PhD lightning talk

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THE FRENCH AEROSPACE LAB

TABLE **OF CONTENT**





PhD Subject

Formal verification of high dimension ML based systems

Context

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You

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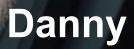
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Runway position

longitude
latitude
altitude



Aircraft position

longitude
latitude
altitude

Aircraft inclinaison

1- yaw 2- pitch 3- roll

You

distanting of

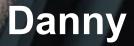
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864

h/m

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N N N







Runway position

1- longitude Known2- latitude Known3- altitude Known



Aircraft position

1- longitude	Known
2- latitude	Known
3- altitude	Known

Aircraft inclinaison

1- yaw 2- pitch 3- roll

You

distant strains

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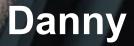
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h/m

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N N N

1/1/1







Runway position

1- longitude Known2- latitude Known3- altitude Known



Aircraft position

1- longitude	Known
2- latitude	Known
3- altitude	Known

Aircraft inclinaison

1- yaw 2- pitch 3- roll Known Known <mark>Unknown</mark>

You

distanting of

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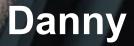
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h/m

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Landing phase

Buenos Aires - Argentina











864



Landing phase

Buenos Aires - Argentina











864



Successful landing Buenos Aires - Argentina



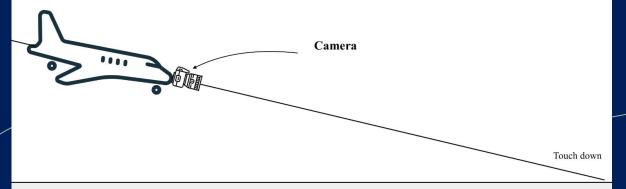
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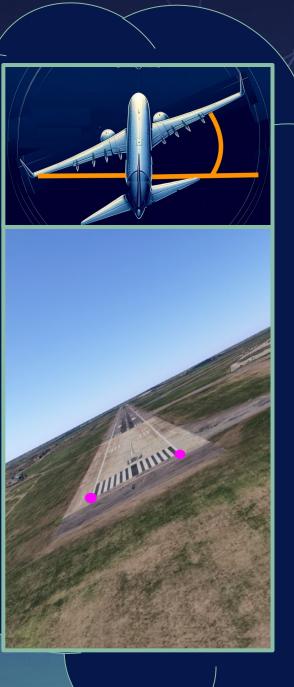


What if we added redundant "roll" function with a camera ?



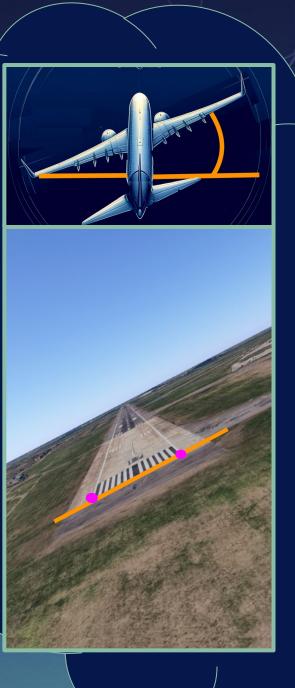






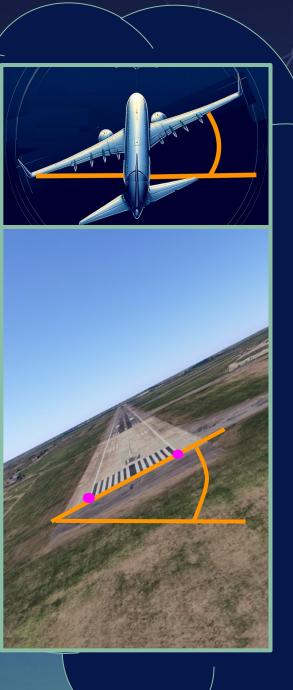






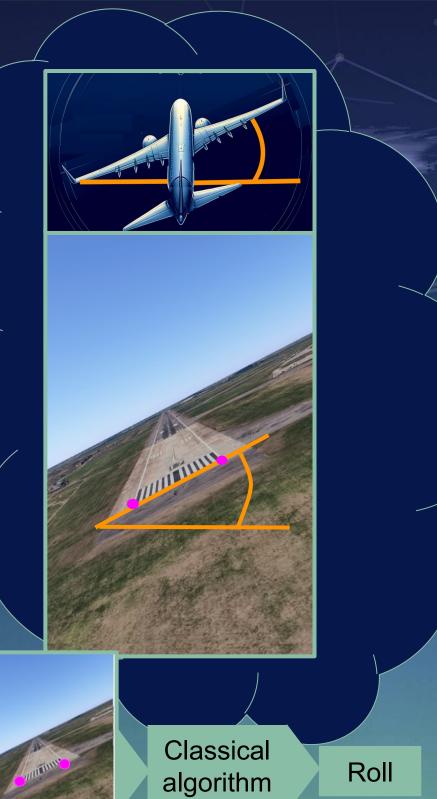














You









Classical algorithm Roll

How can we certify this machine learning model?



