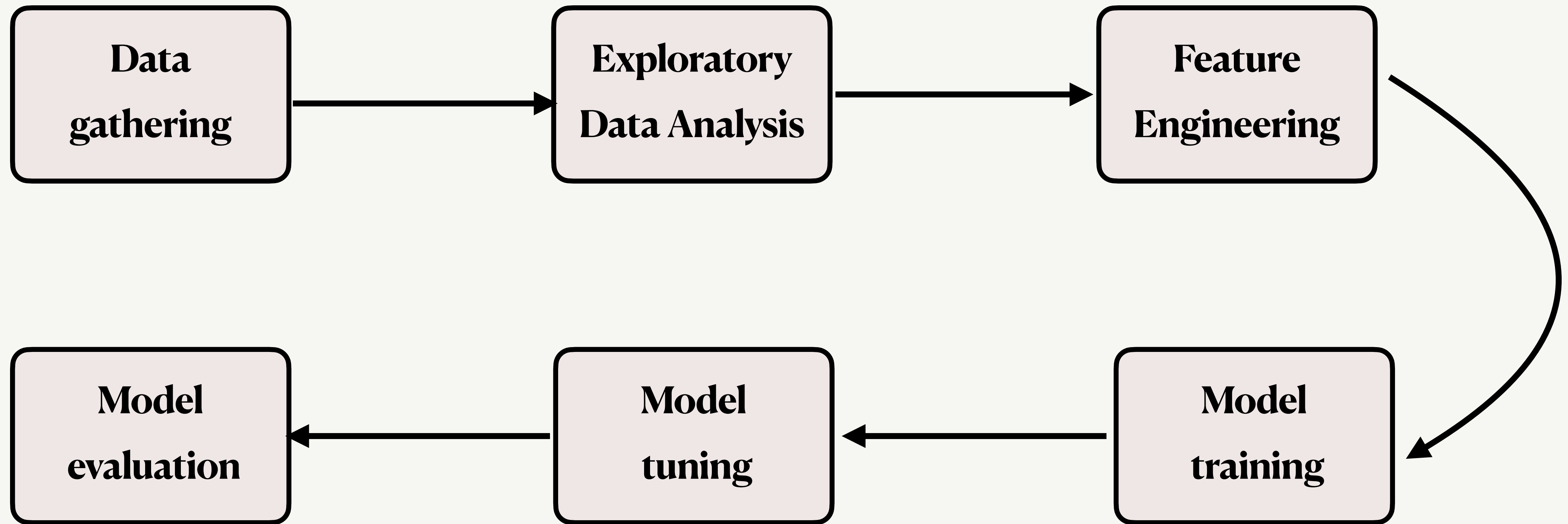


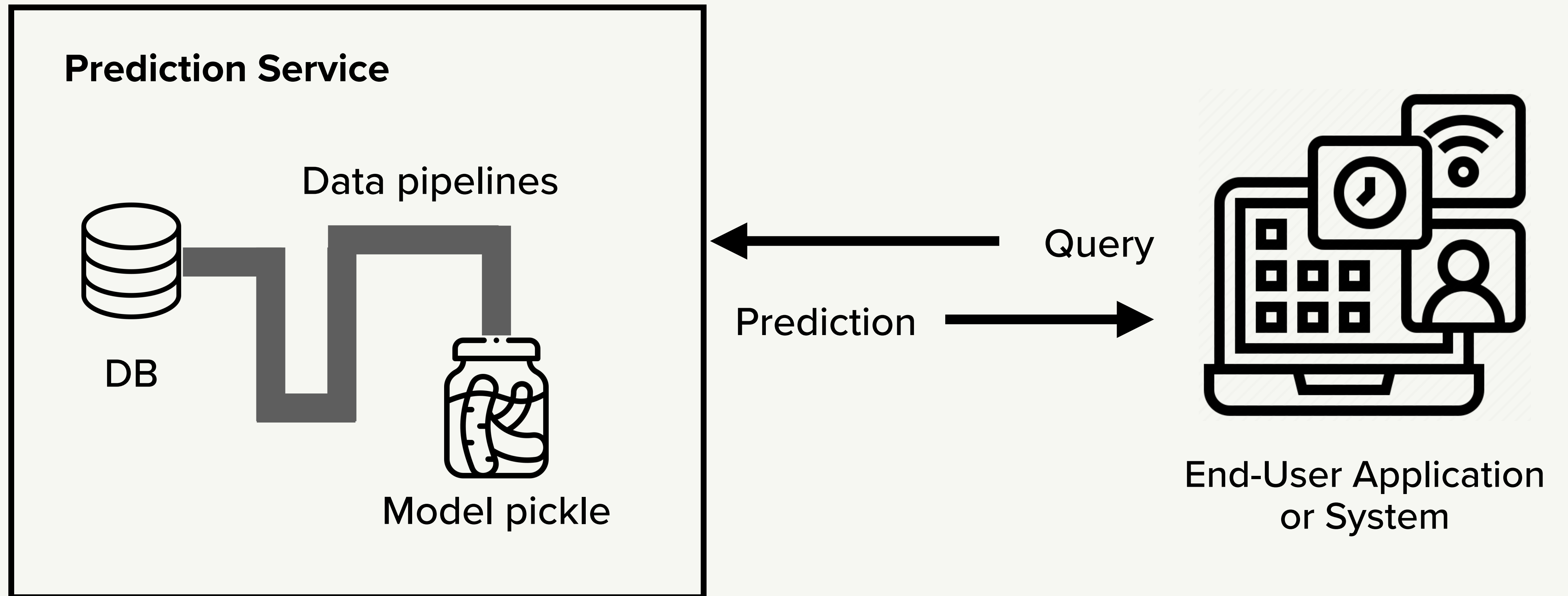
Deploying ML models and all the things that go wrong

Diachuk Marianna, Data Scientist at Restream

Training a model is not the same as...



