

The background of the slide is a dark, blue-tinted photograph of an industrial setting. In the foreground, several yellow robotic arms are visible, likely part of a manufacturing or assembly line. To the right, a person is partially visible, wearing a dark jacket and looking down at a laptop screen. The overall atmosphere is technical and modern.

DATA.ZERO

WP2 Training Seminar

3D Characterisation of cylindrical runout

Luke Todhunter



DAT4.Zero has received funding from the European Union's Horizon 2020 Research and Innovation programme under Grant Agreement No. **958363**

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Runout optimized pressing of head shaft and chuck[A2.3]

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Goal / motivation

- Runout oriented assembly
- Different parts -both with runout from 0 to 0,01mm- become one subassembly with maximum runout of 0,01mm
- Zero defect assembly

Status quo

- [...]
- Only similarity: measuring of imbalance

Details

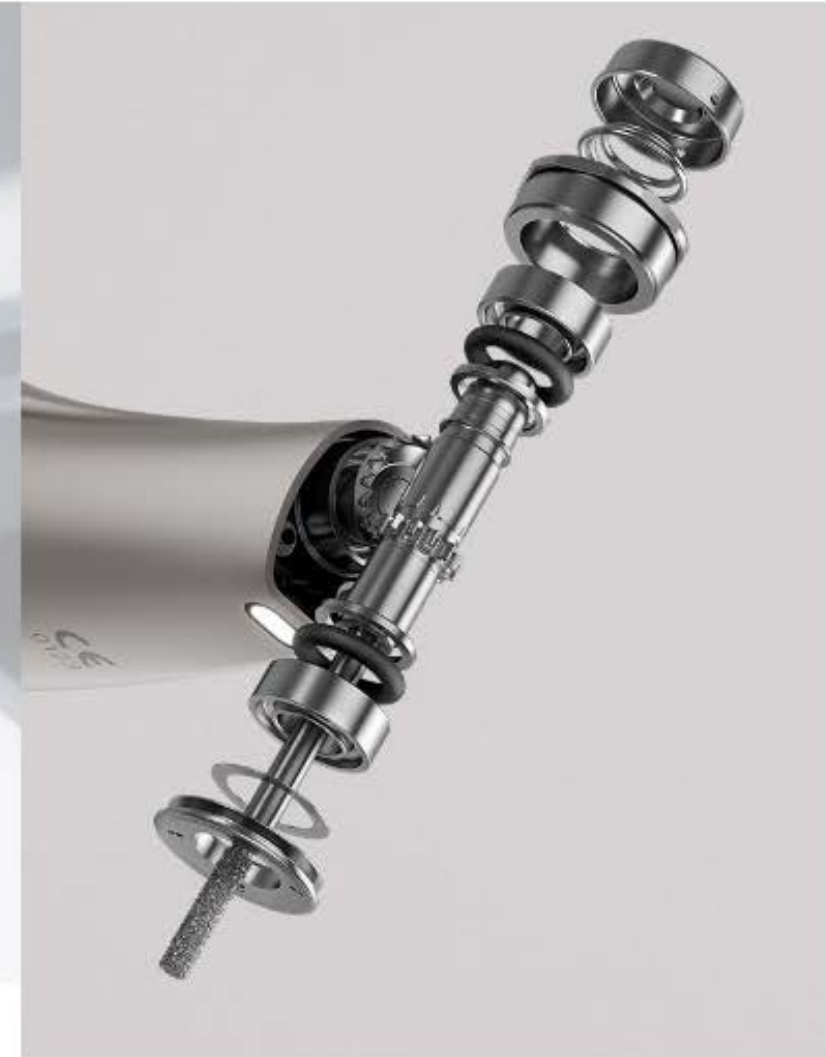
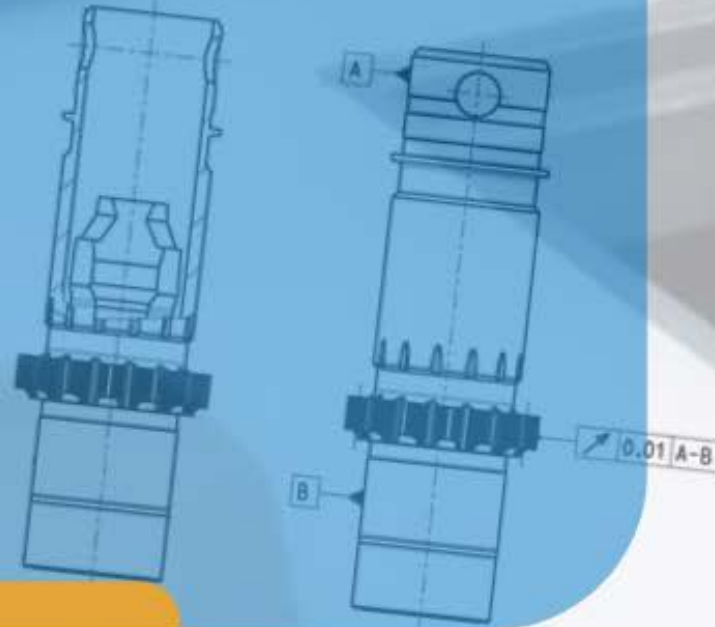
- [...]

Challenges

- Multi references of different parts
- Drill impact vs runout of gears
- Three-dimensional wobble
- Use of the data for assembly

Data

- Q-DAS connection
- Use in automation



Context

- User story 4a - Dentsply
 - Optimise assembly to minimise overall runout
- Requirement 1:
 - A good understanding of the angular direction of runout error

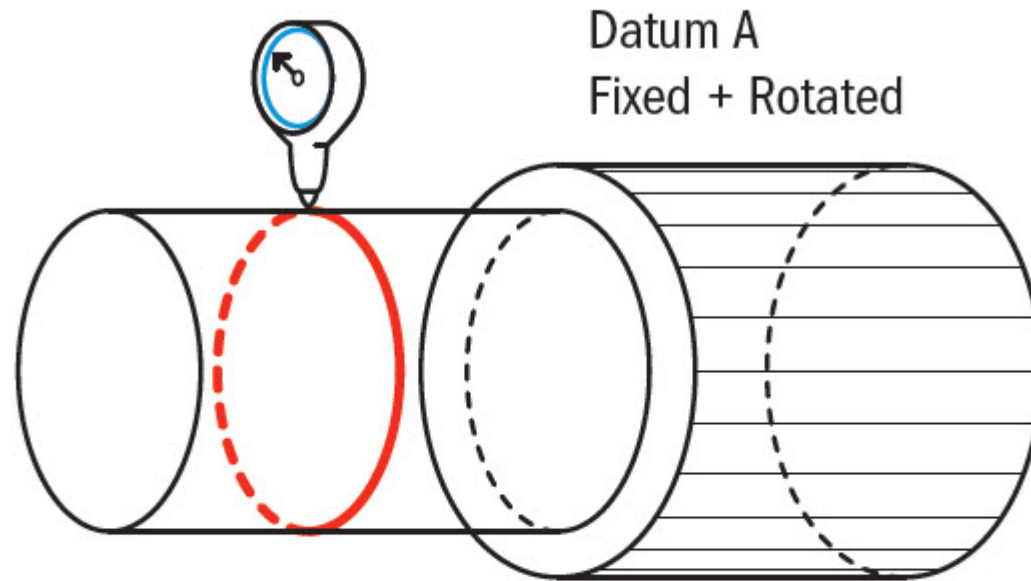


Context

- User story 4a - Dentsply
 - Optimise assembly to minimise overall runout
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 - A good understanding of the angular direction of runout error



