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Symbio Europe

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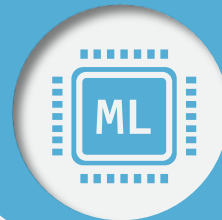
Vice President – Strategic Engagements  
Symbio Europe

AI in QA in AI

AI in QA



QA in AI



# Symbio is a global software engineering and R&D services company

## Think Global, Act Local



We co-create  
**innovative ideas**  
with you and help you  
to co-develop them to  
**transformative digital  
solutions & services**  
that connect, engage,  
and amaze your  
customers.

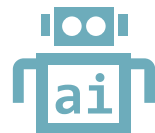


## Our Offering

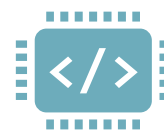
Innovation



Artificial  
Intelligence  
& Robotics



Embedded  
Solutions



Digital –  
Intelligence  
of Things



Quality  
Assurance  
& Testing



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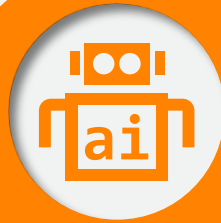
Setting the scene

## PART 02

New opportunities

## PART 03

New Challenges

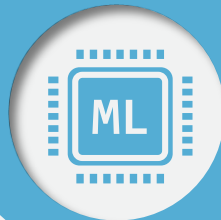


## What is AI?

Looking behind the hype ...

## AI in QA

Making use of AI in QA



## QA in AI

Testing ML based systems

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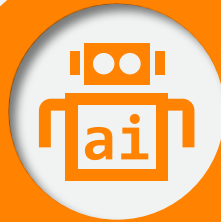
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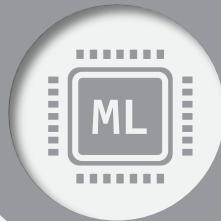


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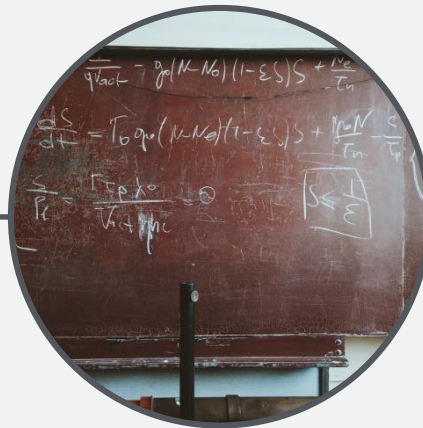
# DEFINITIONS: ARTIFICIAL INTELLIGENCE

## INFORMAL



- › Intelligence exhibited by machines
- › Machines mimicking “cognitive” functions associated with humans
- › Playing strategic games, natural language processing, driving a vehicle etc. ...
- › AI effect: “AI is whatever hasn’t been done yet”

## SCIENCE



- › Study of Intelligent Agents – any device perceiving it’s environment and taking actions maximizing it’s success in some goal
- › Such agents may also learn, hence “Machine Learning” (c. 1959)
- › Or use knowledge i.e. “Knowledge Representation and Reasoning”
- › And bunch of other stuff ...

## MORE BUZZ

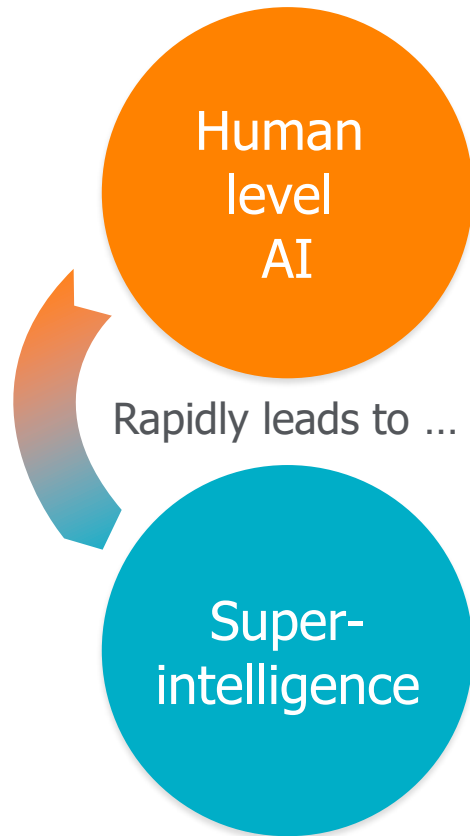


- › Deep Learning – machine learning with a cascade of many layers of non-linear processing units
- › Shallow Learning – not deep
- › GOFAI – Symbolic AI, that was when grandpa was doing studies
- › Artificial General Intelligence or “Strong AI” or “Full AI”, i.e. hypothesised human level AI
- › Superintelligence ...