Training a model is not the same as deploying a model



- Infrastructure
- Integration
- Quality of code
- Testing
- Logging

Issues to focus on during deployment



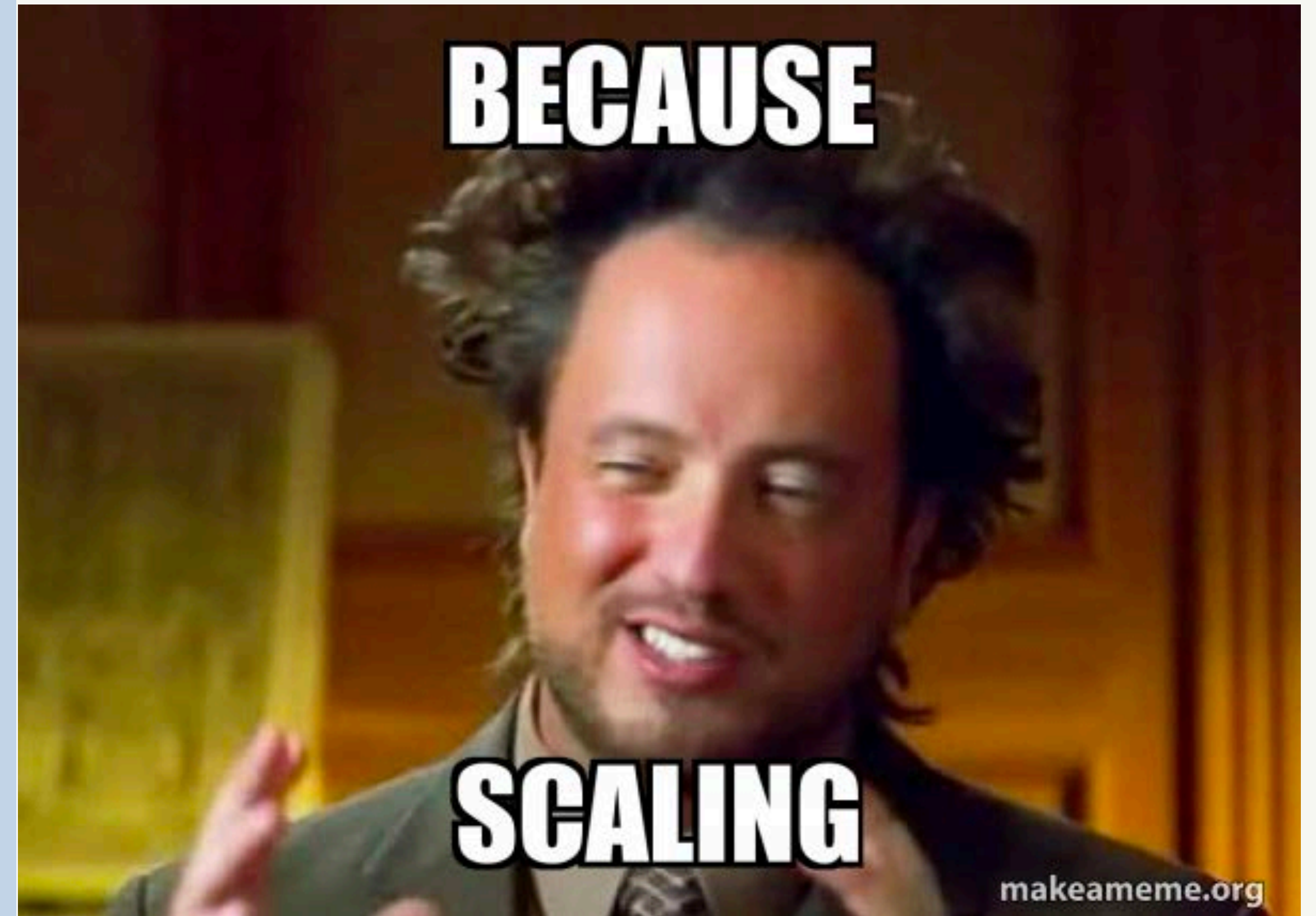
I made my share of mistakes with ML models deployment

but that's what you came for



Who cares about scaling

or how I deployed a model like I never had to deploy any other model in the future