Traditional runout

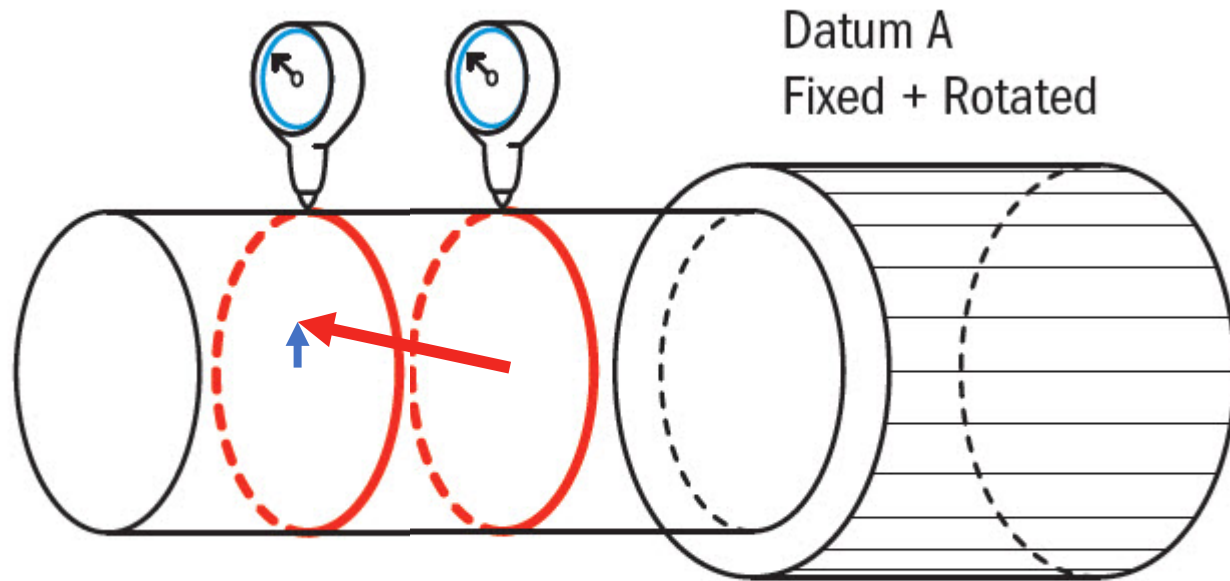


Normal/Circular runout
only checks individual sections
independent of each other

A singular magnitude – **No directional information!**

Mainly identifies **Roundness variation**

3D axial runout



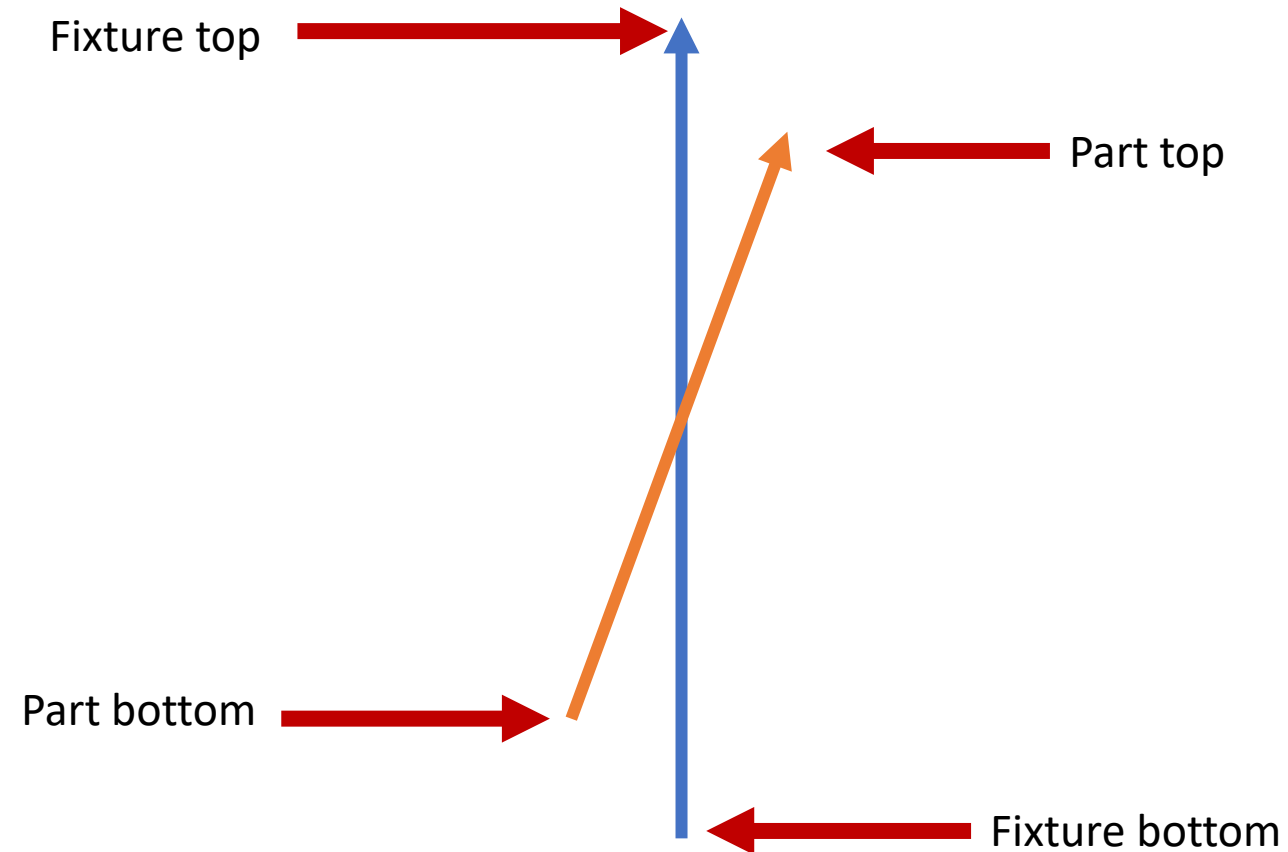
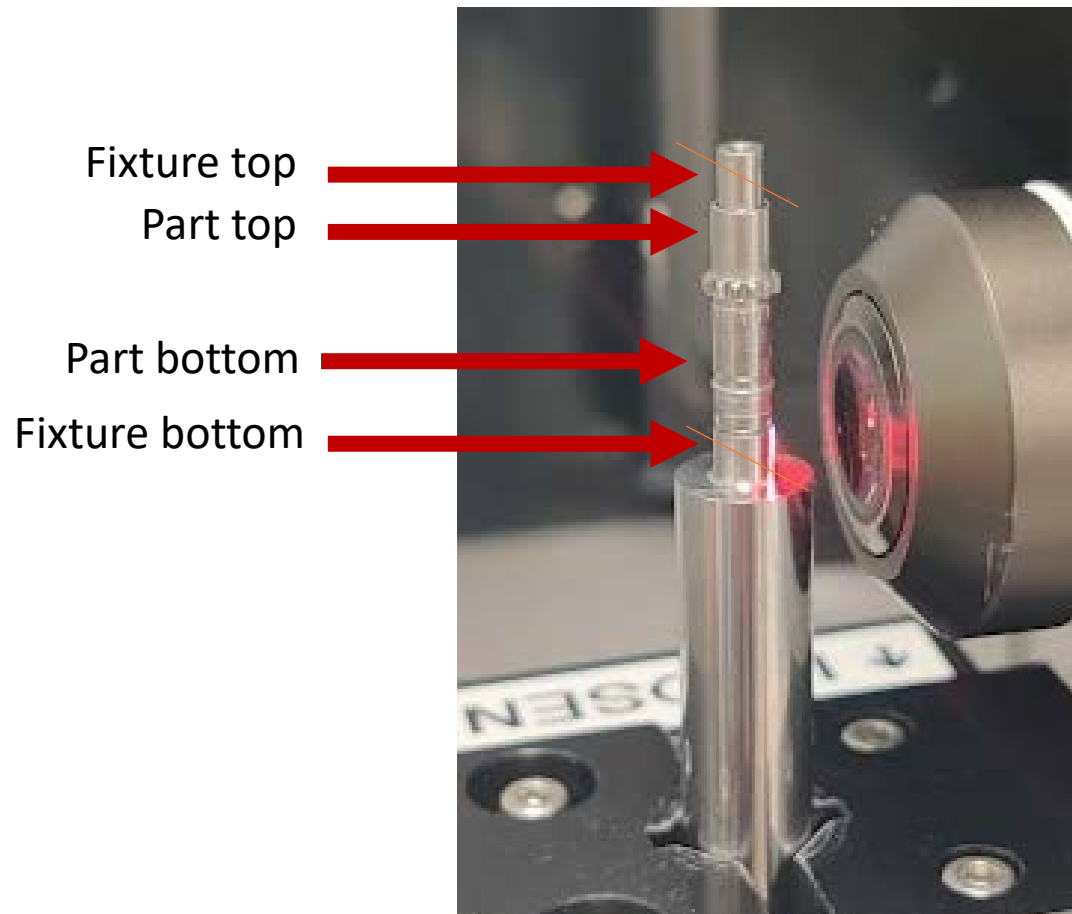
Normal/Circular runout
only checks individual sections
independent of each other

Aim: Magnitude and direction
- **Runout Vector**

Mainly identifies **Axial**
variation

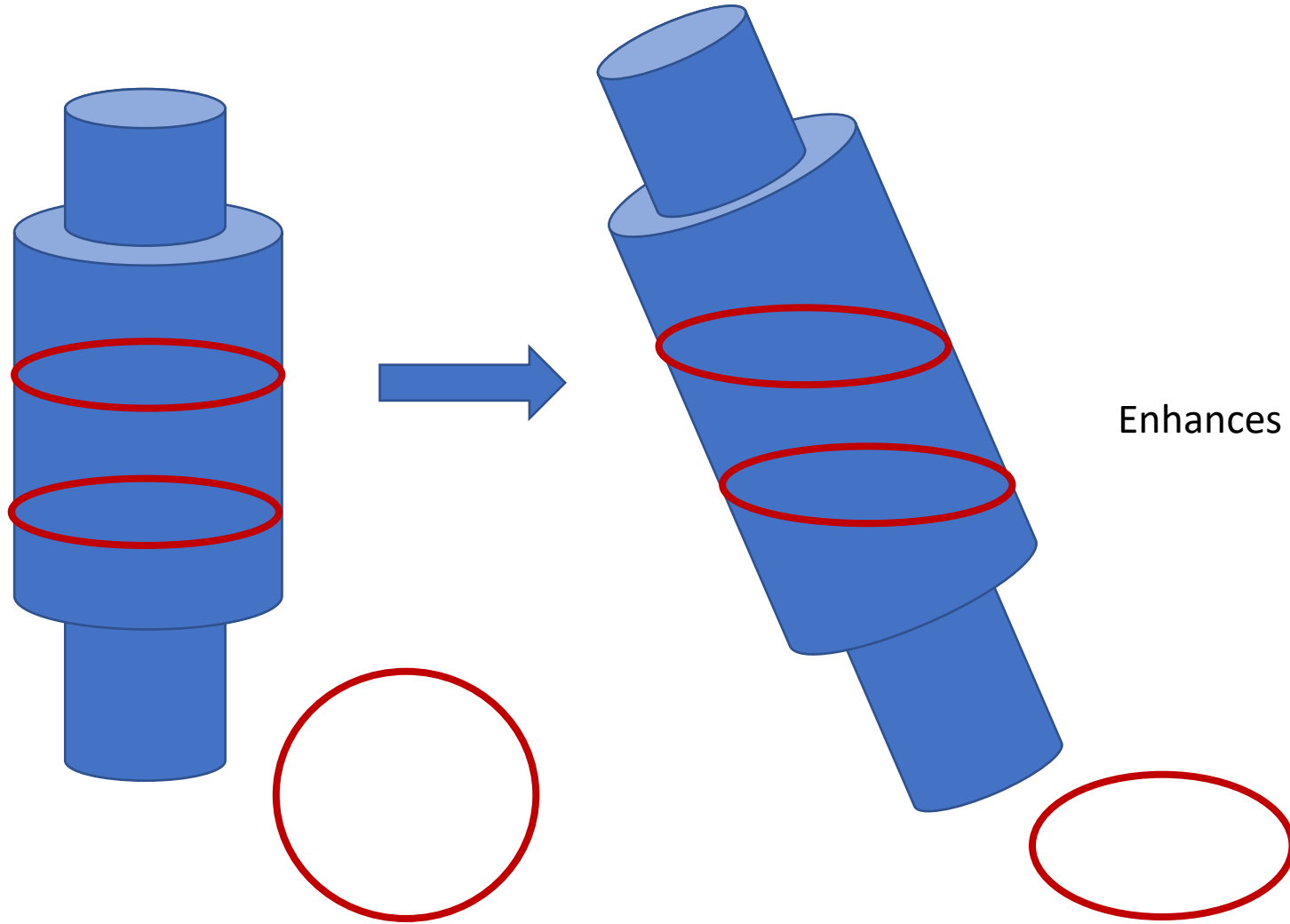
Axial runout - Process

Capture measurement data that contains axis information

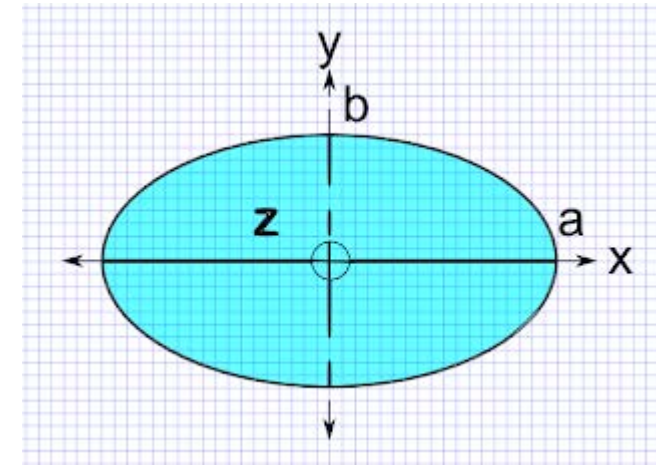
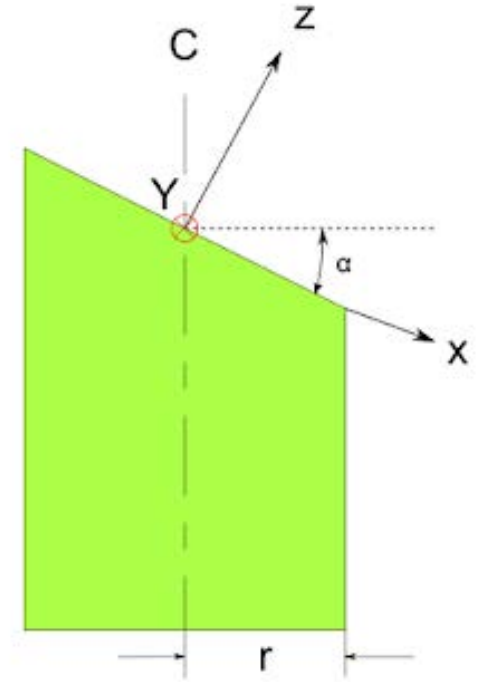