 \bigcirc

Subject



Subject

Possible strategies for ML certification



Subject Formal method

Possible strategies for ML certification

Context Subject



Neural network



Property



Pre/Post conditions



Formalized property



Verification tools



Neural network

Successive image transformations





Neural network



Property



Pre/Post-conditions

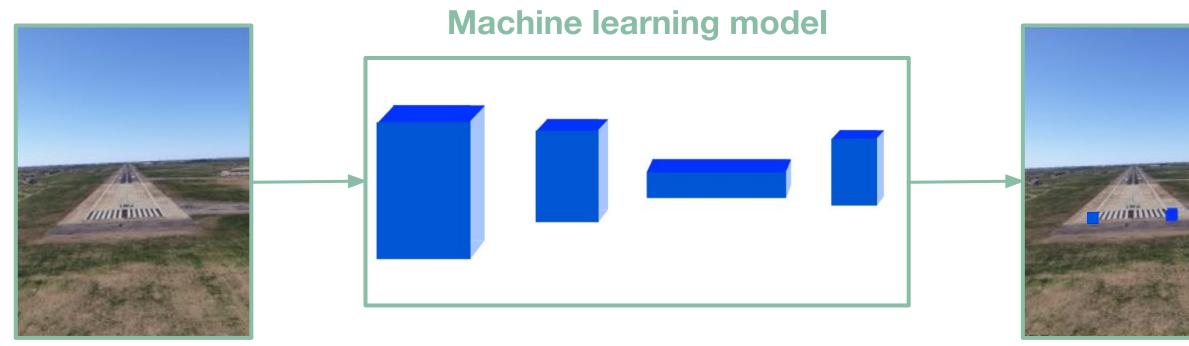


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Neural network

Successive image transformations





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Context
Subject



Neural network



Property



Pre/Post-conditions

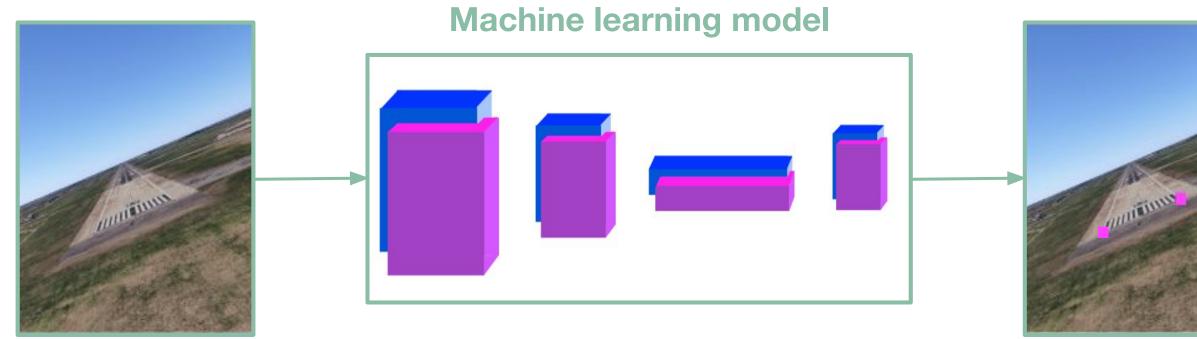






Neural network

Successive image transformations





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1 Context **2** Subject



Neural network

Property



Pre/Post-conditions





Property

Requirement on desired model behavior





Neural network



Property



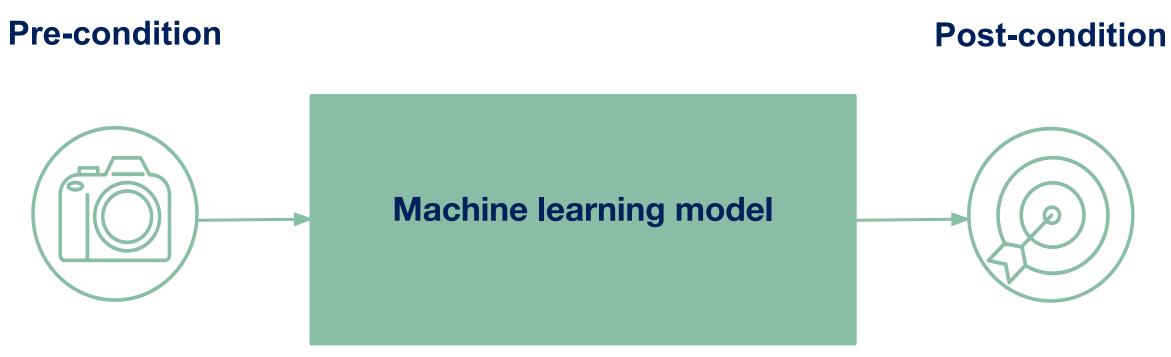
Pre/Post-conditions



65

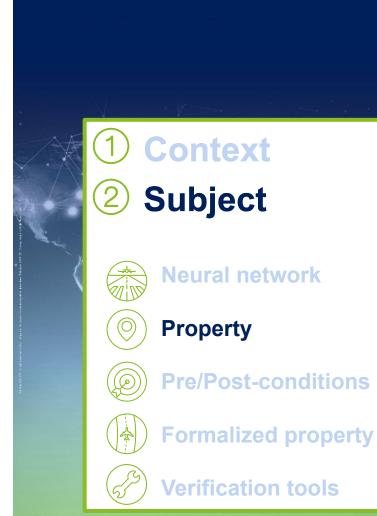




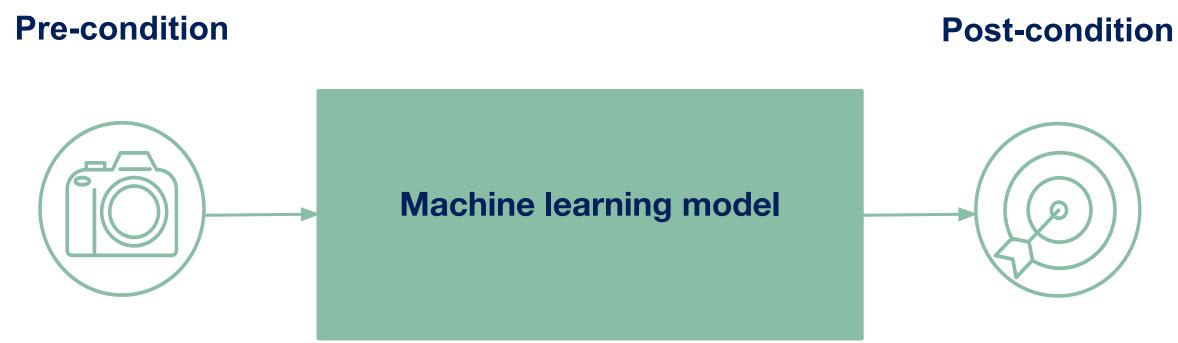


Neural network input constraints

Neural network output constraints



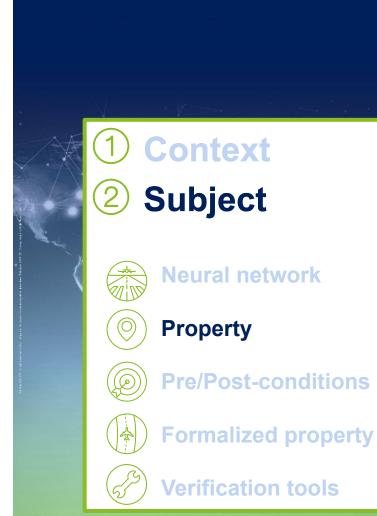




Neural network input constraints

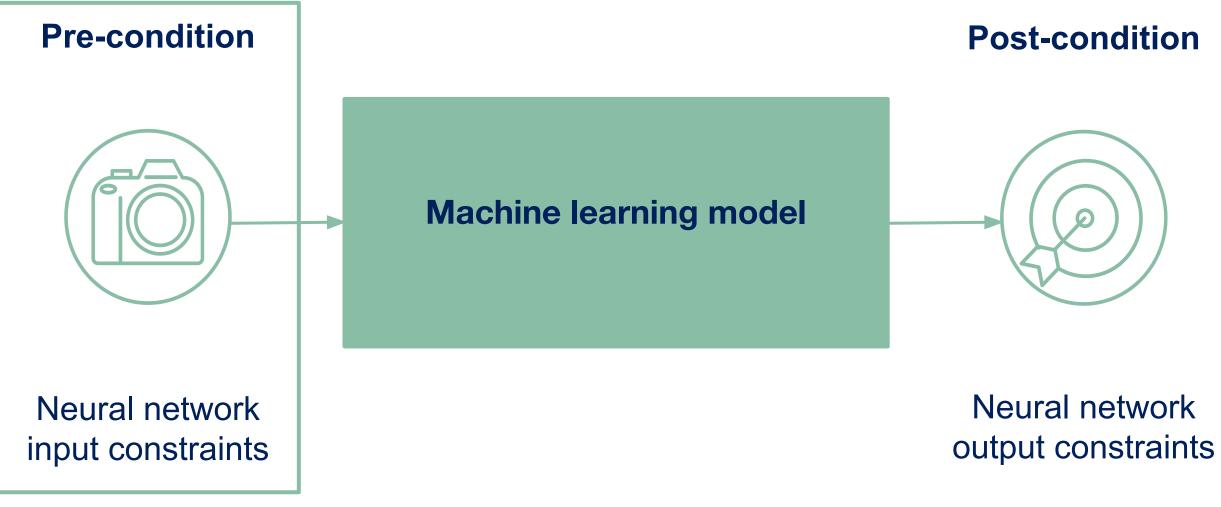
Neural network output constraints

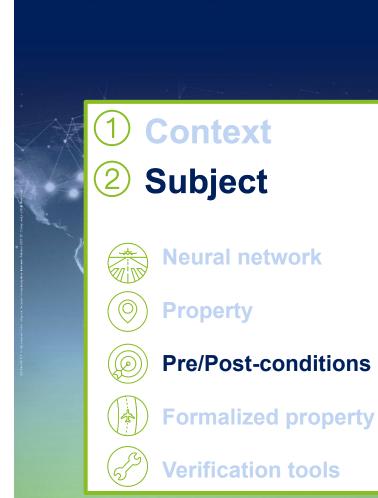
If the input meets the pre-condition, then, the output meets the post-condition