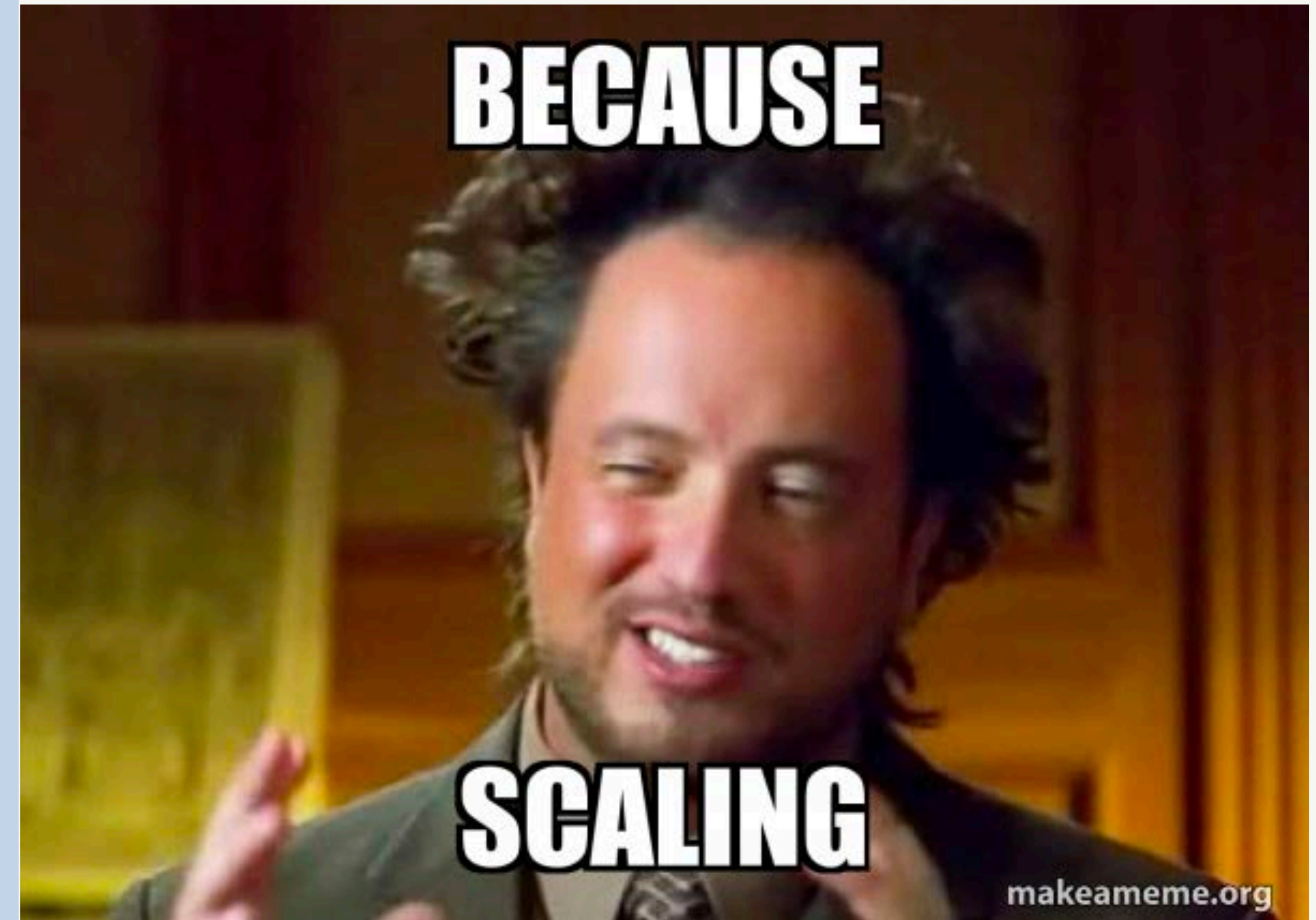


Who cares about scaling

or how I deployed a model like I never had to deploy any other model in the future

- Plan deployment thoroughly
- There's always a reason to scale
- Time used to fix something now is cheaper than it will be later

Lesson Learned



You can't do everything yourself

or how I thought that doing everything
yourself makes it perfect



You can't do everything yourself

or how I thought that doing everything yourself makes it perfect

- Delegation, delegation and once more, delegation
- Synchronisation of tasks in the team can work like a clock
- You're human being, you can't be everywhere at once