Elliptical fitting

A better method for characterising relative angle

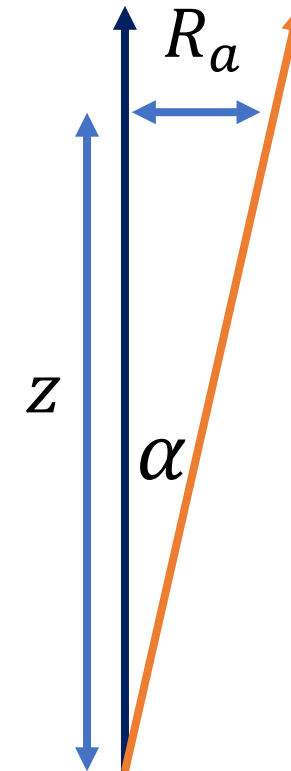
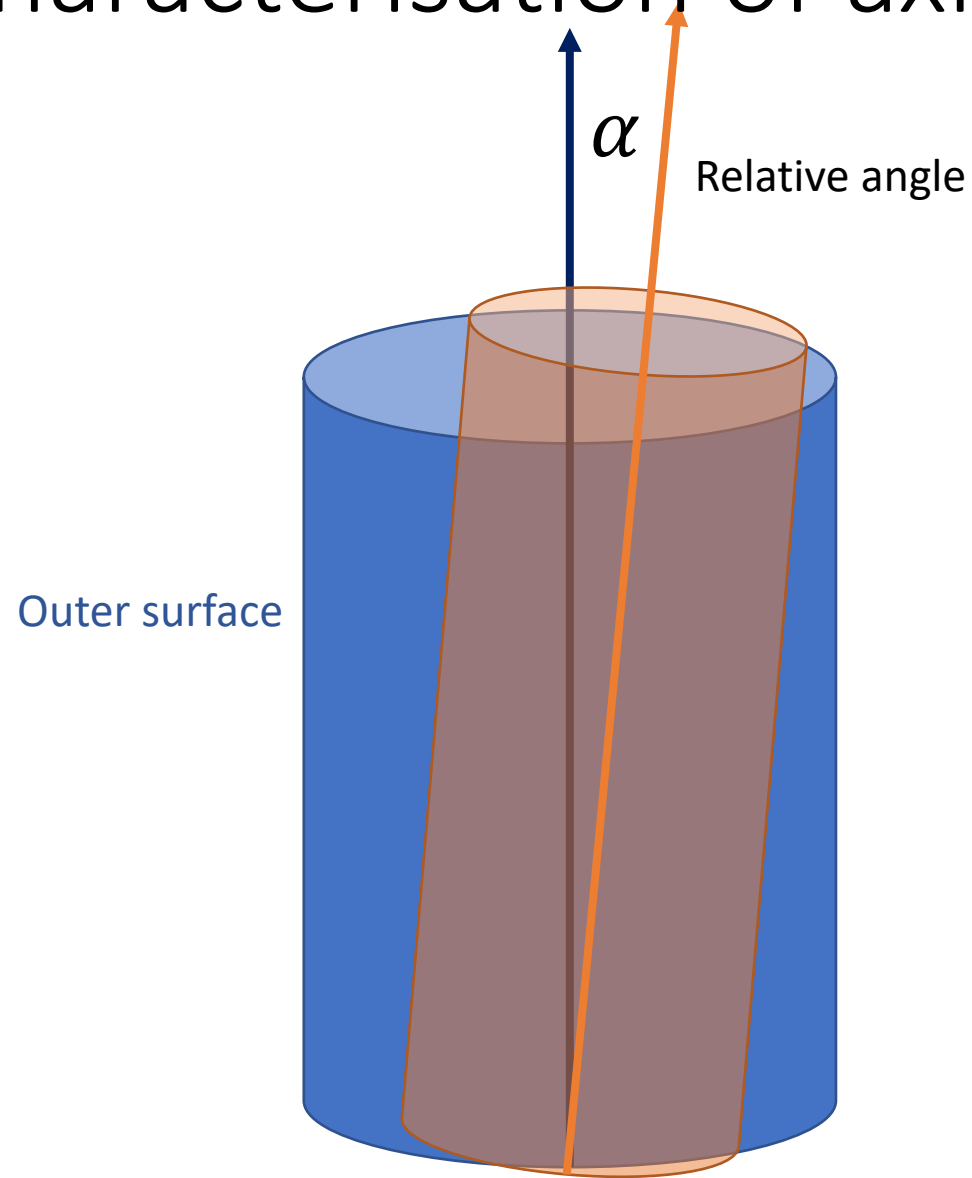


Enhances calculation of axial tilt



Characterisation of axial runout

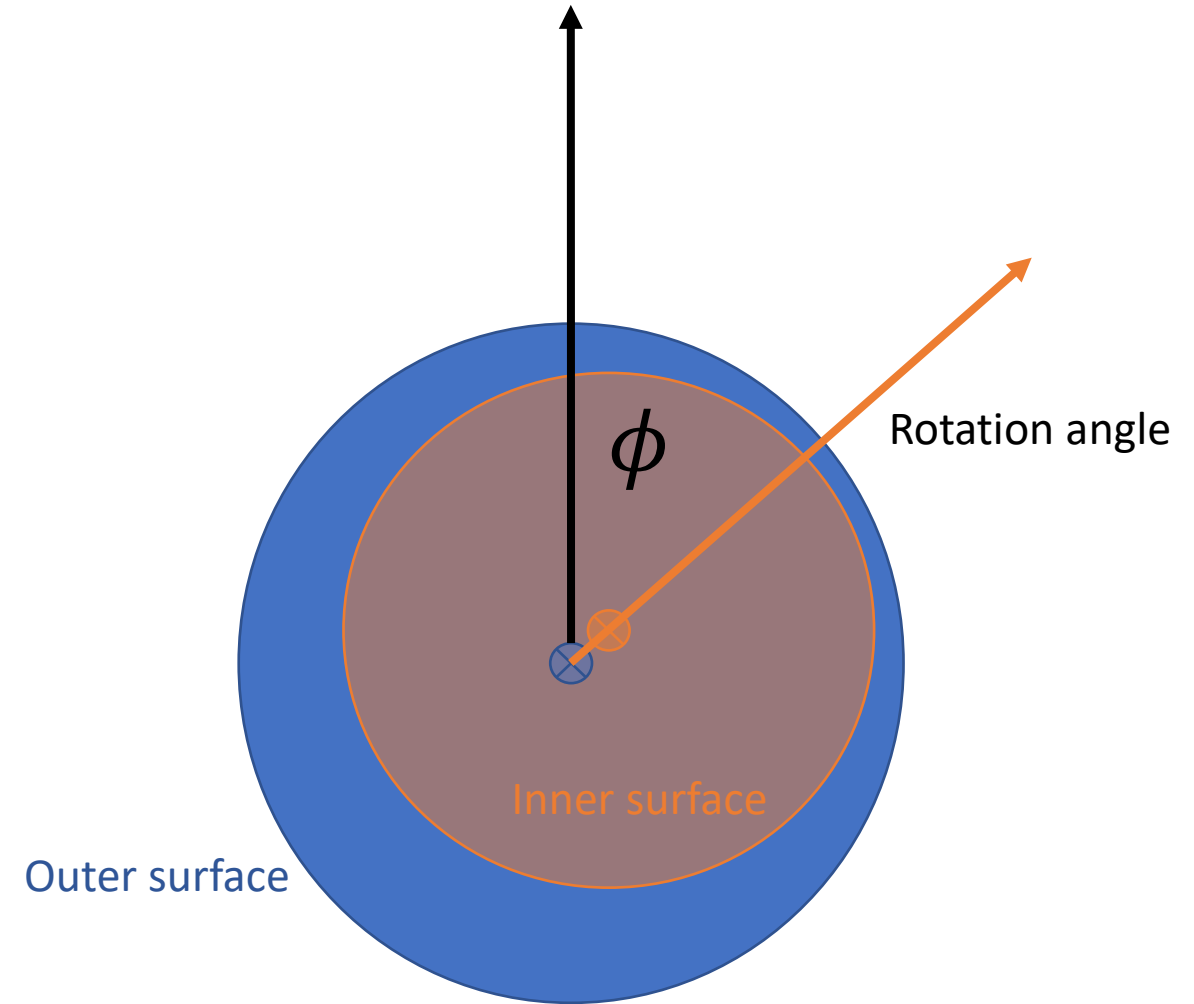
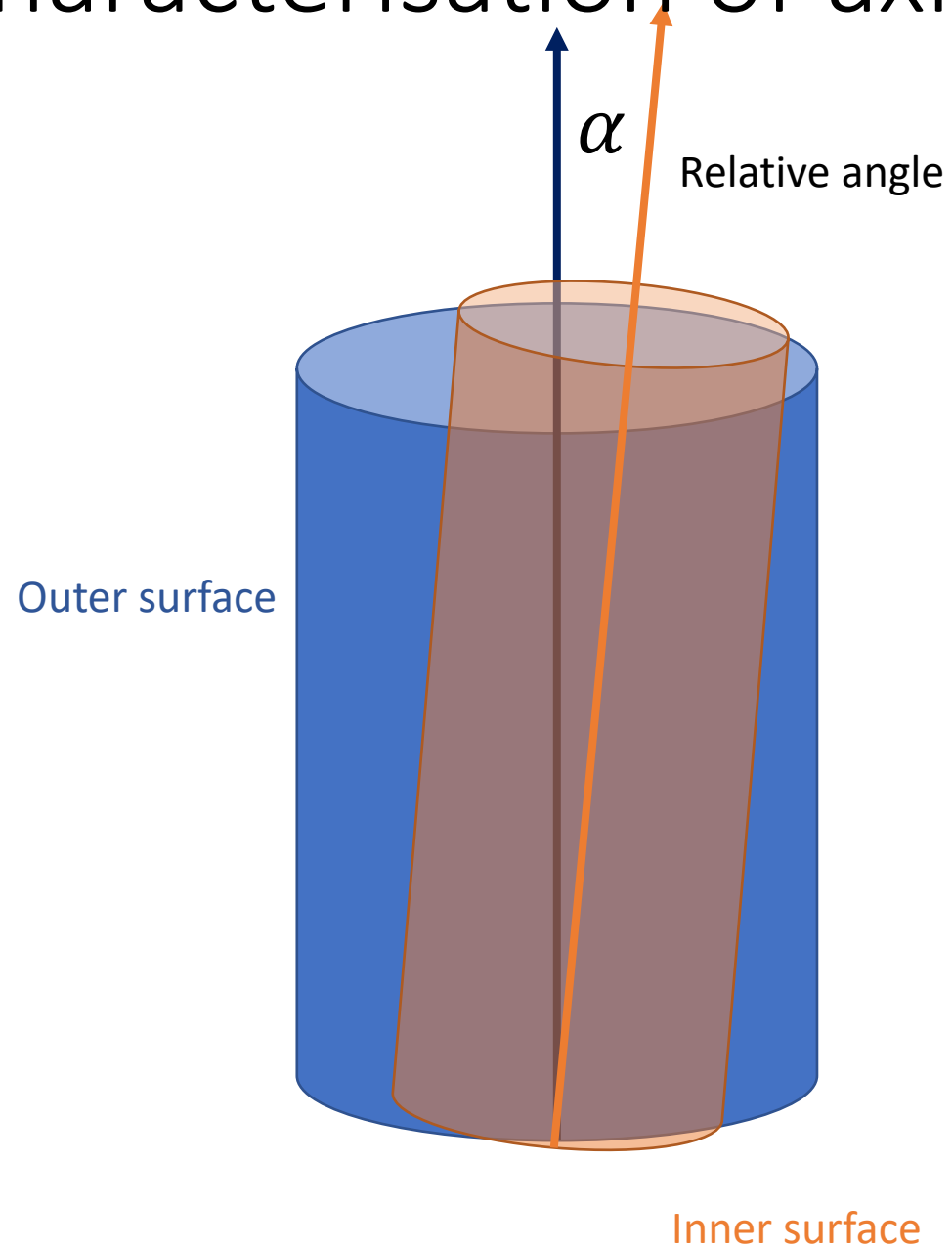
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Inner surface *Axial runout magnitude $R_a = z \tan \alpha$*