# SO WHEN DO WE GET GENERAL AI? OR SUPERINTELLIGENCE?

Difficult to predict but there's so much media buzz on this that let's spend a few minutes on the topic

## 1 Superintelligence?



## 2 Ask the experts?

Combined likelihoods for human level AI from several "AI Expert" polls:

- by 2022 – 10% probability
- by 2040 – 50% probability
- by 2075 – 90% probability

Most of these people expect that Superintelligence might follow in about 30 years. Some are bit more enthusiastic

## 3 Track-record?

If you asked the same question 50 years ago from the same set of people (i.e. the guys who led the field mid 1960s) the answer was that general AI will be there in about 20 years ...



"It's a bit like worrying about the overpopulation on Mars"



# YOUR OFFERING WILL BE DATA-DRIVEN

## WHERE ARE WE TODAY?

### Big Data

zettabytes  
today

### CPU / GPU

MapReduce  
Deep Learning

### Cloud

Compute  
Storage

### DX

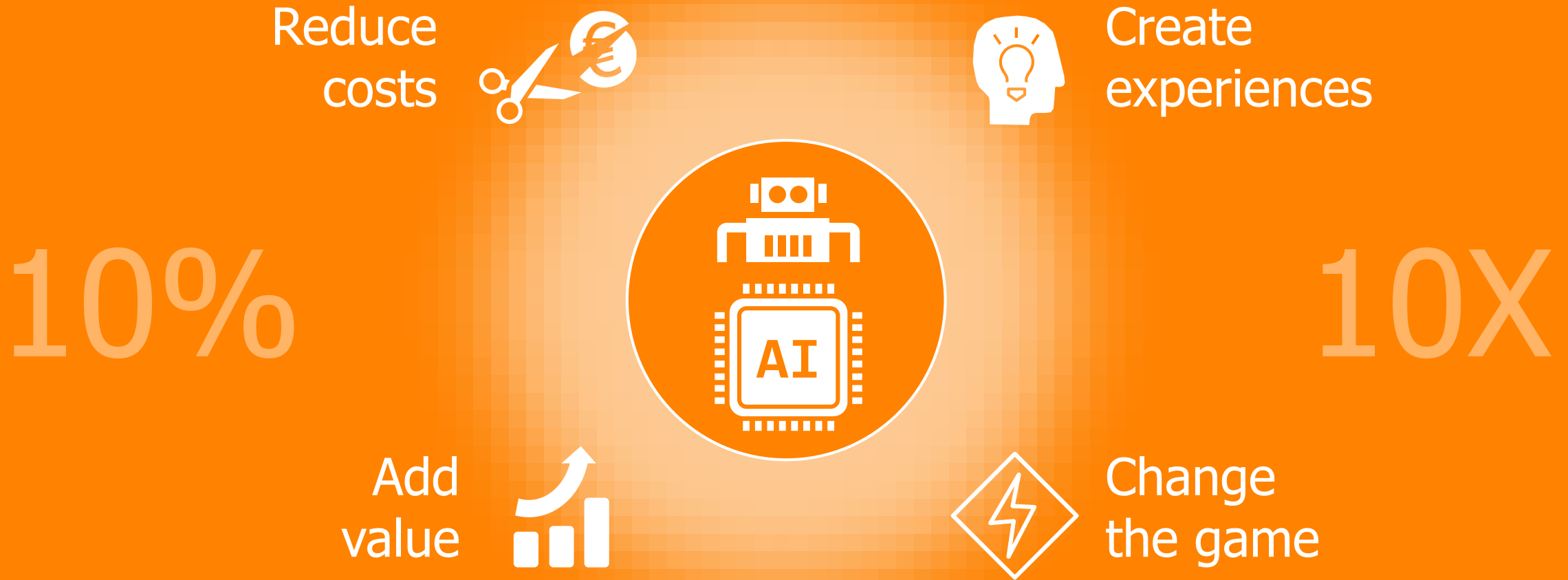
Abstraction level  
Ease of use

Automatic  
Computing  
Engine by  
Alan Turing

Machine  
Learning  
and AI

Super  
Intelligence  
(?)