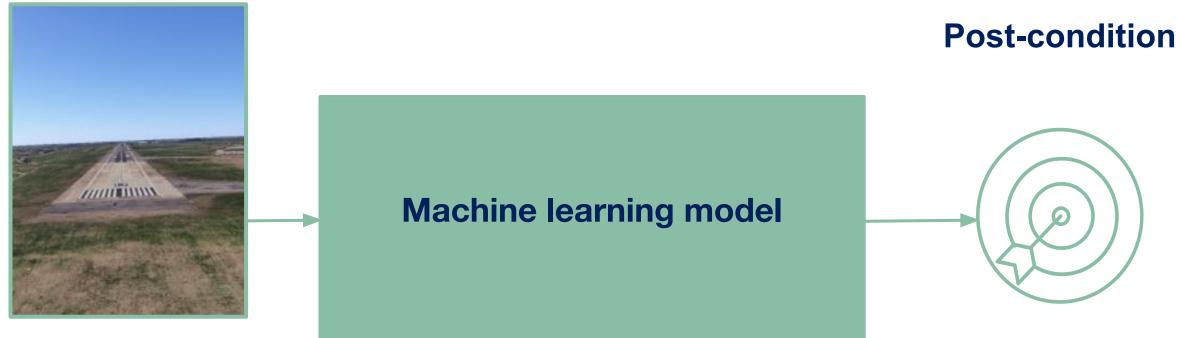
Neural network input constraints



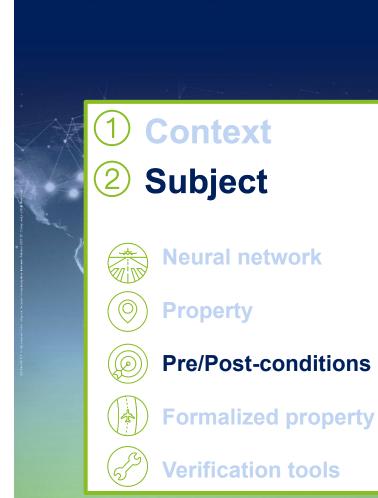




Neural network input constraints



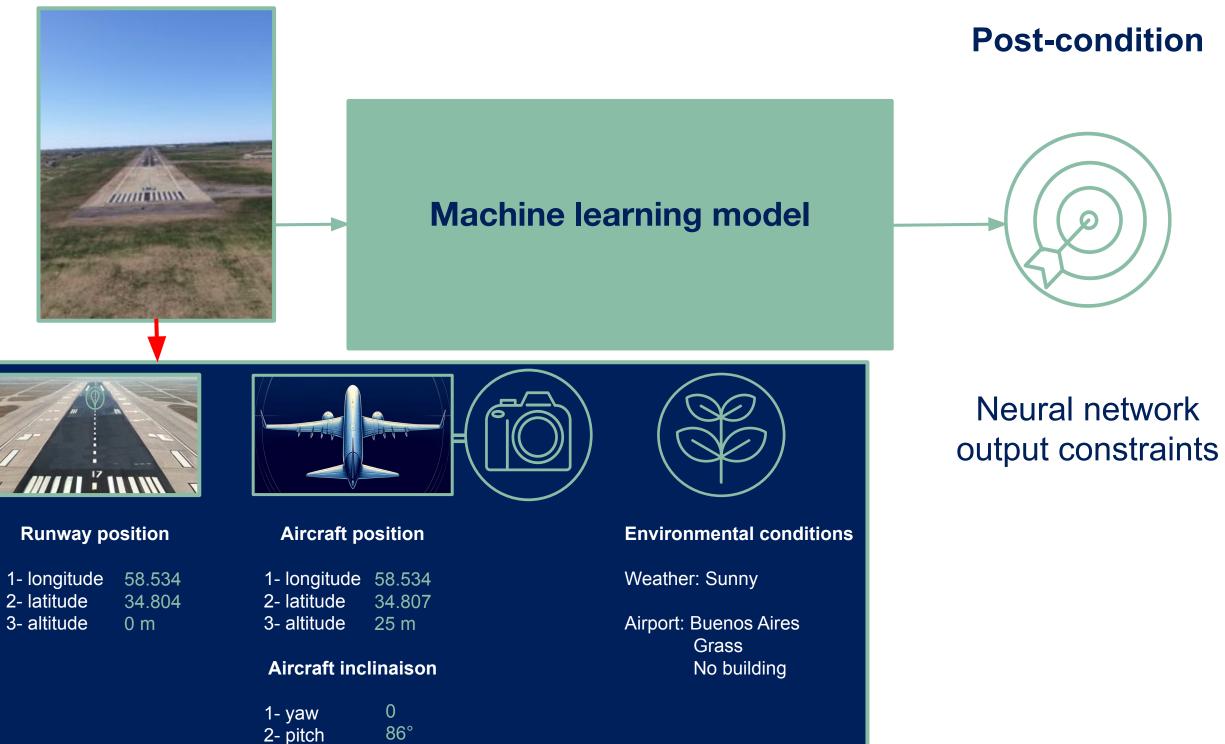
Neural network output constraints

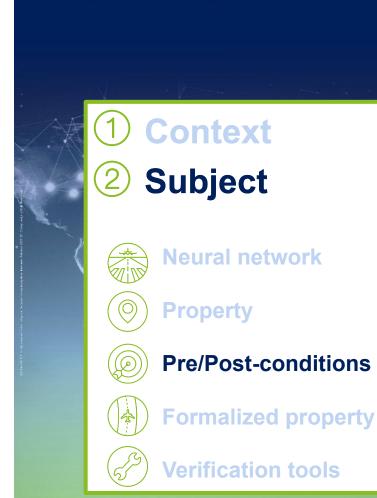




Neural network input constraints

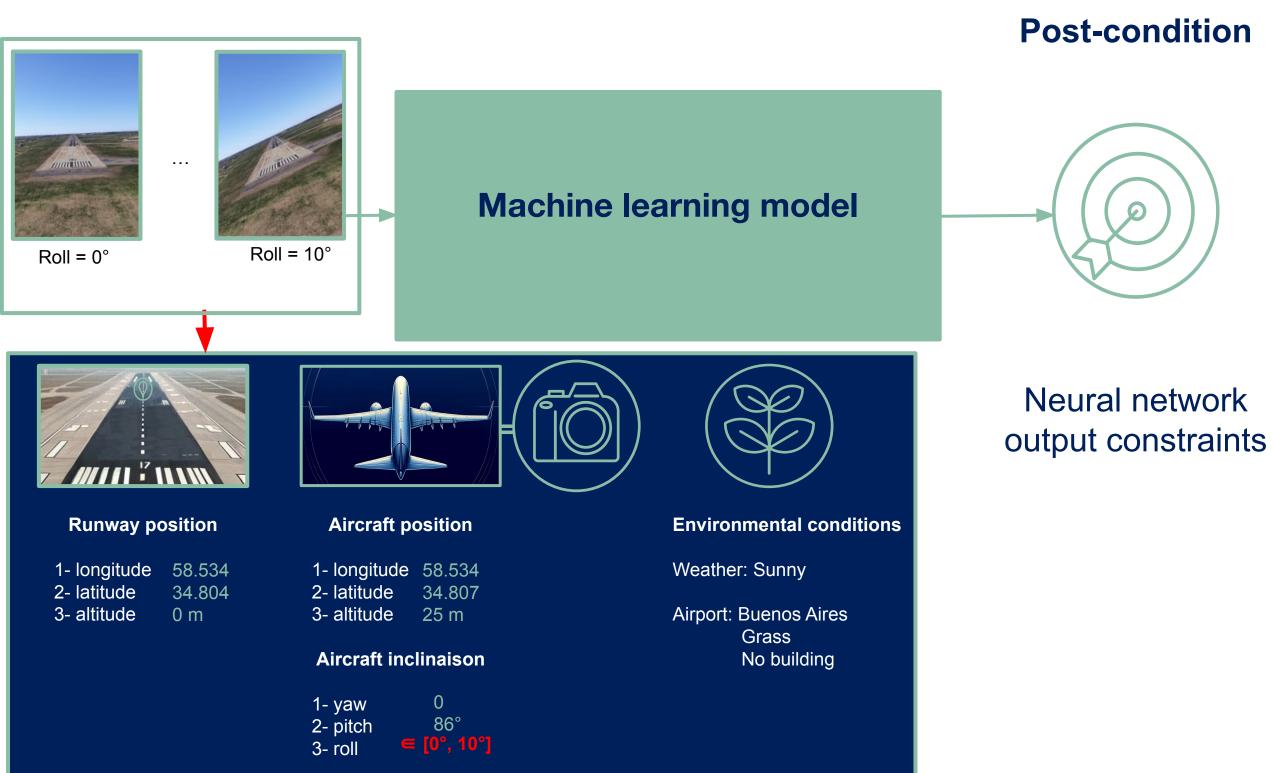
3- roll

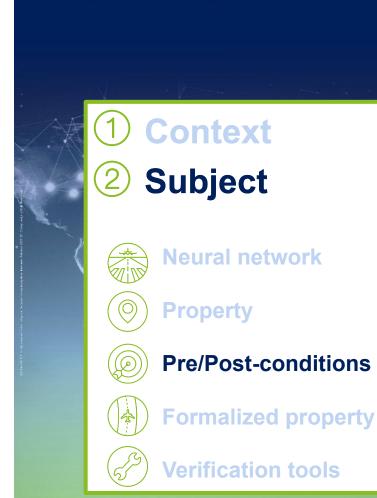






Neural network input constraints

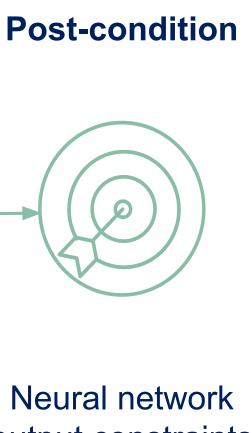






Neural network input constraints **Machine learning model**

Neural network output constraints







Neural network

Property



Pre/Post-conditions



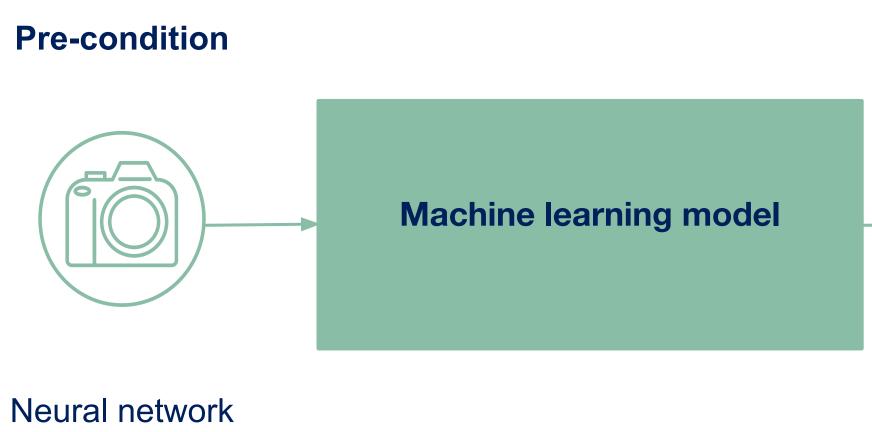
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Post-conditions

Desired model behavior

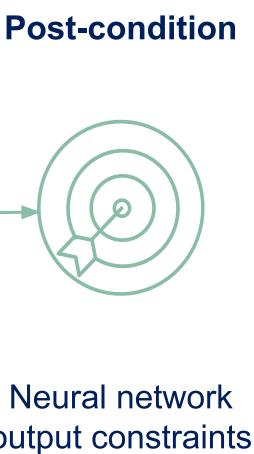


input constraints

Neural network output constraints

Defined by:

- Business requirements -
- Technical constraints
- Performance
- . . . -







Neural network

Property



Pre/Post-conditions



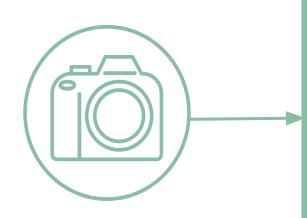




Post-conditions

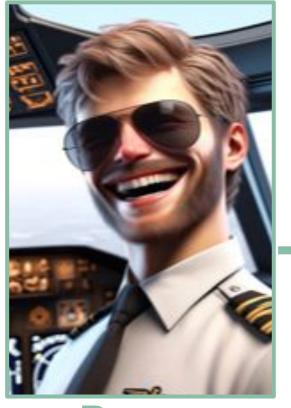