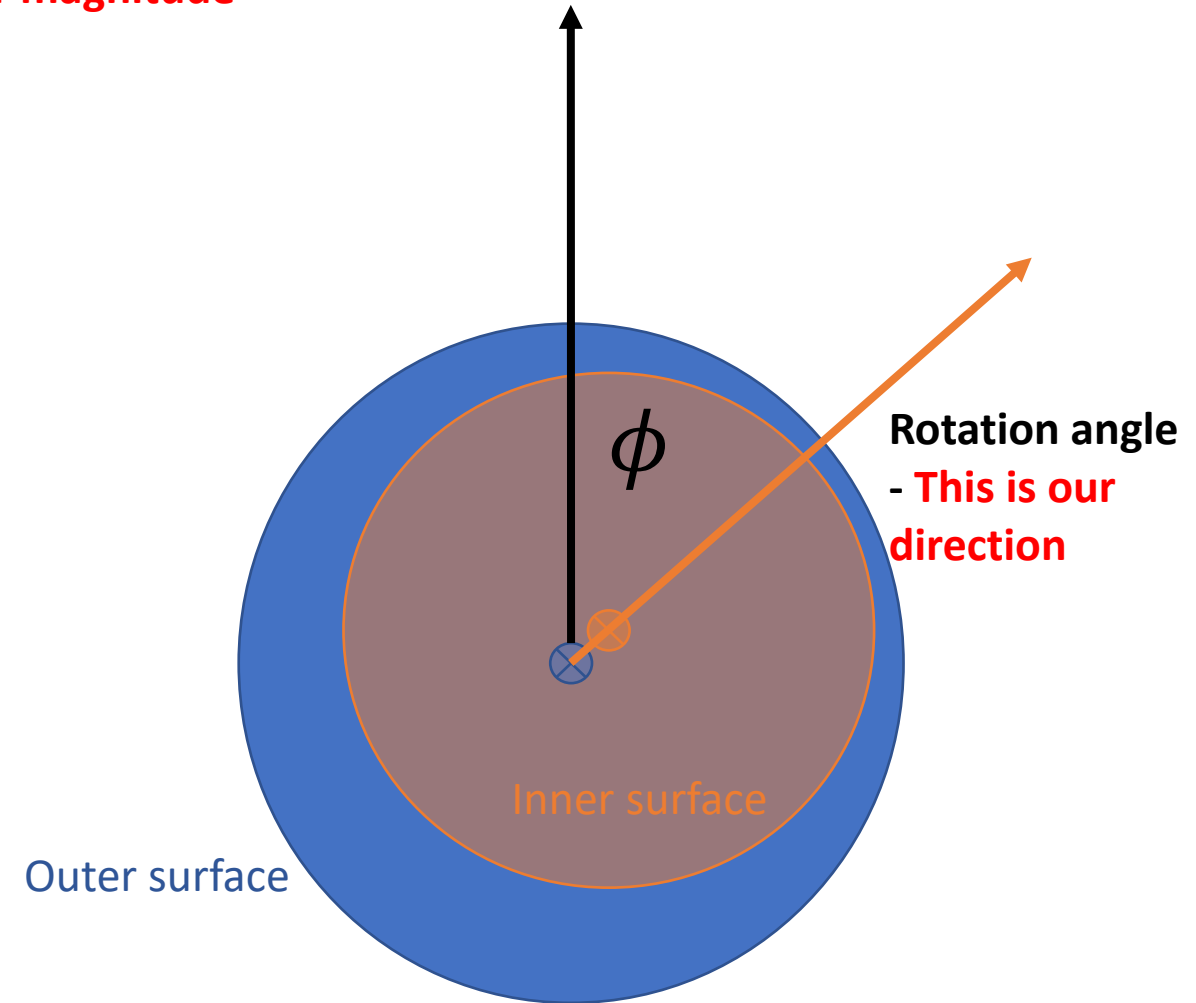
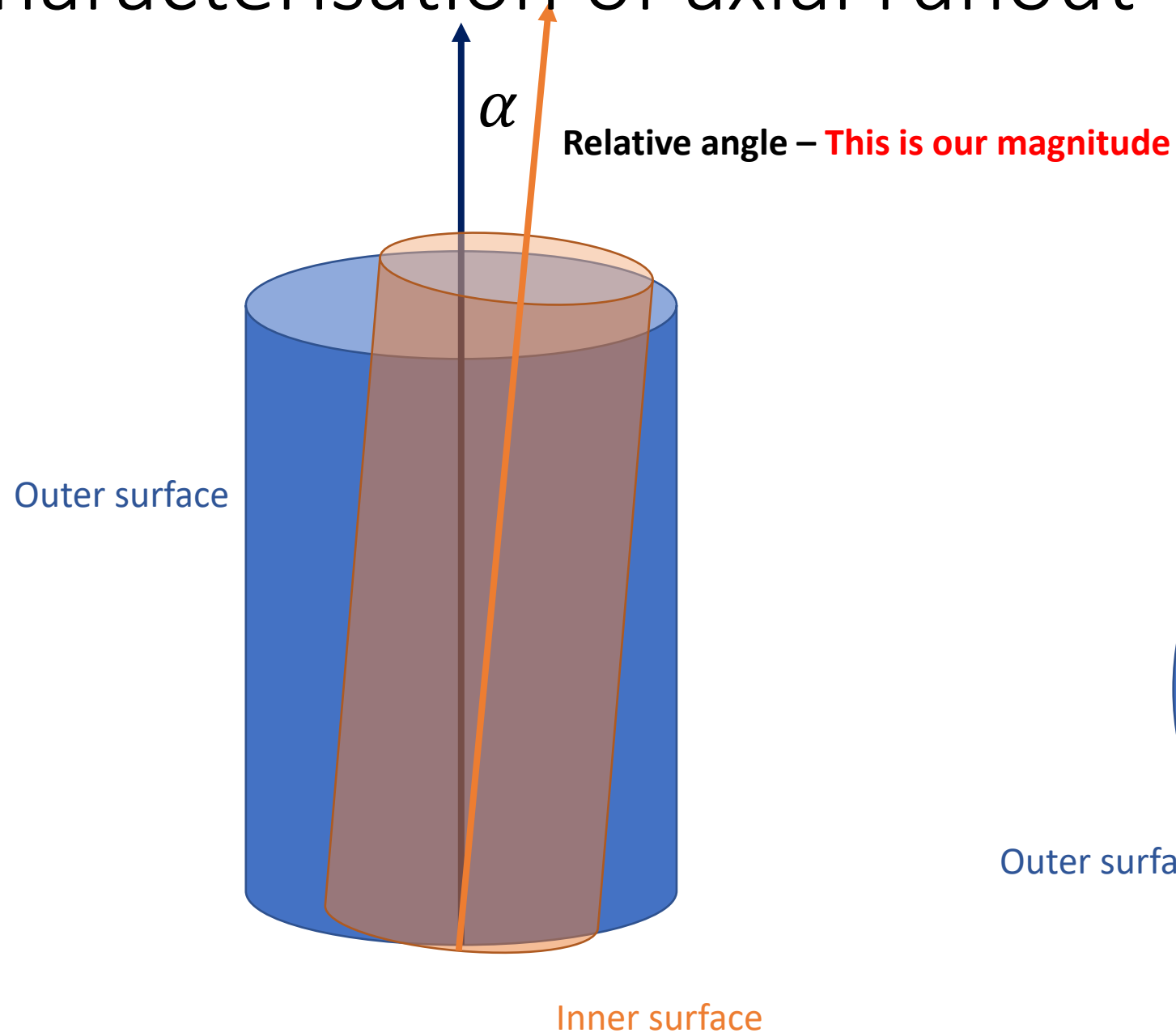
Characterisation of axial runout

