

Data scientists supporting scientific researchers: A win-win situation

Matteo Tanadini
Zurich Data Scientists

30th March 2022



Outline

- 1 Who we are
- 2 Aim of the talk
- 3 Use case
- 4 Conclusions

Section 1

Who we are

- Zurich Data Scientists

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- Team

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- Matteo Tanadini

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 - ▶ private companies and governmental agencies

Section 2

Aim of the talk

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- elaborate on benefits and challenges of the data scientists-researchers collaboration

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- via a use case

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- via a use case
- elaborate on **our role** and how it should evolve

Section 3

Use case

Introducing the use case

Aim: Estimate the risk of local transmission via tiger mosquitos of airborne viruses



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- tiger mosquito
- can help exotic viruses spread locally (eg. chikungunya and Dengue viruses)
- human intervention is required in high risk situations

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- quantify abundance of tiger mosquito adults via *human landing collections*

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“Newer” approach



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- indirectly abundance of tiger mosquito adults via eggs abundances

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- introduce estimated abundances in R_0 formulas
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 - ▶ abundance is measured as eggs abundance (in ovitraps)

The “newer” solution in practice

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- this $f()$ may be as simple as a coefficient or... very complex

Question: How to implement the “newer” solution

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Suggest the best-suited method to use

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- ⇒ knowledge transfer

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What is our current role?

- **enhance their data-driven thinking**
- such that the full potential of data science can be exploited in scientific research
- if the researchers don't get what we do...
- then they won't be able to communicate these findings to the scientific community

What researchers can learn from us

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- how to best collect data

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- how to best design experiments
- be aware of what is potentially feasible/not feasible
- right now many analyses are a compromise between what is appropriate and what can be successfully communicated

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Benefits resulting from the knowledge-transfer of the previous collaboration:

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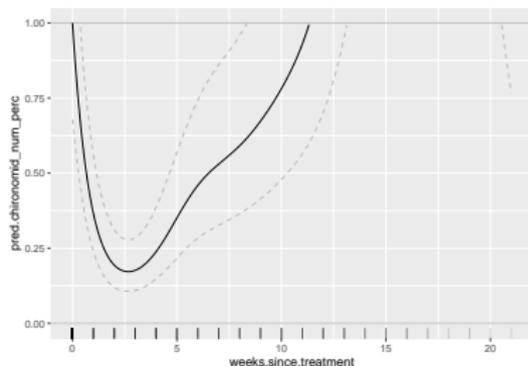
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- better data structure
- Generalised Additive Models explained



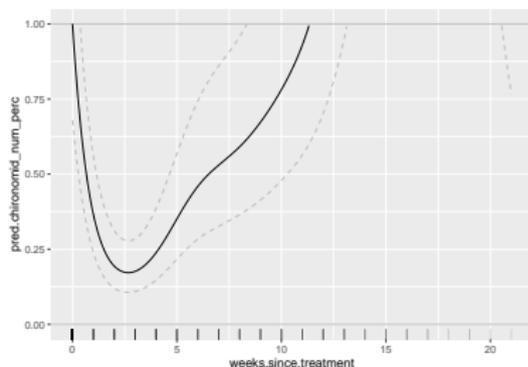
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⇒ this leads to a *win*³ situation



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- nicely commented, reproducible and easy to follow analyses (via dynamic documents)

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- nicely commented, reproducible and easy to follow analyses (via dynamic documents)
- discuss results in depth (focus on graphs)

Section 4

Conclusions

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- data scientists supporting researchers carry the responsibility of make this transition possible
- the transition to data-driven thinking research is interest of all actors involved

End

Looking forward to your questions, comments, and inputs!

Section 5

Additional slides

How to maximise (formal) knowledge transfer (ie. Teaching)

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- via inclusion during all steps of any collaboration project
- via new targets of teaching (group leaders?)

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- to get to the next level of data science (AI) projects
- to see the full potential of data science exploited