Lesson Learned



You are not alone

or how I failed in cooperation



You are not alone

or how I failed in cooperation

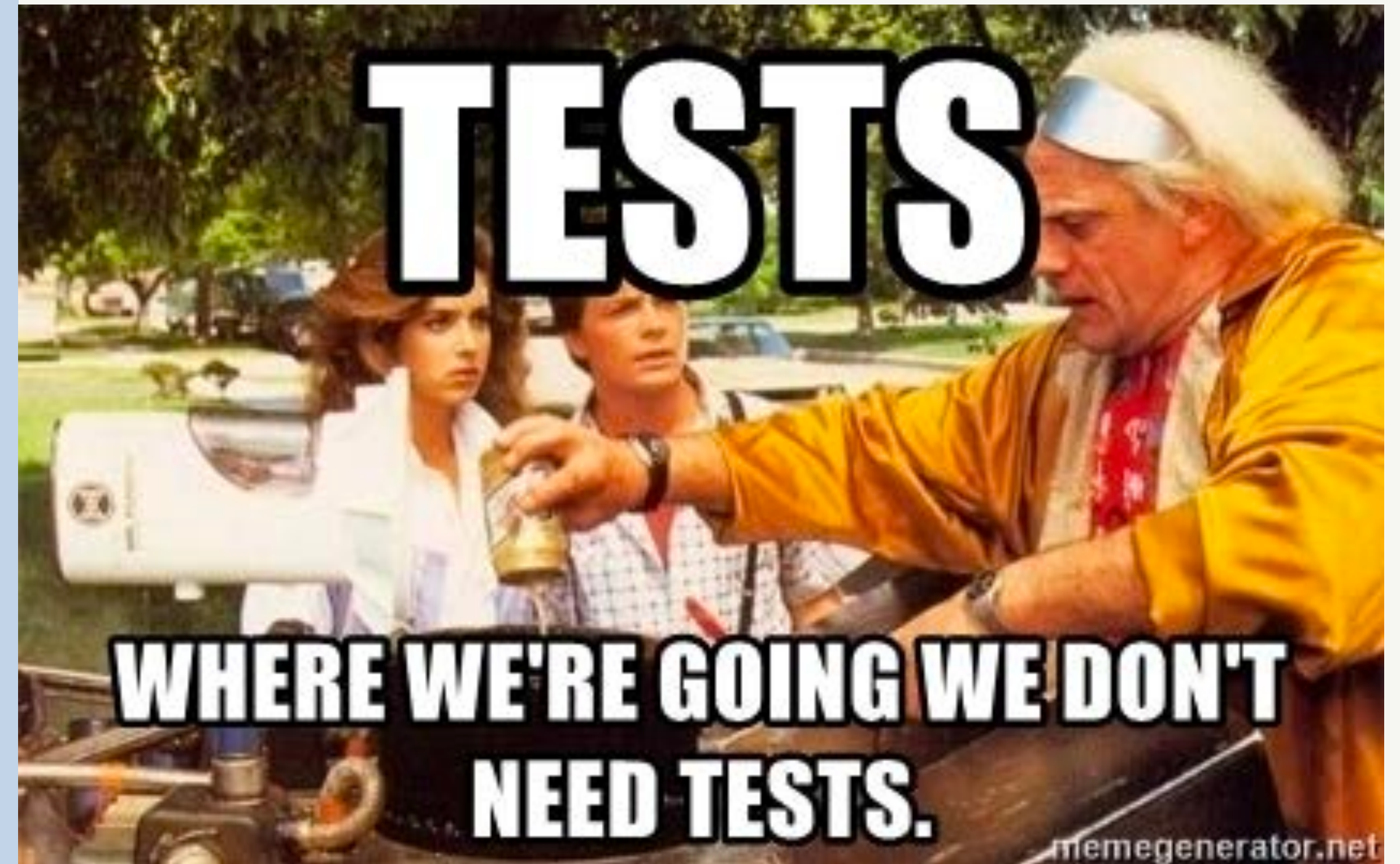
- Cooperation shouldn't be just with data scientists
- Include other specialists in resources and projects planning
- Spread the word about the essence of data science field