DATA.ZERO



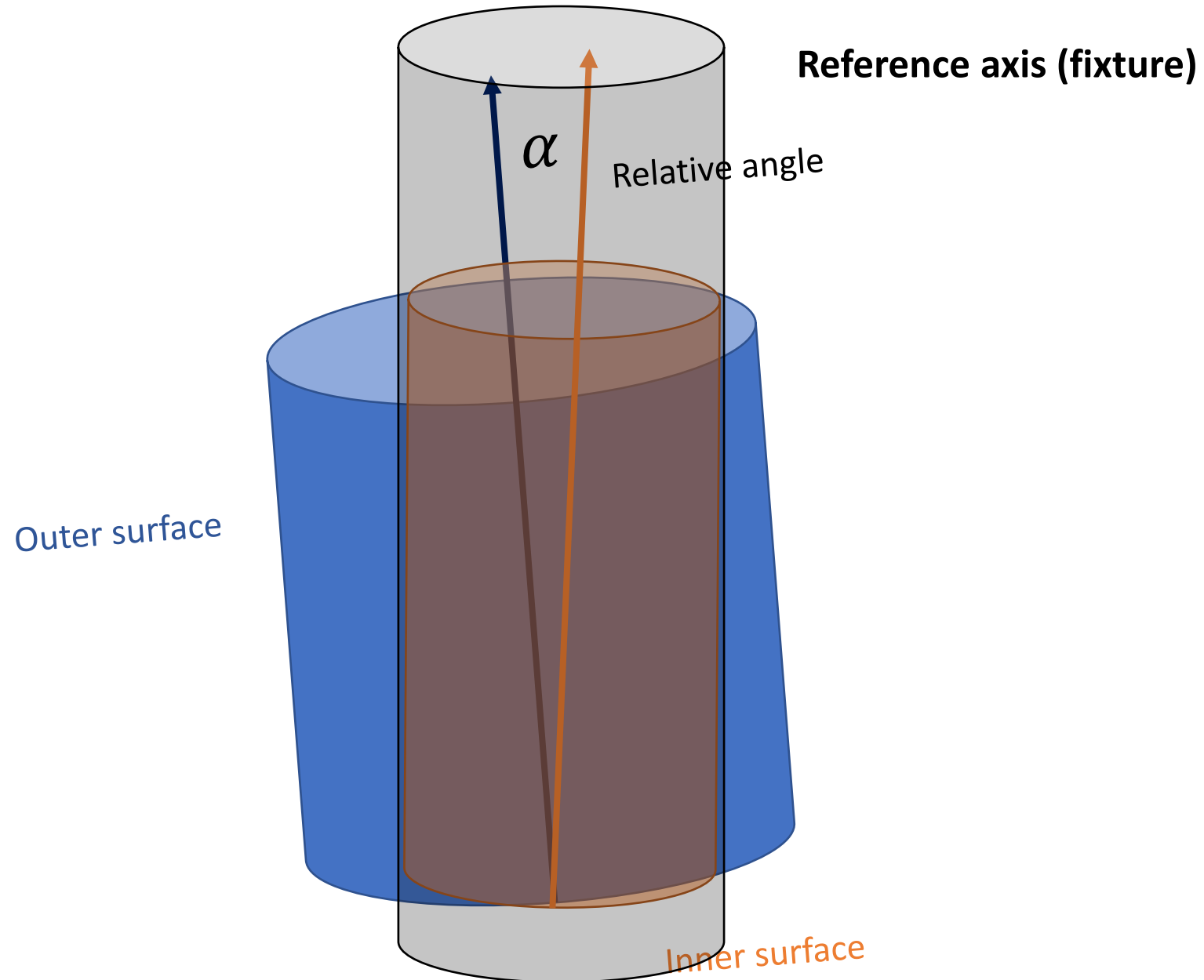
Characterisation of axial runout

DATA.ZERO



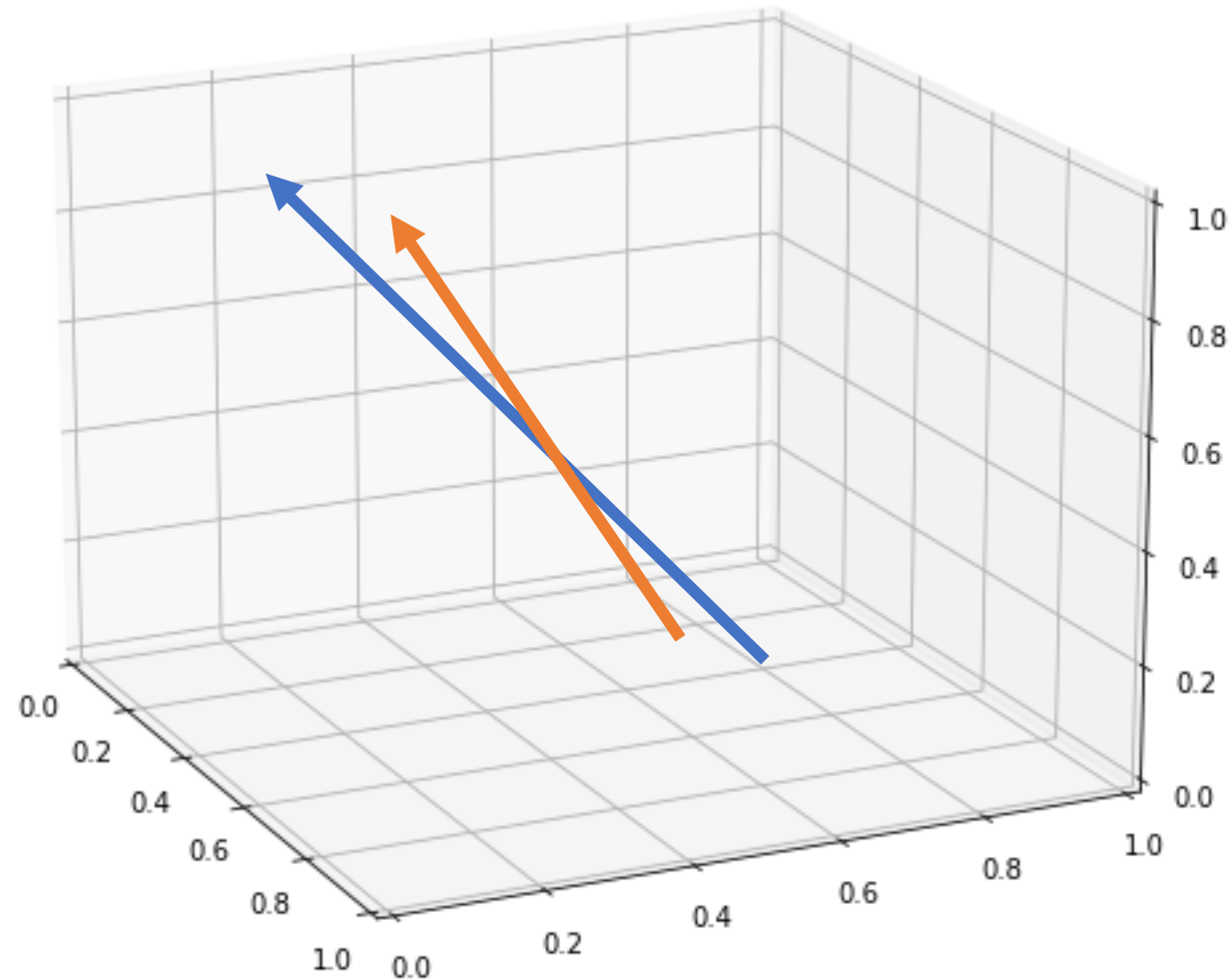
Characterisation of axial runout

DATA.ZERO



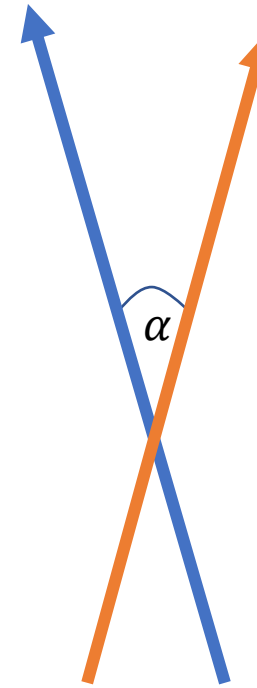
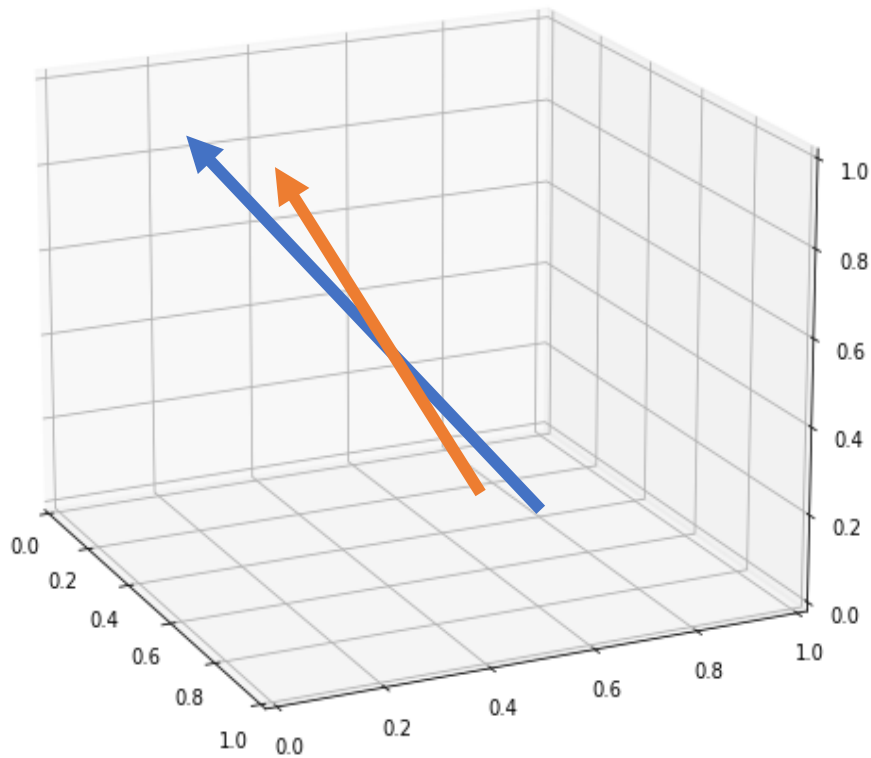
Axial runout - Process

Define vectors for each measurement pair (reference and part)