Desired model behavior

Pre-condition

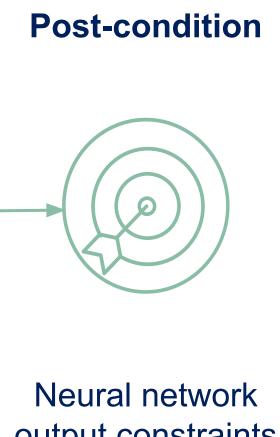


Machine learning model

Neural network input constraints



Danny



output constraints

Defined by:

- Business requirements
- Technical constraints
- Performance -

-. . .





Neural network

Property



Pre/Post-conditions







Post-conditions

Desired model behavior

Pre-condition Machine learning model Neural network

Post-condition

Neural network output constraints

Delimited zone

input constraints











Neural network

Property



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Pre/Post-conditions







Formalized property

Mathematical traduction





Neural network



Property



Pre/Post-conditions



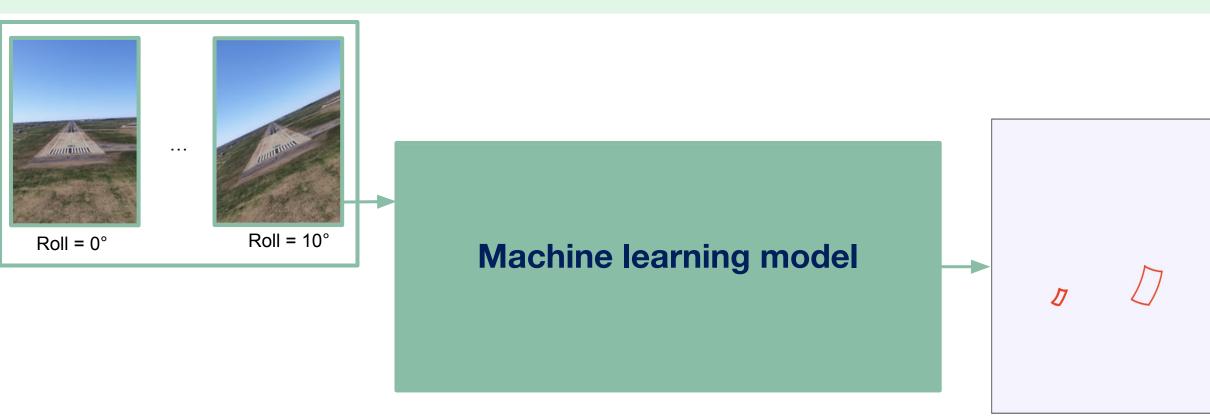
Formalized property



Verification tools

Formalized property

Mathematical traduction







Neural network



Property



Pre/Post-conditions

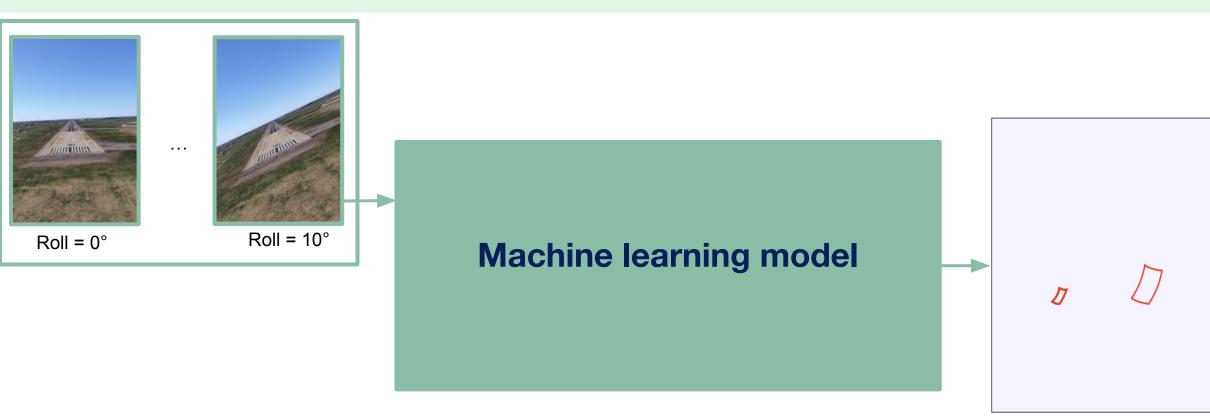


Formalized property



Verification tools

Mathematical traduction



What does it means mathematically : "The input meets the precondition"?