Lesson Learned



Data scientists don't need test

or how me and my team deployed
pipelines without tests

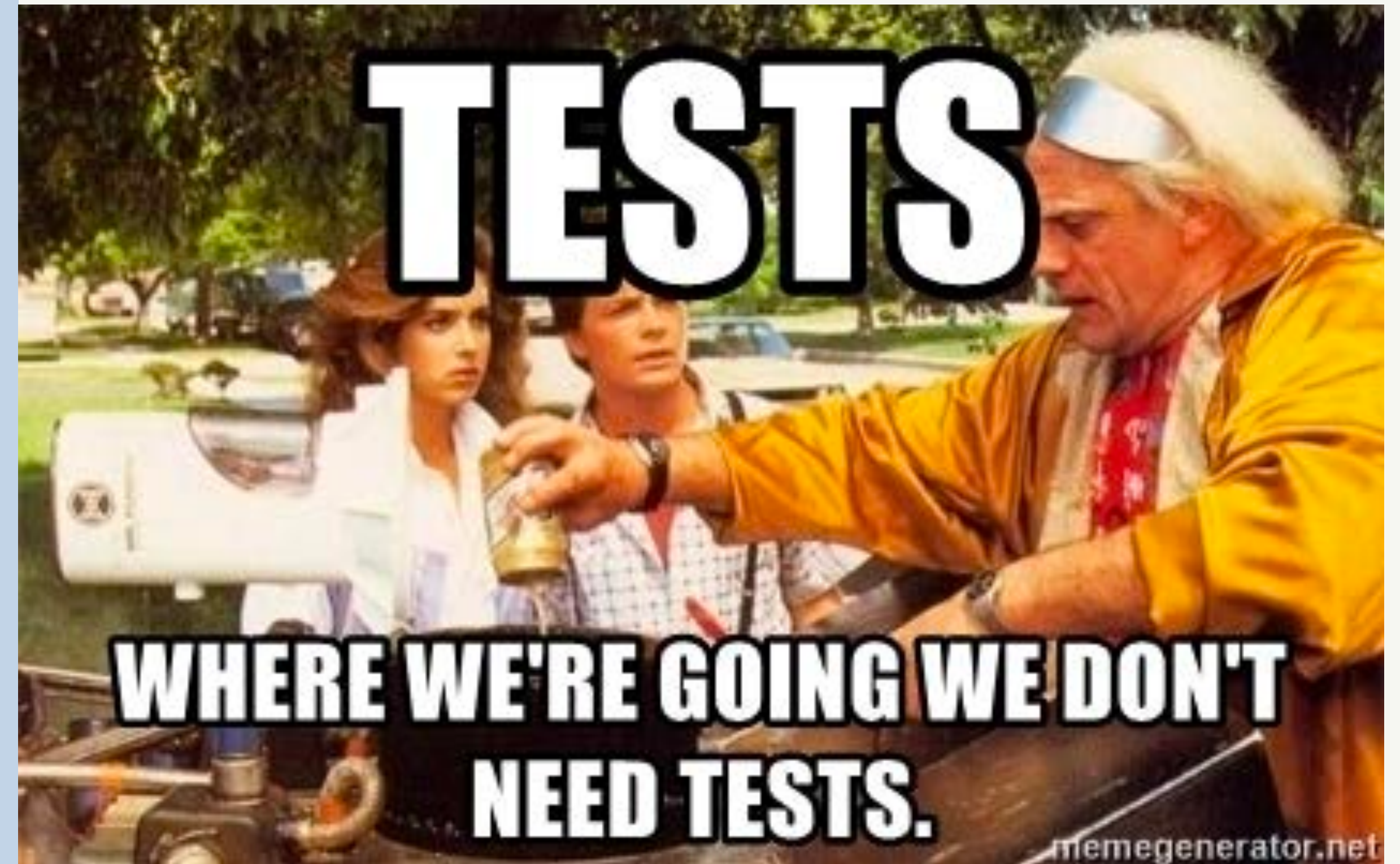


Data scientists don't need test

or how me and my team deployed pipelines without tests

- Work with ML models needs even more tests than any other piece of code
- It's easier to write tests than to fix bugs in production 🙄

Lesson Learned



The past is in the past

or how I didn't have time for monitoring models



The past is in the past

or how I didn't have time for monitoring models

- No monitoring -> no gain
- Model's lifecycle is not over after the deployment
- Tests may pass but other mistakes might still stay there
- The essence of data scientists's work is to find the answer to the question "Why?"