Axial runout - Process

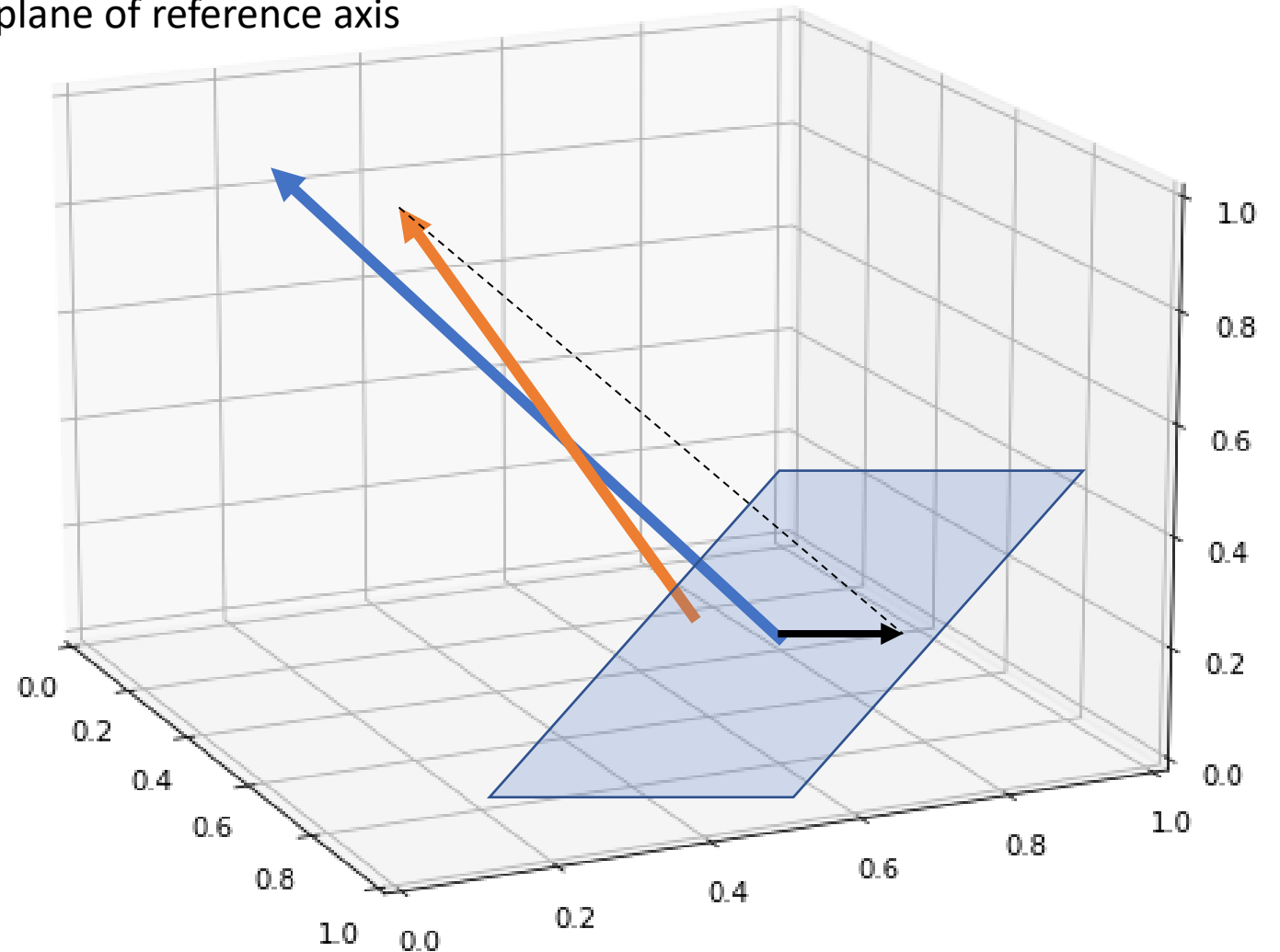
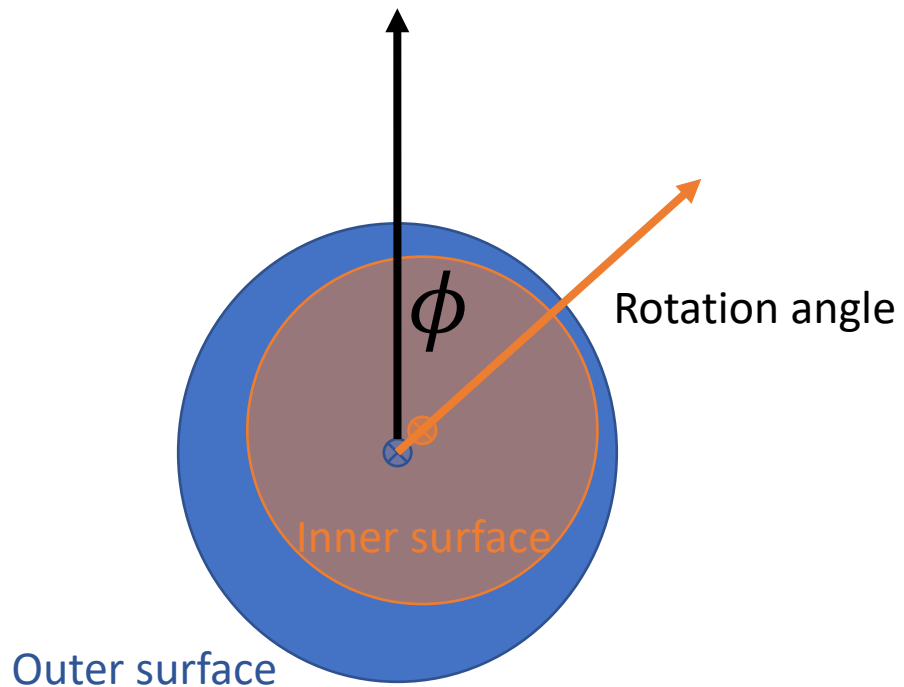
Relative angle: Calculate angle between vectors



$$\alpha = \arccos \frac{a \cdot b}{|a||b|}$$

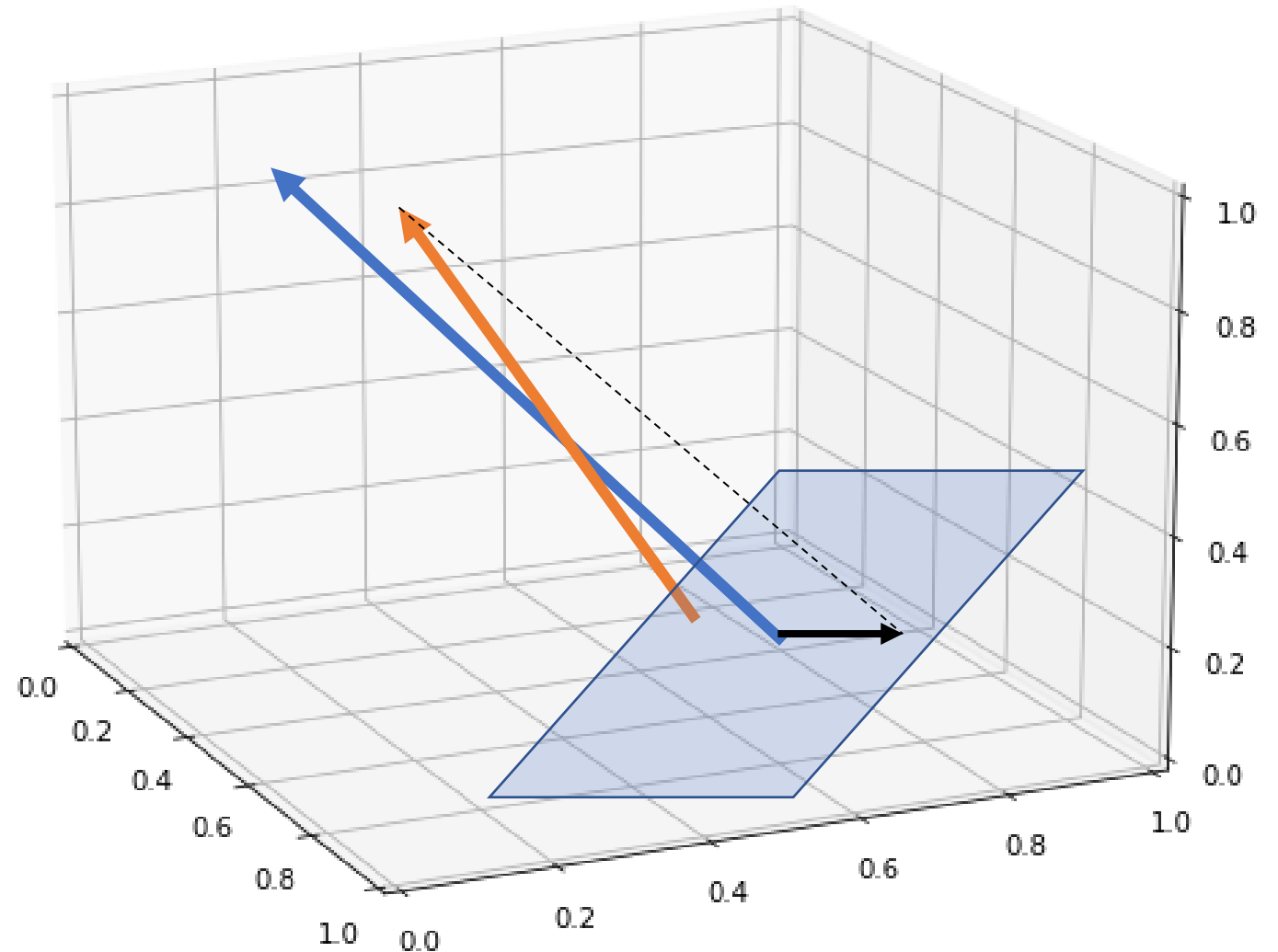
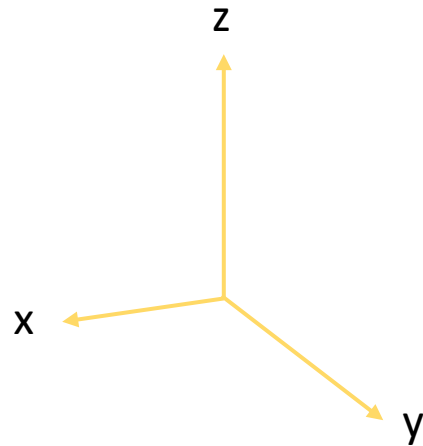
Axial runout - Process