Neural network



Property



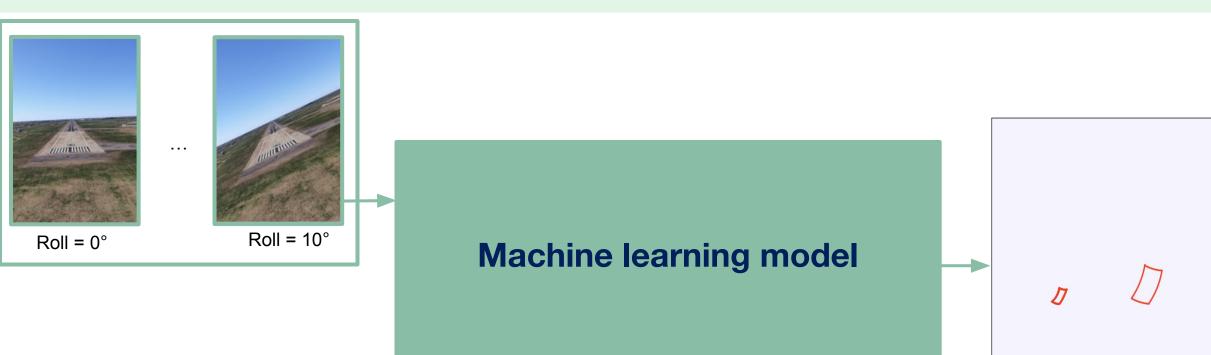
Pre/Post-conditions



Formalized property



Mathematical traduction



1	2	3
4	5	6
7	8	9

9 pixels image





Neural network



Property



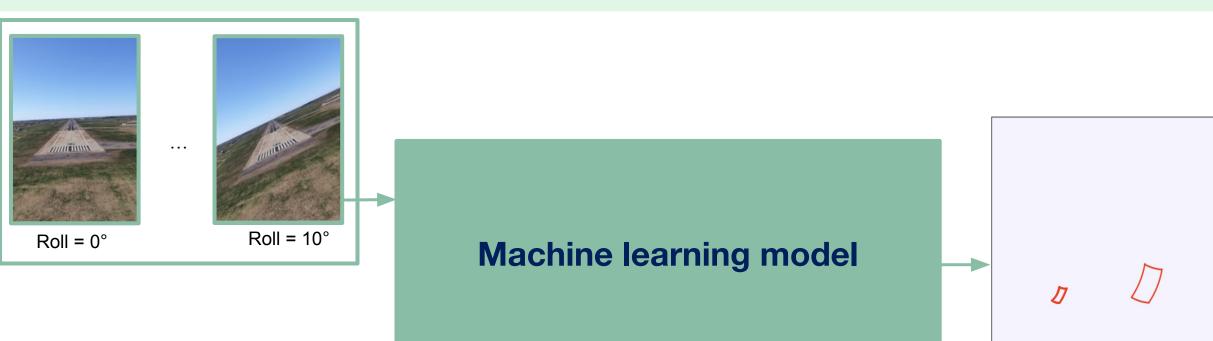
Pre/Post-conditions



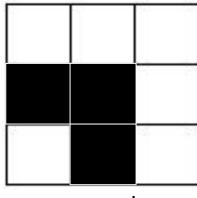
Formalized property



Mathematical traduction



1	2	3
4	5	6
7	8	9



example





Neural network



Property



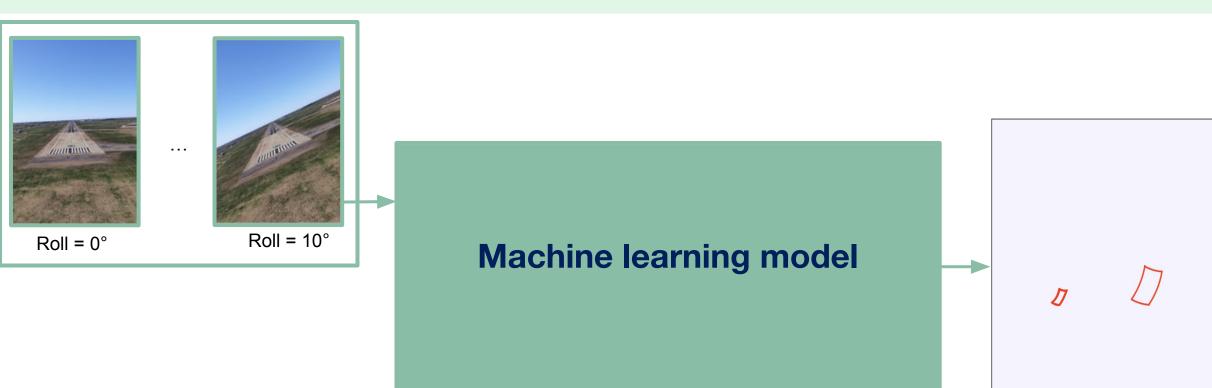
Pre/Post-conditions



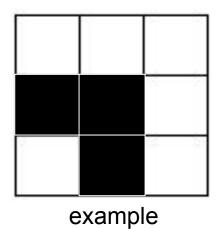
Formalized property



Mathematical traduction



1	2	3
4	5	6
7	8	9



1 2 3 4 5 6 7 8 9





Neural network



Property



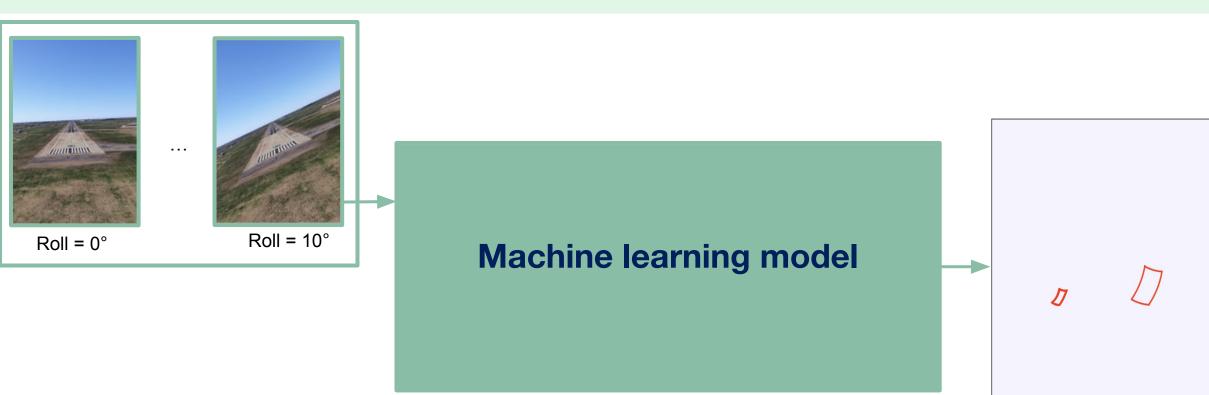
Pre/Post-conditions



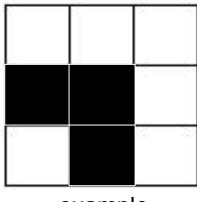
Formalized property



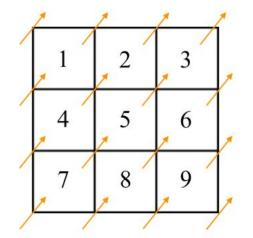
Mathematical traduction

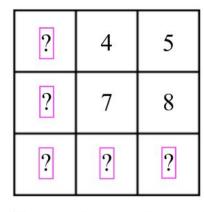


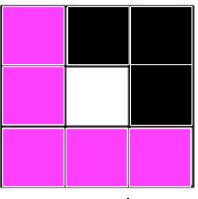
1	2	3
4	5	6
7	8	9



example







example





Neural network



Property



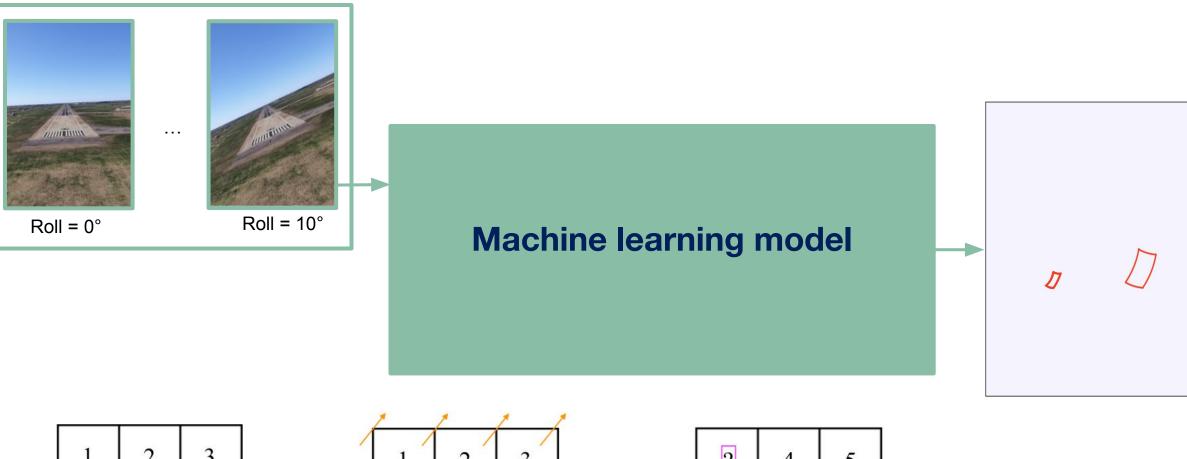
Pre/Post-conditions



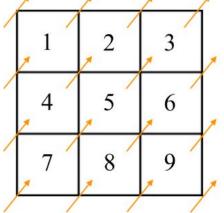
Formalized property



Mathematical traduction



1	2	3
4	5	6
7	8	9



?	4	5
?	7	8
?	?	?