# A SIGNIFICANT OPPORTUNITY FOR ALL

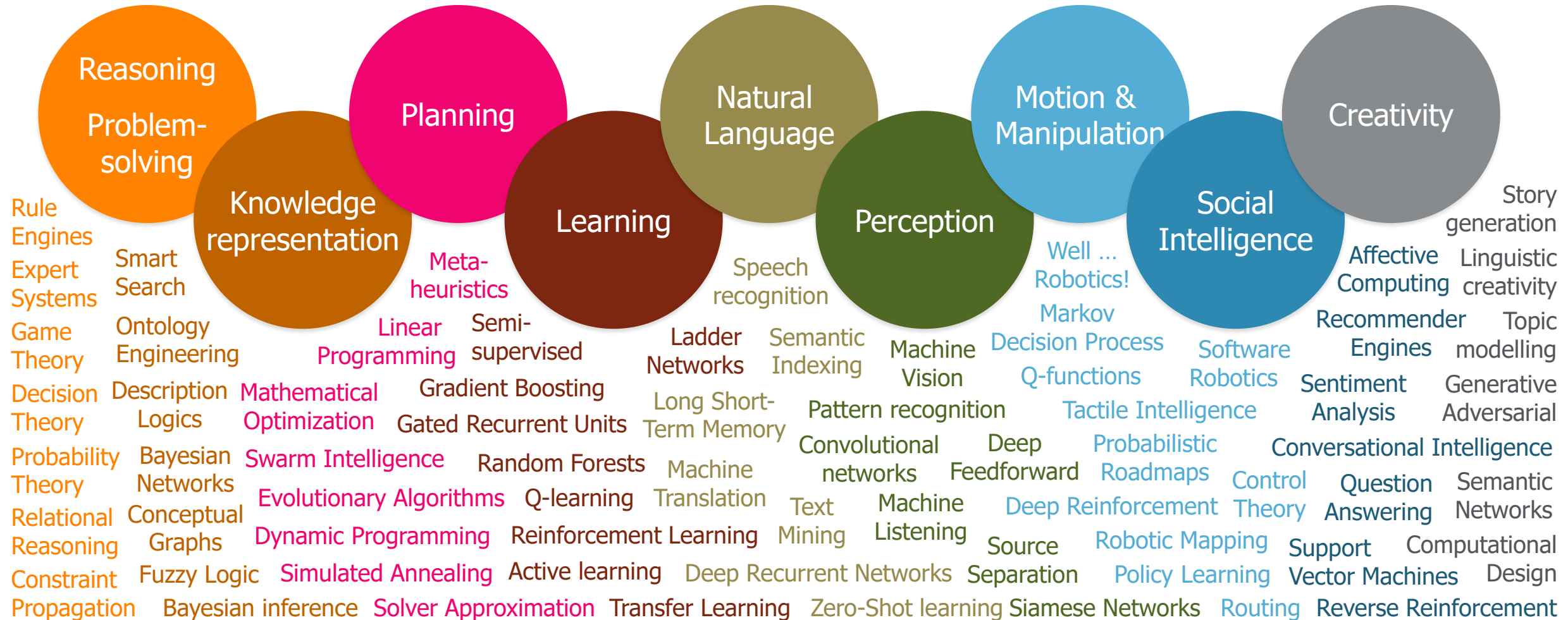




Why?

# IT'S A TOOLBOX!

AI is not a single field of research or one specific approach



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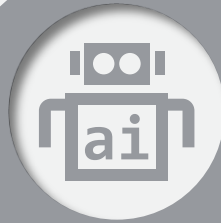
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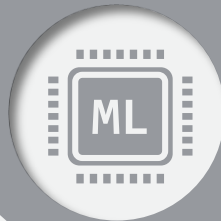


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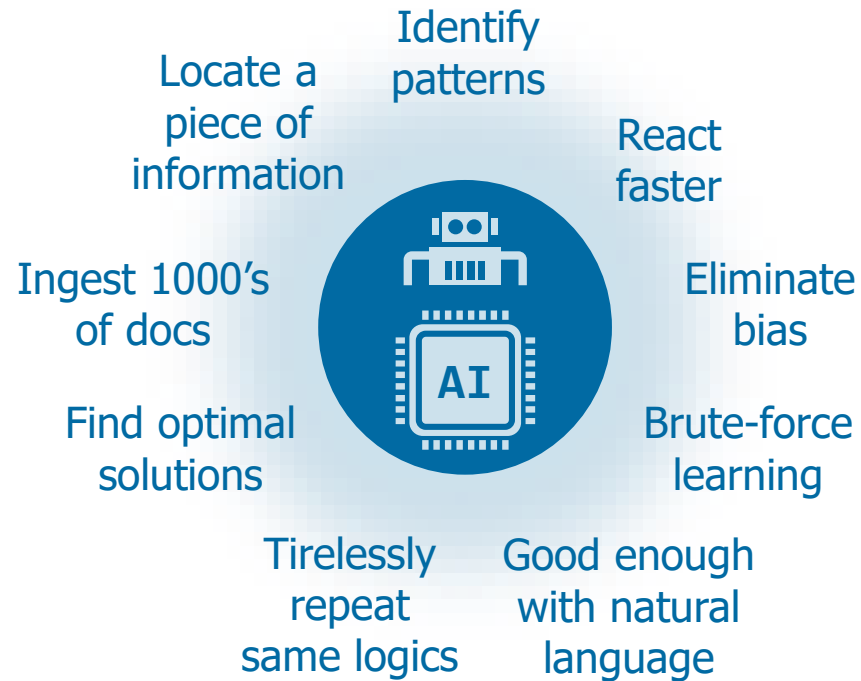