Rotation angle: Project part axis onto normal plane of reference axis



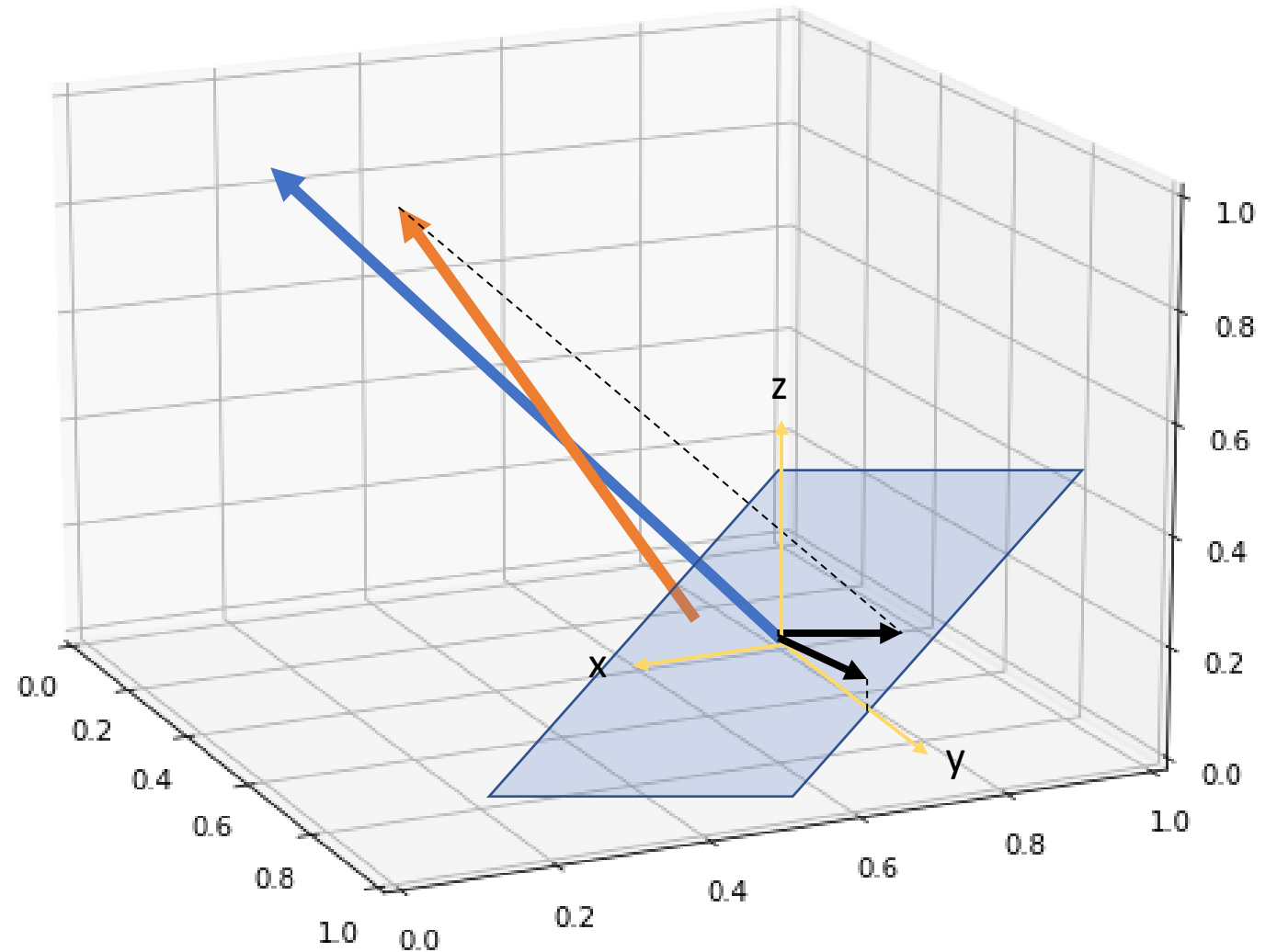
Axial runout - Process

Rotation angle: Calculate angle between part projection and y-axis projection (ensuring consistent direction)



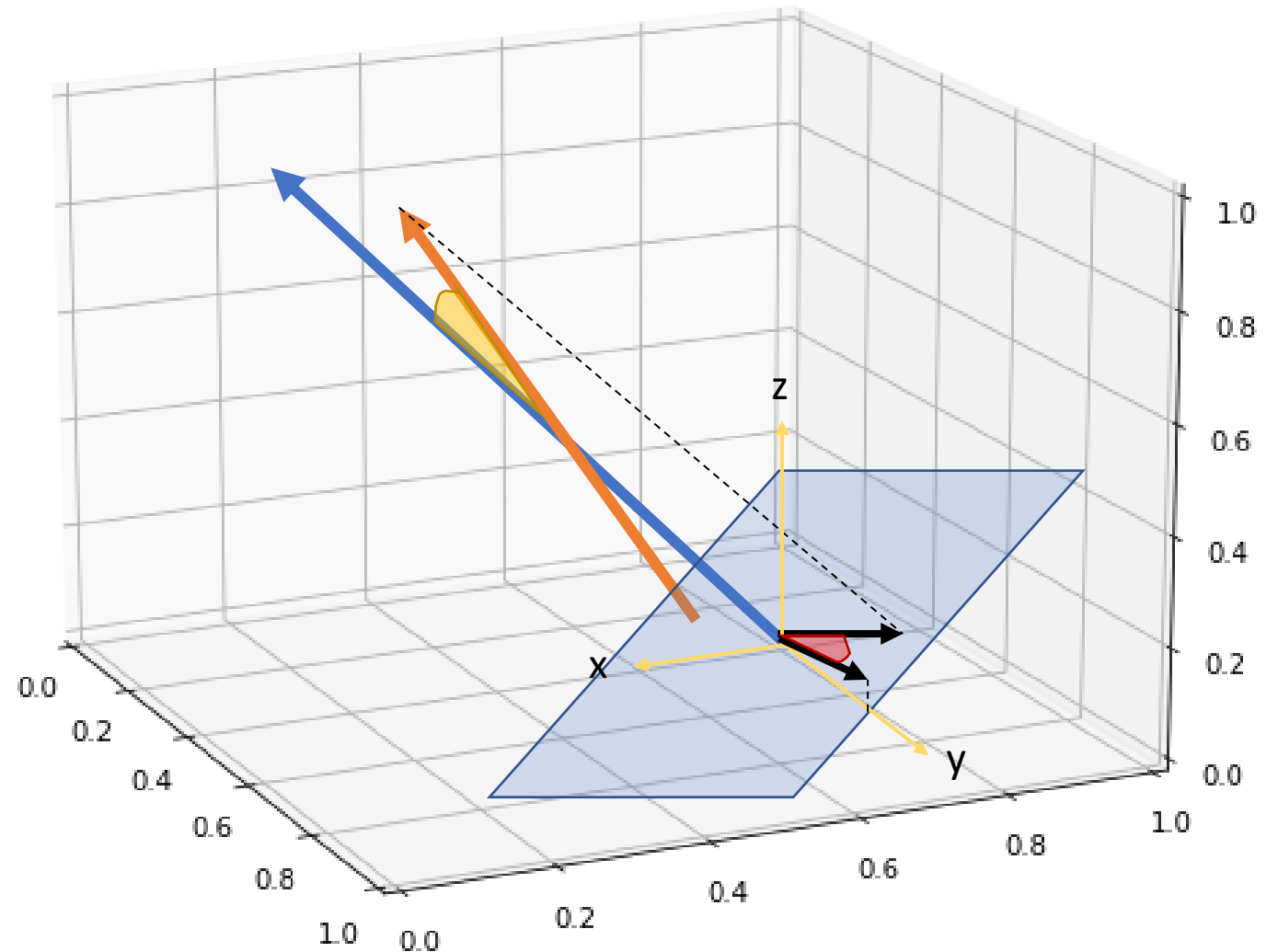
Axial runout - Process

Rotation angle: Calculate angle between part projection and y-axis projection (ensuring consistent direction)



Axial runout - Process

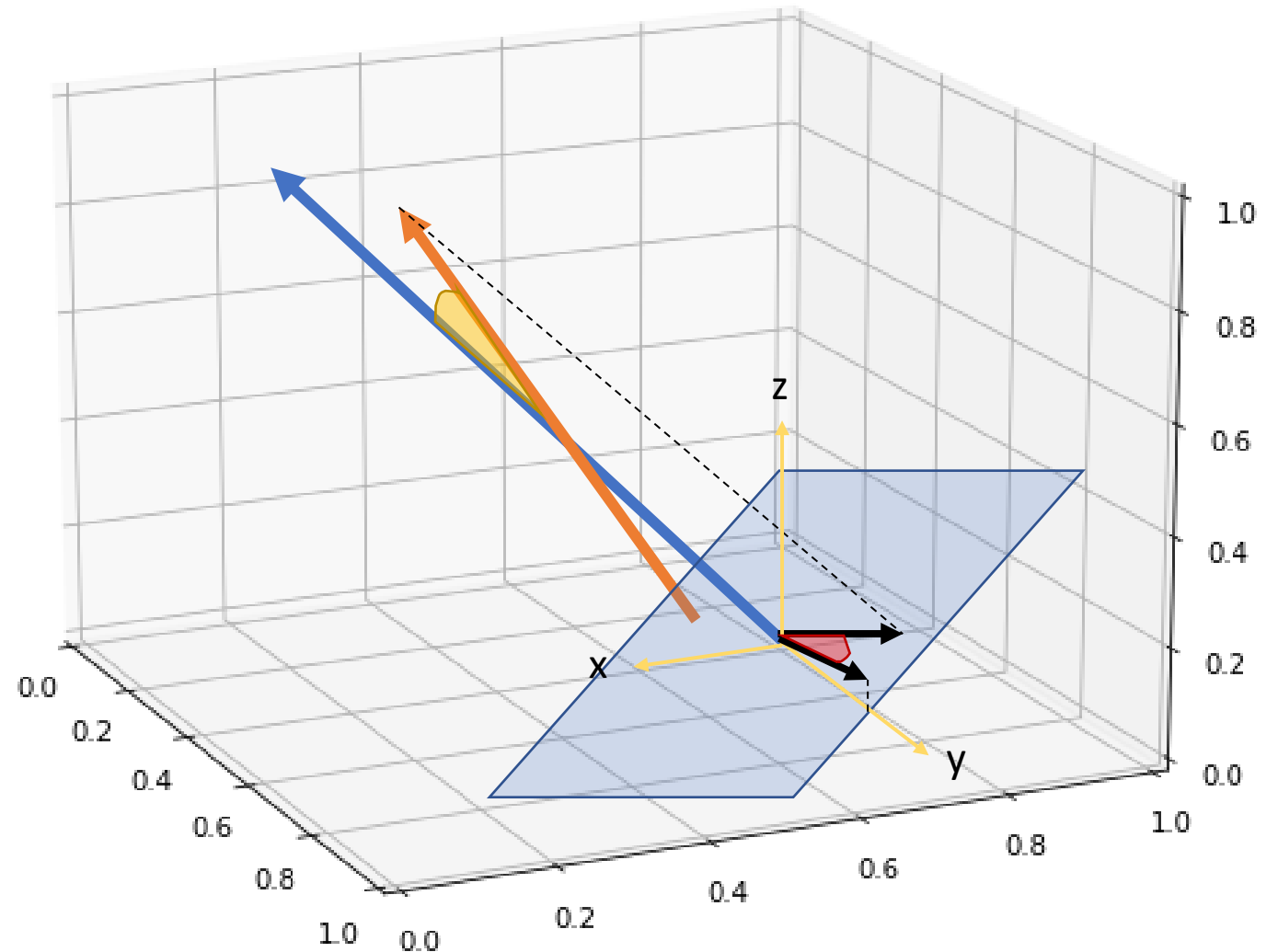
Rotation angle: Calculate angle between part projection and y-axis projection (ensuring consistent direction)



Axial runout - Process

Rotation angle: Calculate angle between part projection and y-axis projection (ensuring consistent direction)

Runout vector obtained!