5 pixels unknown in black and white





Neural network



Property



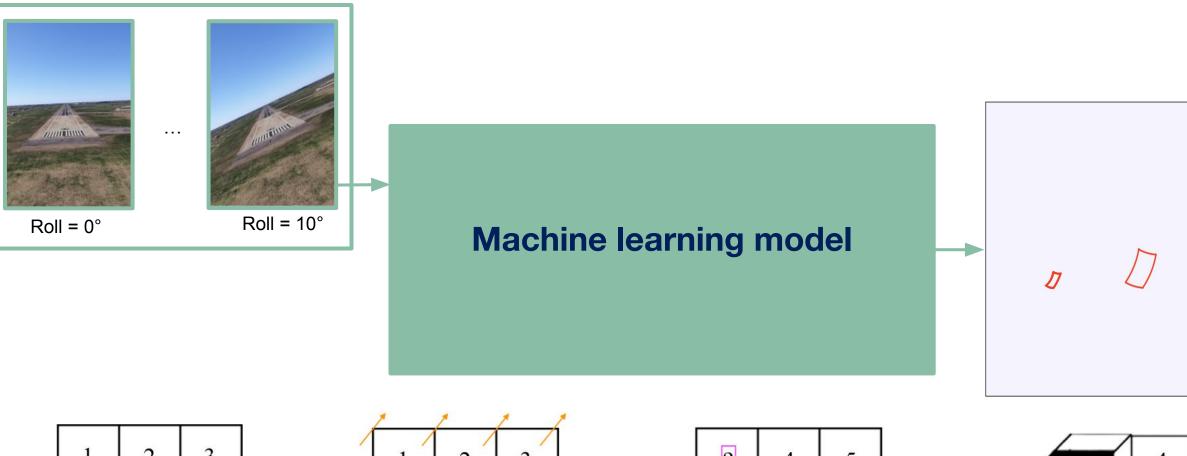
Pre/Post-conditions



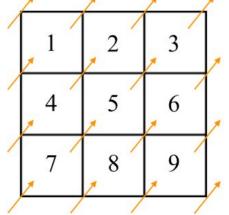
Formalized property

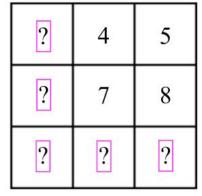


Mathematical traduction

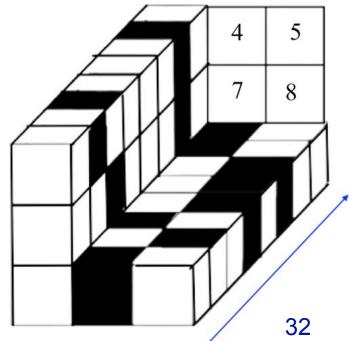


1	2	3
4	5	6
7	8	9





5 pixels unknown in black and white 32 combinations



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Neural network



Property



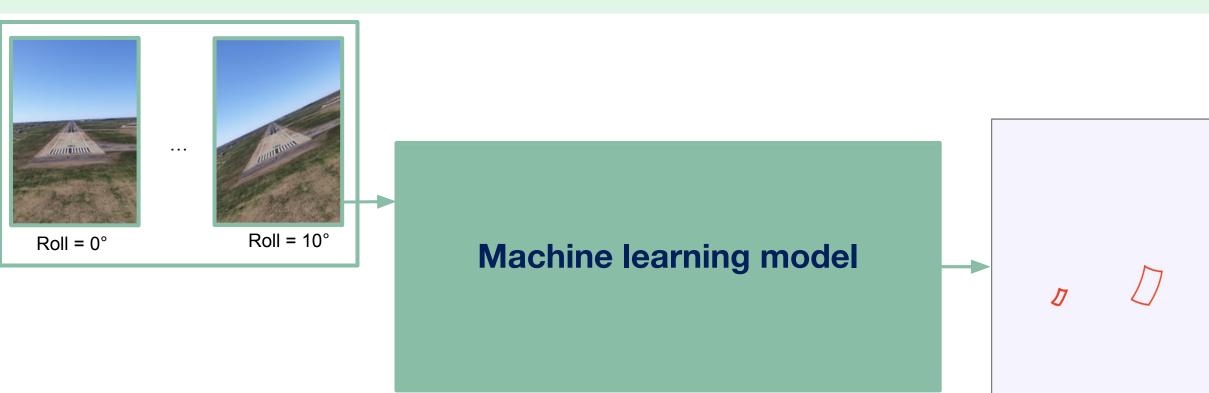
Pre/Post-conditions



Formalized property



Mathematical traduction









Neural network



Property



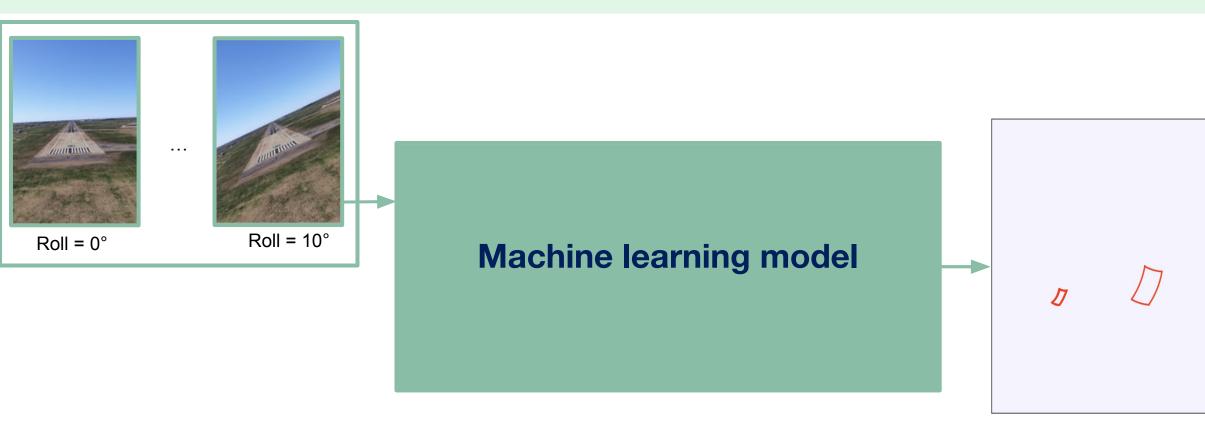
Pre/Post-conditions



Formalized property



Mathematical traduction













Neural network



Property



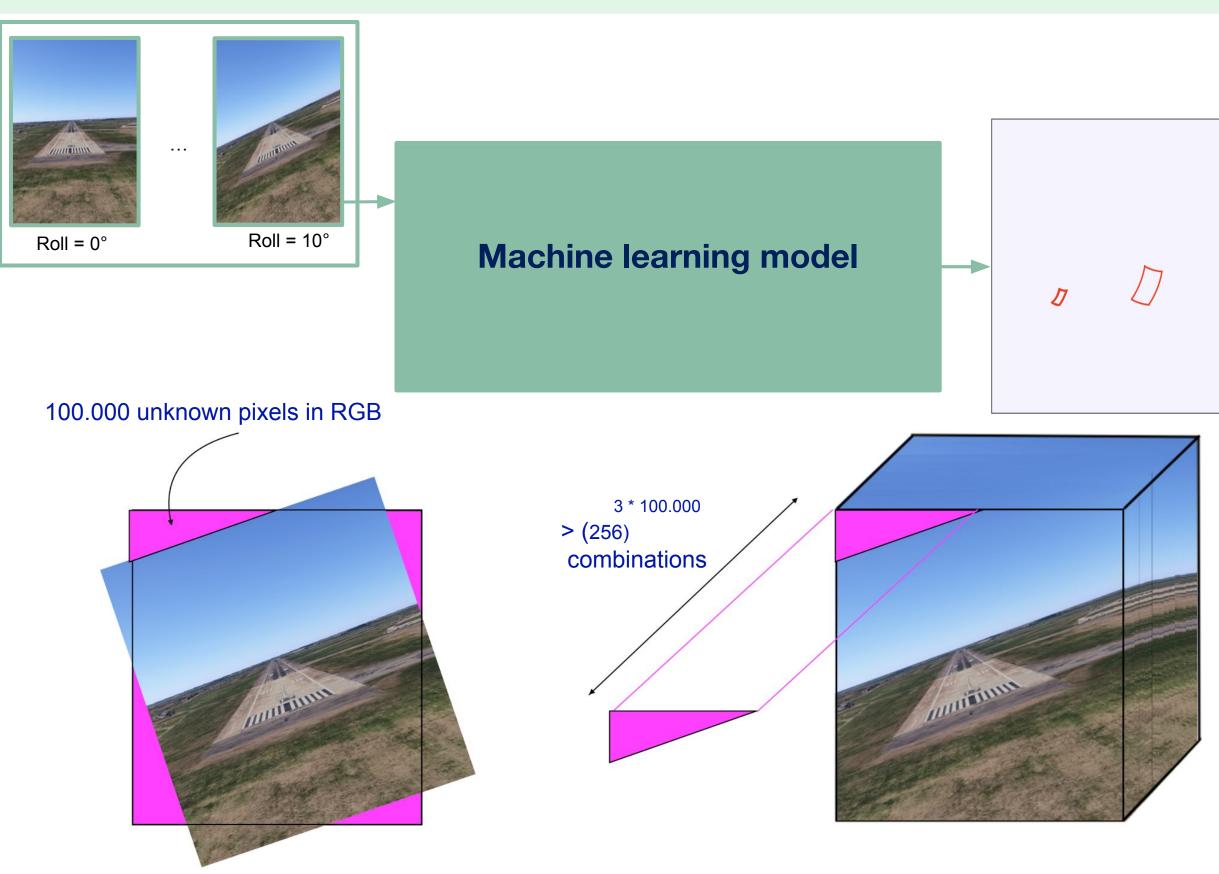
Pre/Post-conditions



Formalized property



Mathematical traduction







Neural network



Property



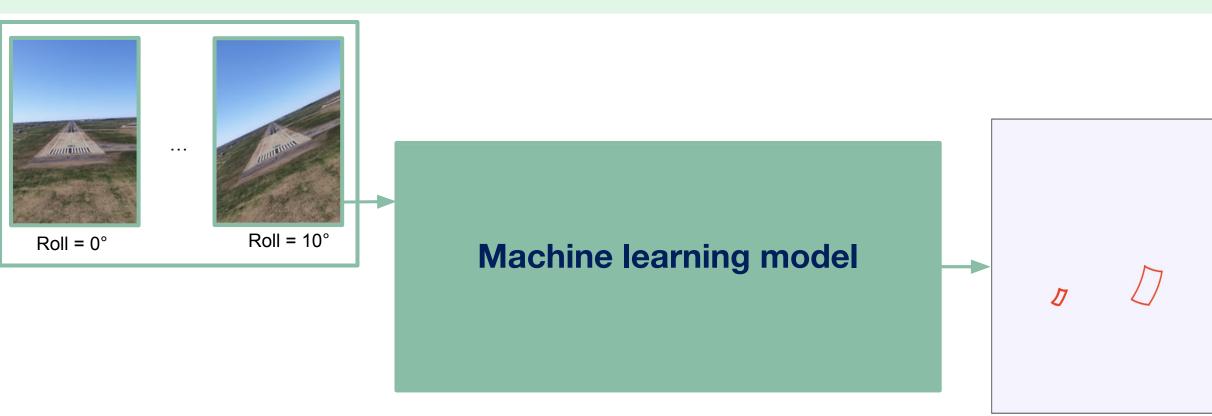
Pre/Post-conditions