## QA in AI

Testing ML based systems

# HOW DO YOU COMBINE HUMAN LABOUR AND INTELLIGENT MACHINES?

## Machines can do much better in many areas

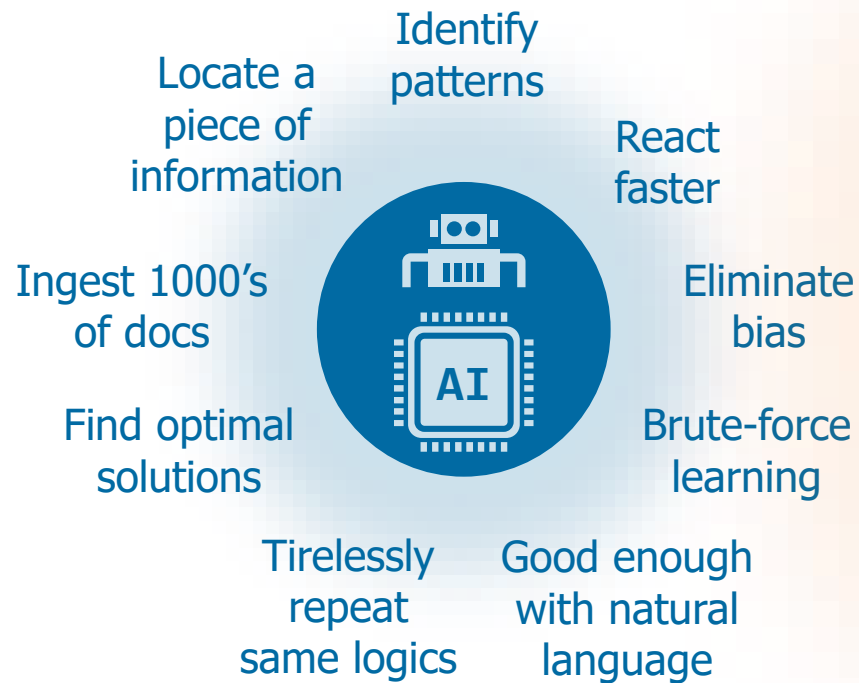


## The right person will win hands-down in others



# WHAT YOU REALLY NEED IS A SMART COMBINATION OF BOTH

## Machines can do much better in many areas



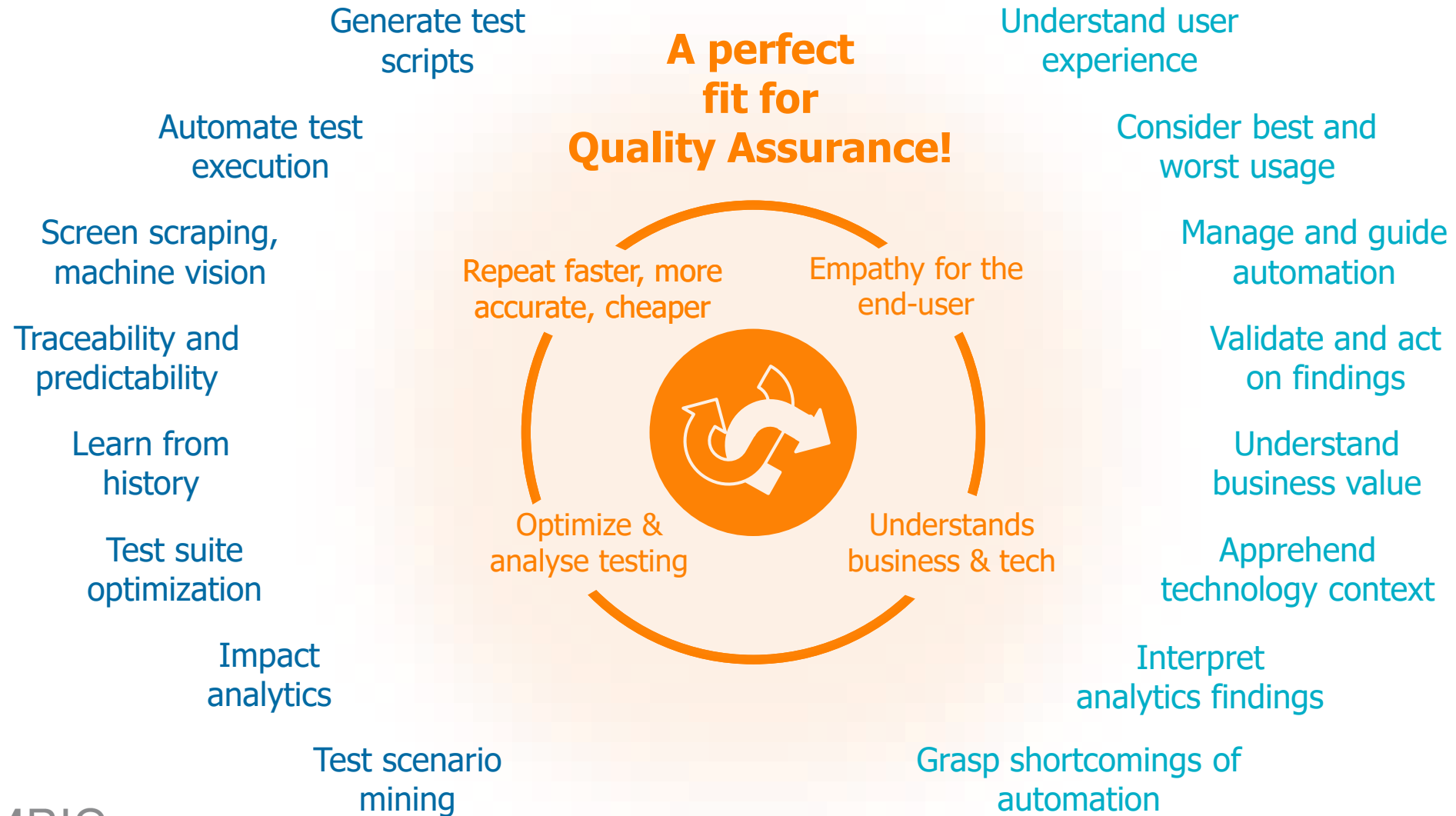
## What if you combine those strengths?



