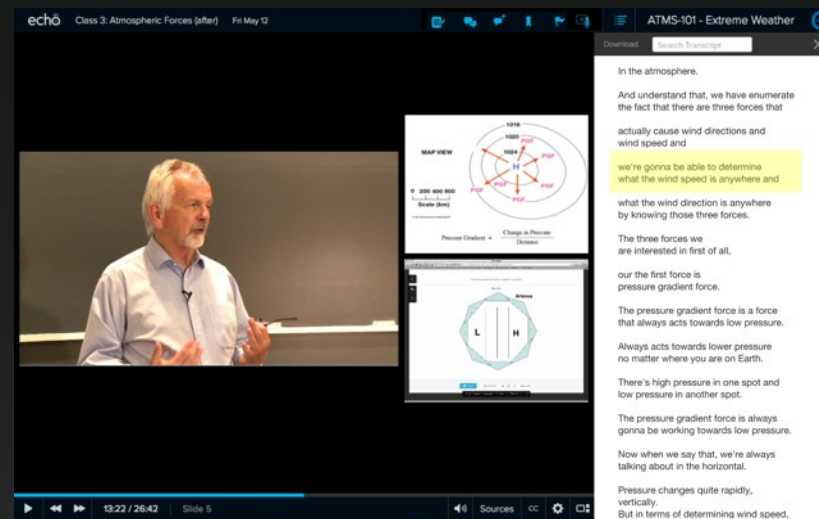


Automatic transcription of classes

<https://aws.amazon.com/solutions/case-studies/echo360/>



- **Amazon Transcribe**
 - Call an API, job done
 - 31 languages, 6 in real-time
- Students can focus on **listening** to and **engaging** with the lecture instead of trying to write down every word
- Students can later **review** accurate transcriptions, synchronized with slides and video thanks to timestamps
- Could easily be combined with **Amazon Translate**

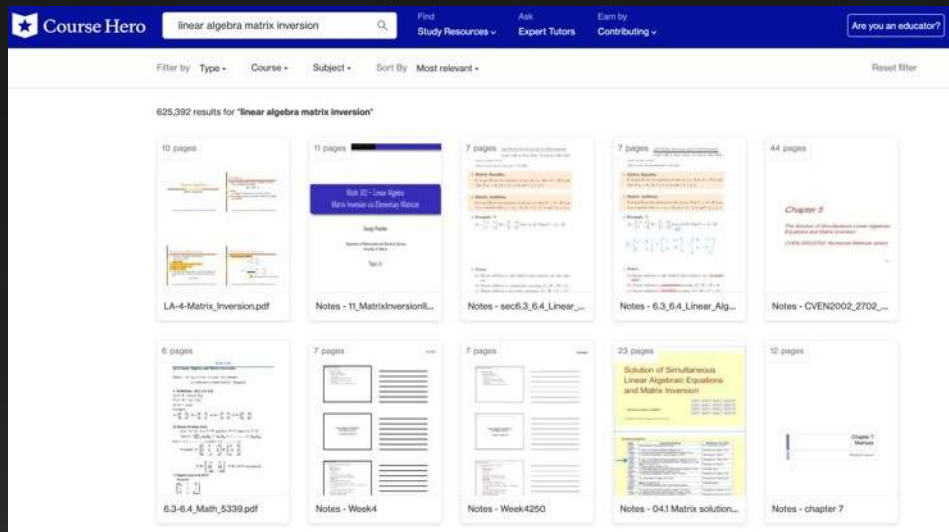


Personalizing studying

<https://aws.amazon.com/blogs/publicsector/personalizing-studying-with-machine-learning-course-heros-approach/>



- Helps student **gain mastery** or **dive deep** on over particular concepts
- 25 million course-specific materials
- Materials are checked and classified by machine learning jobs running on **Amazon EMR** and **Amazon SageMaker**



Predicting student success

<https://aws.amazon.com/solutions/case-studies/ivy-tech-community-college-of-indiana/>



- Identify **behaviors** of successful and unsuccessful students
- Predict with **80%** accuracy, which students are likely to fail a course within the first two weeks of a 16-week term
- Struggling faculty can be identified and addressed **before** they impact students
- Flagging fraudulent activity far **earlier** using Natural Language Understanding tools to analyze course evaluations

« We have an analysis kit we run every day, looking at data, comparing patterns over previous years' information, and in a matter of seconds, we can tell if a student is likely to succeed or fail. The results have been phenomenal. »

Lige Hensley, CTO, Ivy Tech

<https://ml.aws>

<https://aws.amazon.com/education/>

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<https://medium.com/@julsimon>