Formalized property



Mathematical traduction



Find strategies to reduce dimensionality





Neural network



Property



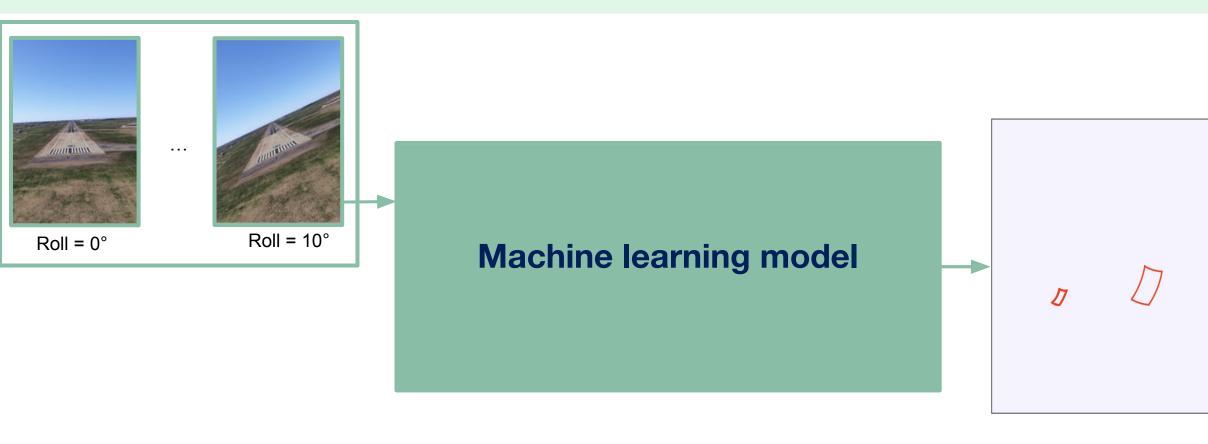
Pre/Post-conditions

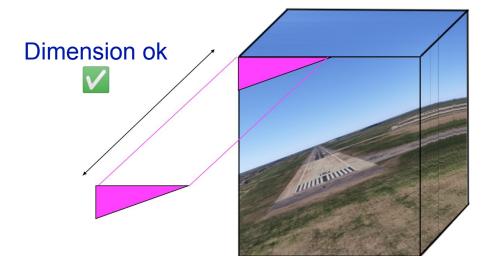


Formalized property



Mathematical traduction





Find strategies to reduce dimensionality





Neural network



Property



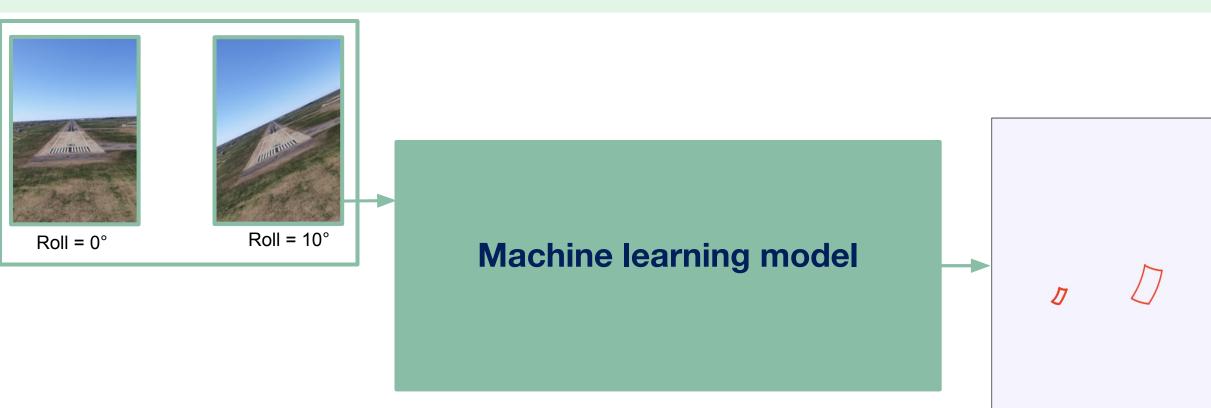
Pre/Post-conditions

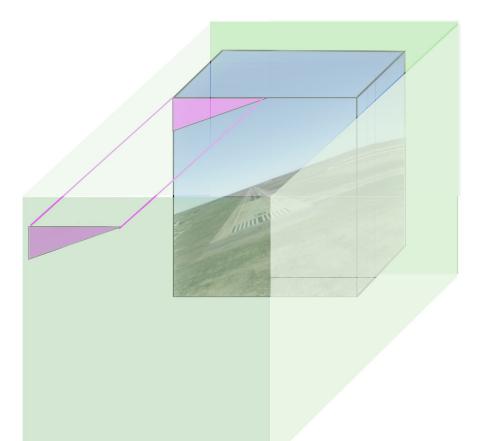


Formalized property



Mathematical traduction





Creating simplified representation that encompasses all possible cases

abstraction





Neural network



Property



Pre/Post-conditions



Formalized property





Automated validation software





Neural network



Property



Pre/Post-conditions

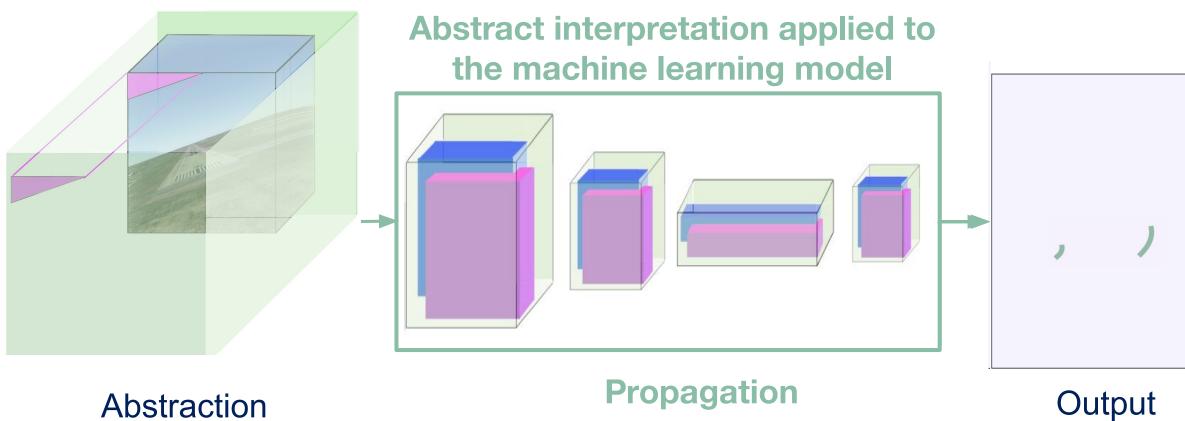


Formalized property





Automated validation software



Abstraction





Neural network



Property



Pre/Post-conditions



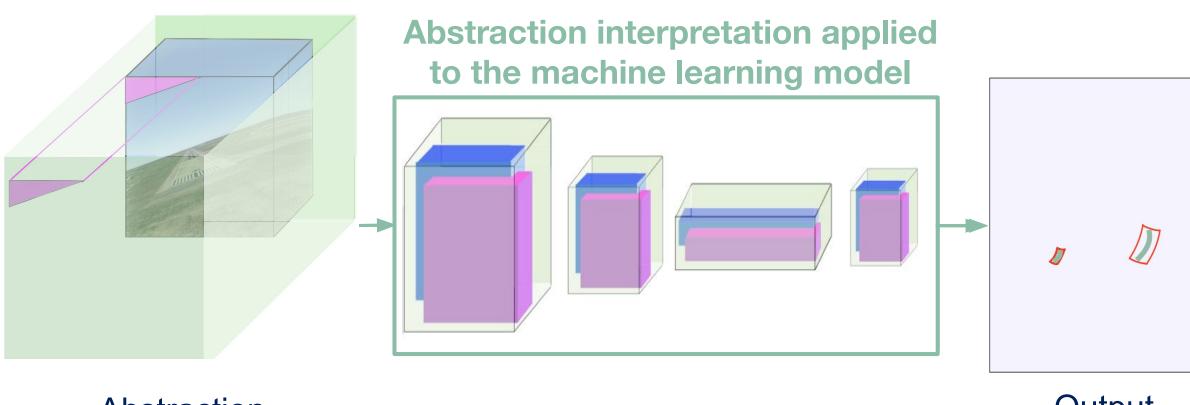