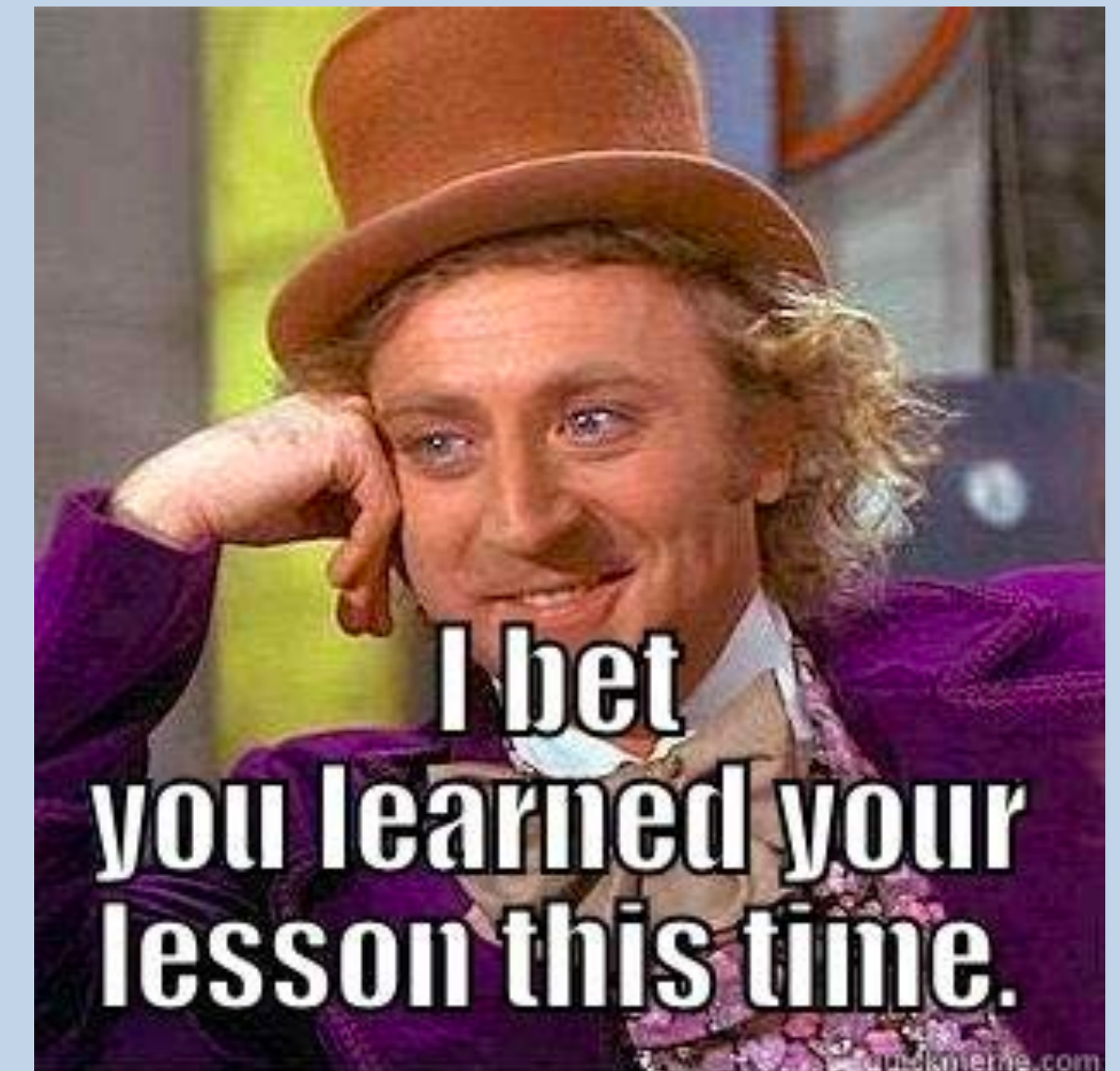


Lesson Learned

What I've learned

and let's hope I won't fail again

- Expect and plan for scaling
- Work as a team (delegate!)
- Work with other teams and departments
- Data scientists need tests even more than anyone
- Model monitoring is the key to success



**Thank
you!**



In production no one can hear you scream.