## The right person will win hands- down in others



# TOOLS ARE HERE TO AUGMENT – NOT TO REPLACE ROLE OF THE TESTER WILL CHANGE





# THE NEW TESTER: THE AUTOMATOR NOT THE AUTOMATED

Understands  
automation  
possibilities

Knows  
how to use  
the tools

Uses  
the power  
of analytics  
and data  
science

Grasps  
end-user and  
business  
context

Approves &  
interprets  
automation  
results

Handles  
the more  
complex  
issues



# CASE ROBOT AIDED TEST AUTOMATION

## Multi-Device UI Test Automation with RATA

IF YOU CAN TOUCH IT, RATA CAN TEST IT!

- › Non-intrusive black-box testing for devices
- › Both functional and non-functional testing
- › Platform independent – either touch or software based

### Solution

- › Optical Character Recognition (OCR) and Icon detection enable UI changes without extensive rework of test automation scripts
- › Model-based approach for generating pseudo-random test paths
- › Optofidelity robot, imaging and high speed cameras
- › Automated reporting and analytics

RATA has been successfully applied in several industries

- › Automotive
- › High tech / consumer devices
- › Industrial Equipment



Check out our RATA video:  
[https://www.youtube.com/watch?v=n\\_915xMRGAo](https://www.youtube.com/watch?v=n_915xMRGAo)