Formalized property



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Automated validation software



Abstraction

Output + postcondition





Neural network



Property



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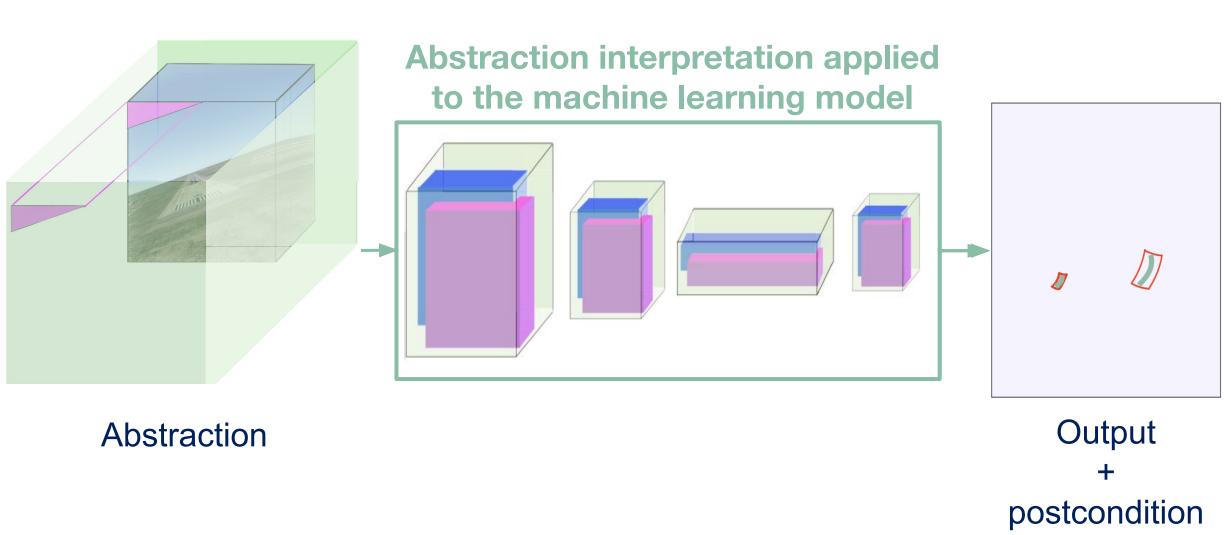
Pre/Post-conditions







Automated validation software



The postcondition is verified for the abstraction

Verified property





Neural network



Property



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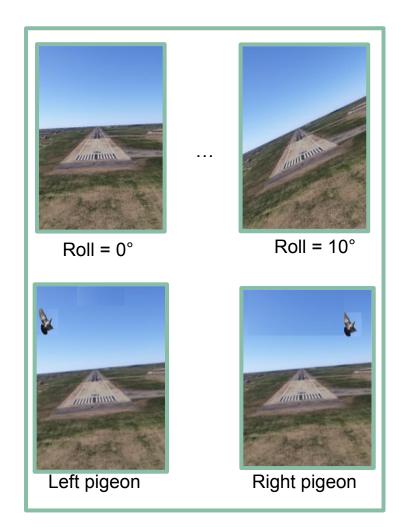
Pre/Post-conditions







Not only is the roll set

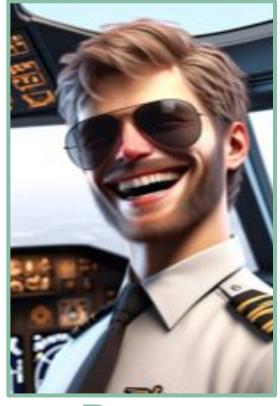








Find properties



Danny







Find properties

Abstraction

14 11 2023





Not only is the roll set

Find properties

Abstraction



Thank you for your attention

Noémie Cohen

Supervisor: Claire Pagetti, Xavier Pucel ONERA Christophe Gabreau, Mélanie Ducoffe AIRBUS







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