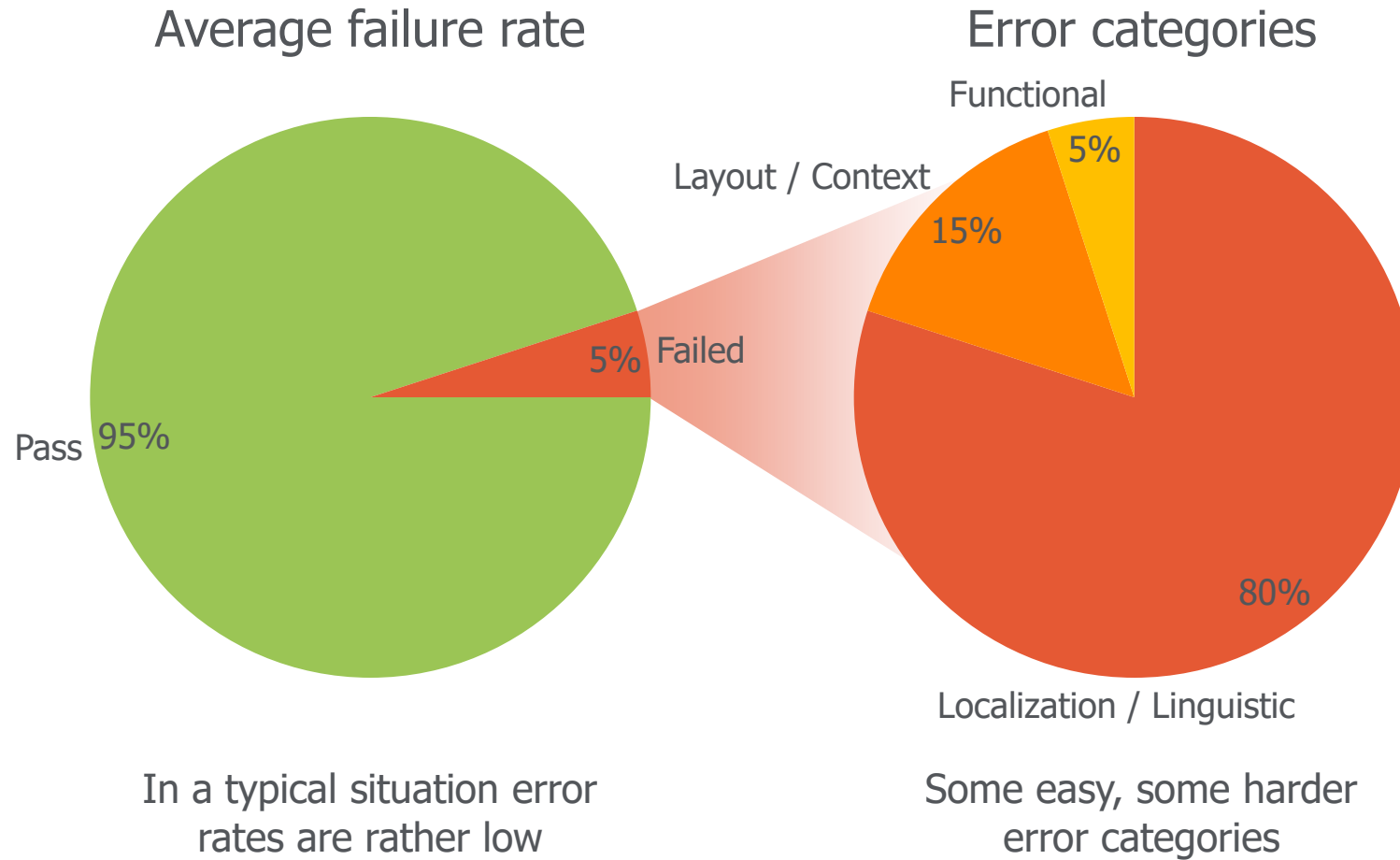


### THE REAL VALUE

- ✓ One-time set-up – can run 24/7
- ✓ Accelerated R&D and testing cycles
- ✓ Improved test quality and repeatability
- ✓ Also performance, stability, longevity testing
- ✓ Easy benchmark / competitor analysis
- ✓ Ability to test new kinds of control and activation methods
- ✓ Test device performance without modifications or connections

# FUTURE DIRECTIONS: MACHINE LEARNING FOR LQA

Localization Quality Assurance is a complex domain with a lot of routine work



Machine Learning to help?

Some **easy** categories

- › Text overlapping UI elements
- › Typos, grammar errors, unlocalized, truncated, ...

Some **harder** ... e.g.

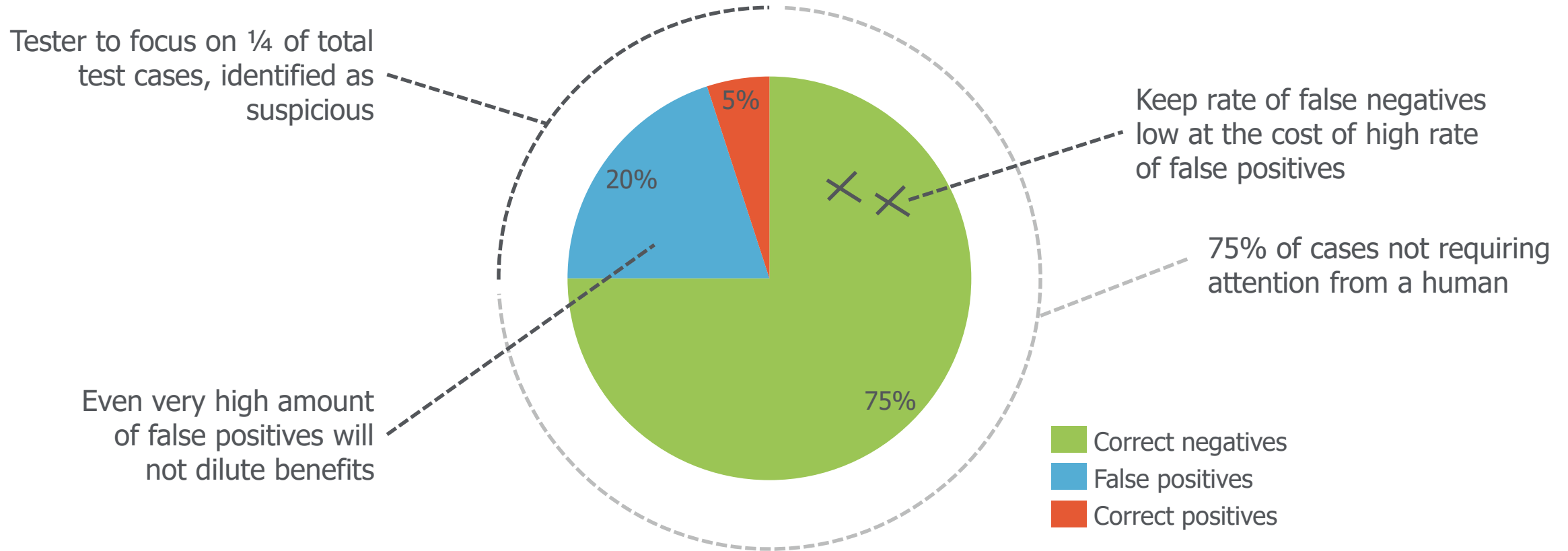
- › Technically correct translation but still not right
- › Inconsistency issues

Machine translation tools are developing fast ...

... but they will not solve our problem

# SO A DEAD END ... ? NOT SO!

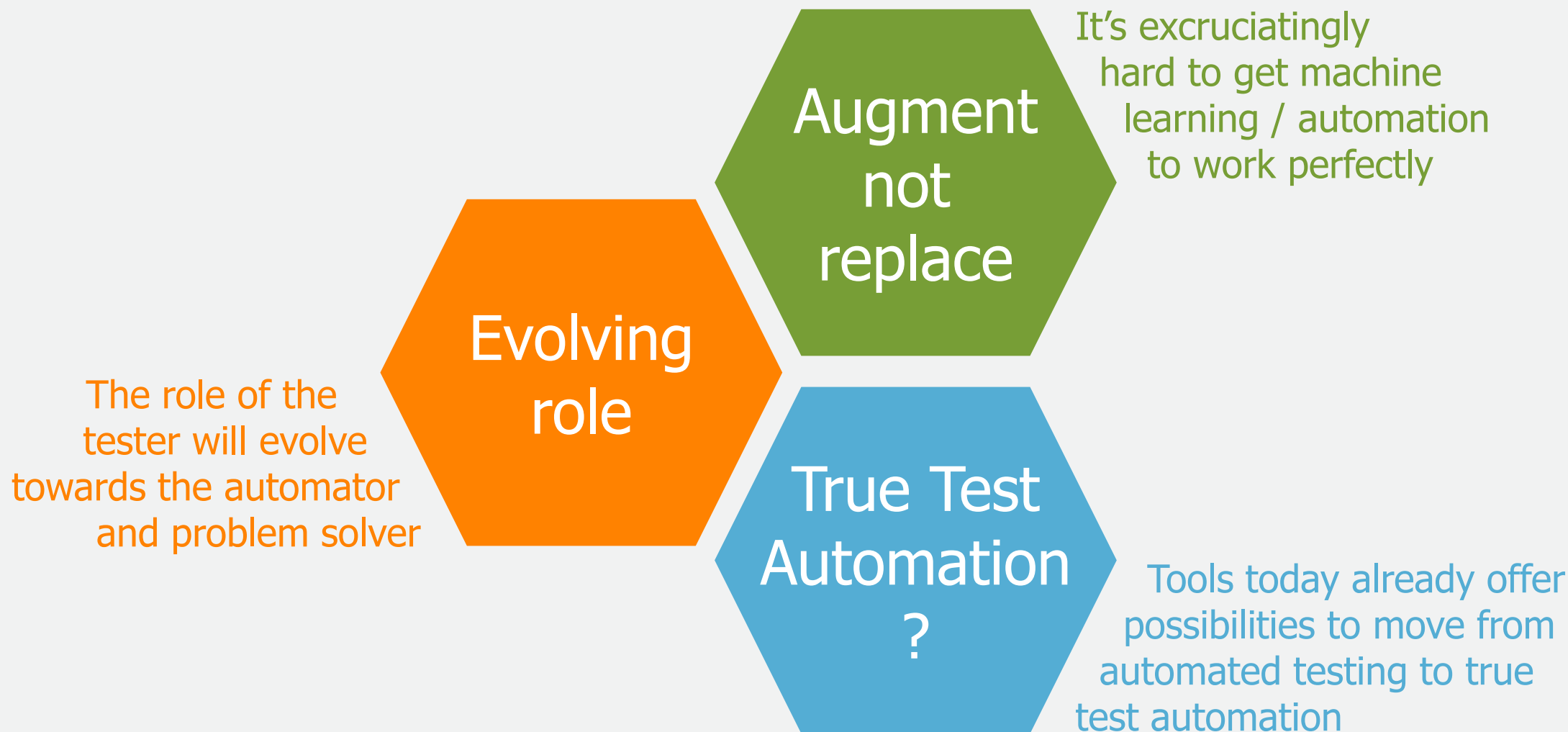
You need that smart combination of human and machine capabilities



Other benefits will also follow

- > Immediate feedback
- > Regression testing
- > Cycle time reduction
- > Improved quality
- > Less error leakage
- > More interesting for tester

# IN SUMMARY





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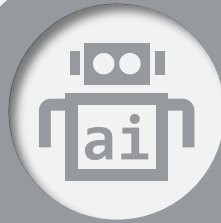
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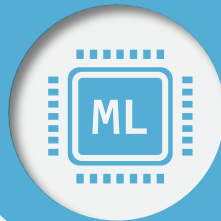


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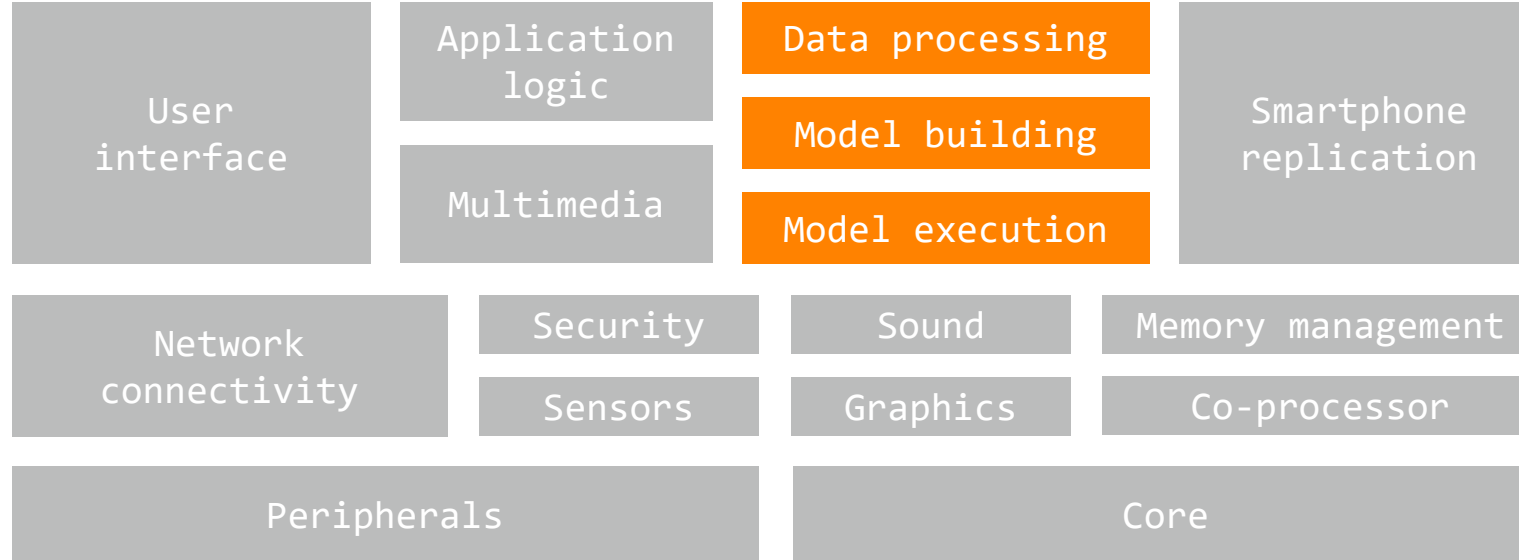
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# THE CHALLENGE:

## HOW DO YOU ENSURE QUALITY FOR DATA-DRIVEN PRODUCTS?

System architecture for our hypothetical ML enabled solution



### First observation

In a real application the Machine Learning (ML) components will only constitute a small part of the whole

### A bit of reflection

You probably used ML since you wanted to

1. infer something out of data
2. get self-customizing behaviour; and/or
3. you just plain could NOT even start to program it by hand

More likely with central impact to CX

New paradigm:

**Data is code!**

Let's consider a bit what could go wrong here ...

## Your data?



“It’s a picture of a basketball player”

Your training data might be biased or otherwise unfit in so many different ways

## Your data again?



“This API just started to behave funny”

Earlier you had code dependencies – now you’re also getting data dependencies

## Your ML engineer?



“It’s just not converging how it should be”

Although tools are getting easier to use, a small typo could still produce a valid model that just does not train right

## Other people?



“That seems to be an ostrich”

Adversarial attacks appear to work well. Consider for example automated process for handling insurance claims...



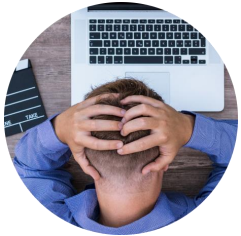
Data defines the behaviour of your system

Test the data! Test the distributions. List possible sources of error. E.g. can your system end up impacting it's own training data?



Your data pipeline processes data in complex fashion

Keep track of data dependencies. Version your models and data. Clean up your data. Monitor it. Consider live validation.



The tools are a bit black box and complex to use

Consider coding patterns to follow. E.g. unit tests to just validate (ML related) code, not behaviour. Consider approaches such as mutation testing.



There could be complex ways to misuse them too

Is it realistic (e.g. it's possible and there's good benefit / risk ratio)? If so – good luck!  
Use detection schemes ...<sup>\*</sup>

# IN SUMMARY

Quality Assurance  
just got way more  
complex ... luckily  
we are already  
familiar with code

Think of  
data as  
code

Look for  
and use  
(good old)  
best practices

However many data  
science / machine  
learning practitioners  
may not be familiar  
with Agile & DevOps

Be bold ...  
and also  
be careful  
out there!

You definitely should  
start exploring, the  
future will be  
data driven ... just  
keep in mind it's all  
rather new

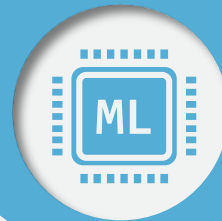


Folks, it's time for

# Q&A

AI in QA in AI

AI in QA



QA in AI