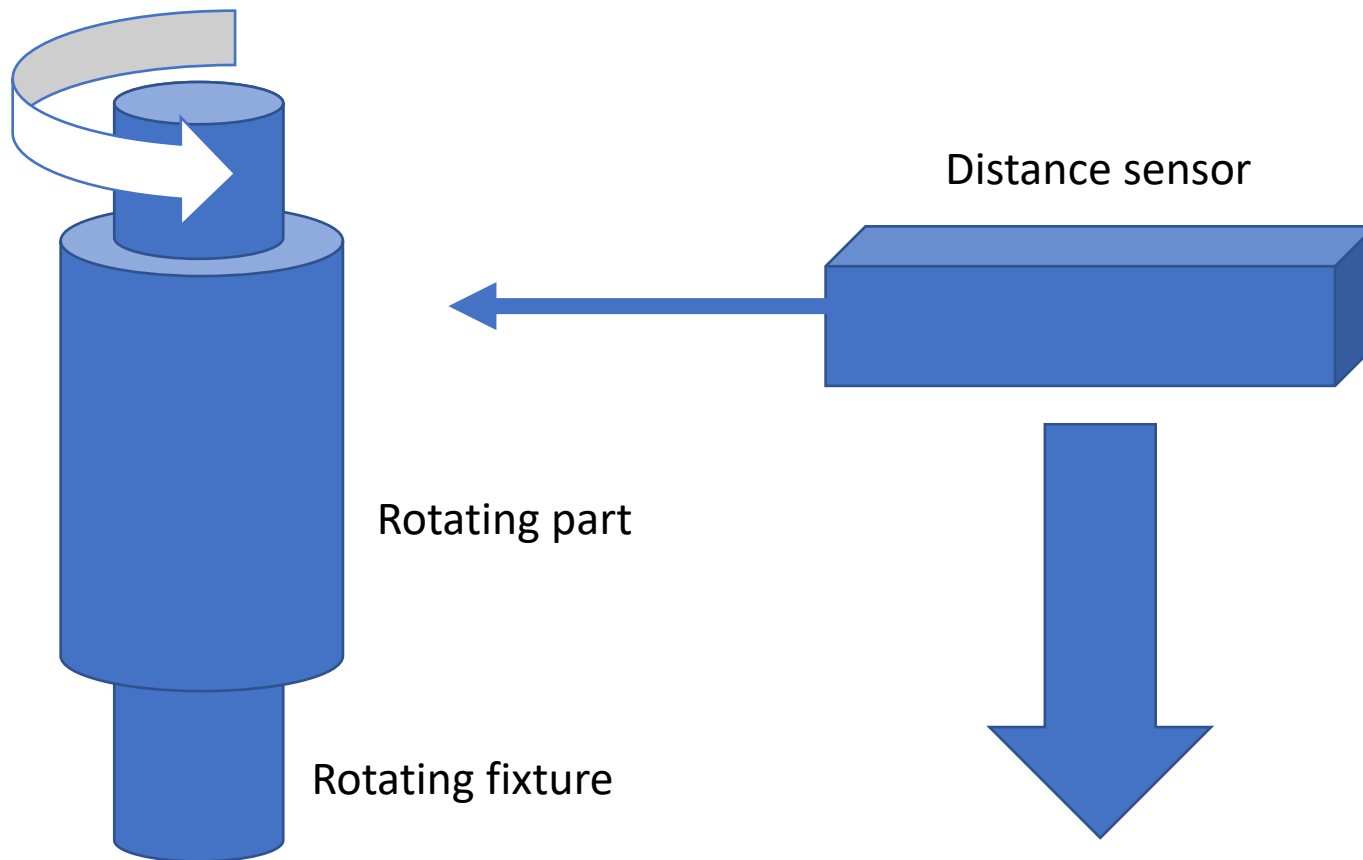
Model validation

- How well does this model extract the correct angles from measurement data?
- Under what scenarios does the model fall apart?
- Ideally – this would be performed by comparing the model predictions to known angle values taken via real measurement
 - But this is the problem we're currently solving – there are no known angles!

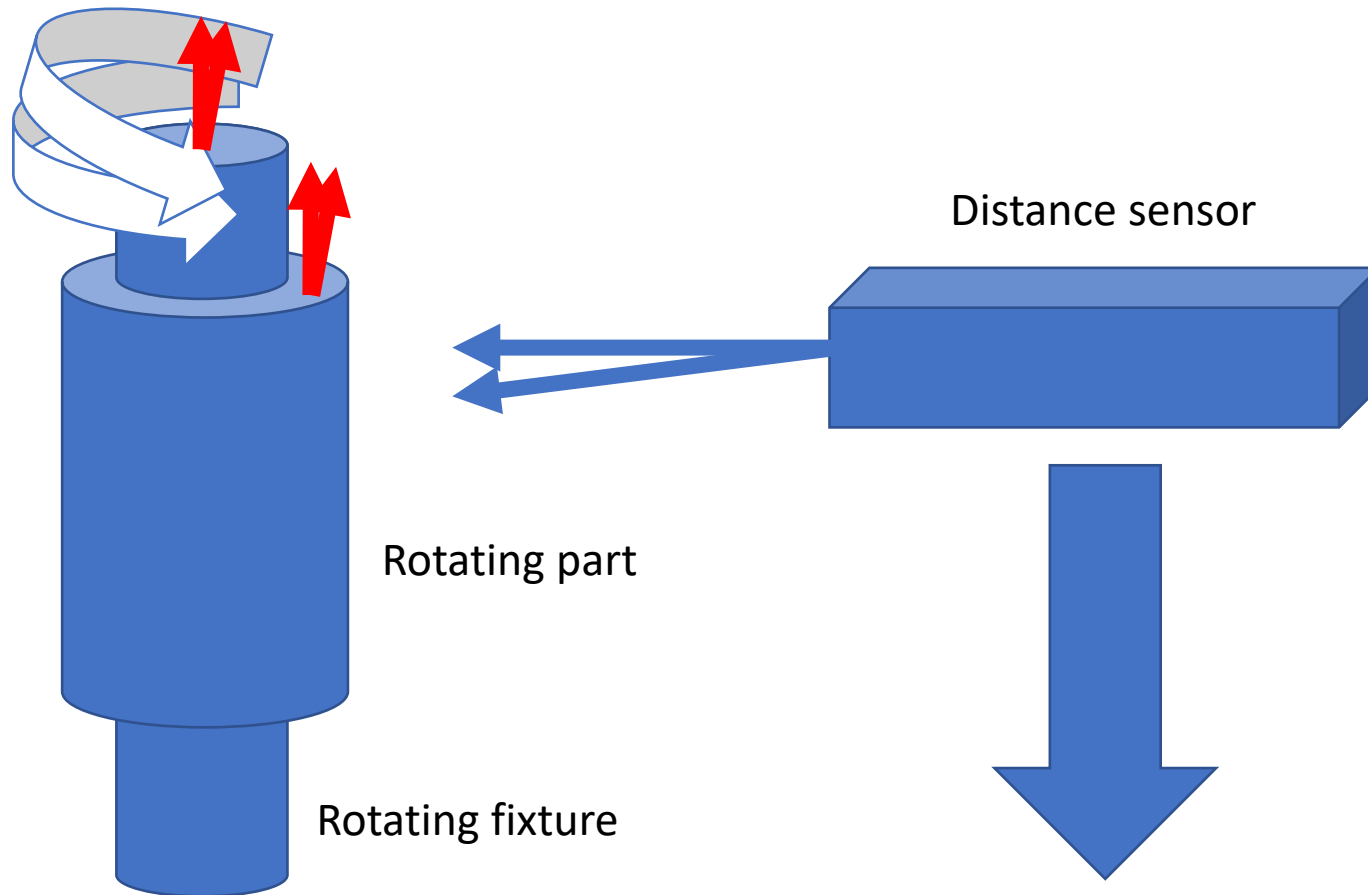
Model validation

- How well does this model extract the correct angles from measurement data?
- Under what scenarios does the model fall apart?
- Ideally – this would be performed by comparing the model predictions to known angle values taken via **real measurements**
 - But this is the problem we're currently solving – there are no known angles!
- Solution: Comparing the model predictions to known angles taken via **virtual measurements**

Physical measurement setup



Physical measurement setup



Virtual instrument

- Objects:
 - Distance sensor(position, orientation)
 - Fixture/inner part surface(radius, position, orientation)
 - Part outer surface(radius, position, orientation)
- Actions:
 - Capture distance from sensor to object
 - Rotate fixture about central axis
 - Rotate part about fixture axis
 - Repeat for 360 degrees
 - Move sensor to next position
 - Repeat for all required measurements

```
def FullMeasurement(sensor, part, fixture, measurementID):  
    mitaka.ZShift(0.025)  
    fixtureTop = MeasurementSystem(sensor, fixture)  
    fixtureTop.Capture(rotationAxis)  
    fixtureTop.SavePoints('fixtureSim_top_' + str(measurementID))  
  
    mitaka.ZShift(-0.003)  
    partTop = MeasurementSystem(sensor, part)  
    partTop.Capture(rotationAxis)  
    partTop.SavePoints('sim_top_' + str(measurementID))  
  
    mitaka.ZShift(-0.008)  
    partBottom = MeasurementSystem(sensor, part)  
    partBottom.Capture(rotationAxis)  
    partBottom.SavePoints('sim_bottom_' + str(measurementID))  
  
    mitaka.ZShift(-0.005)  
    fixtureBottom = MeasurementSystem(sensor, fixture)  
    fixtureBottom.Capture(rotationAxis)  
    fixtureBottom.SavePoints('fixtureSim_bottom_' + str(measurementID))  
    mitaka.ZShift(-0.009)
```

Virtual instrument

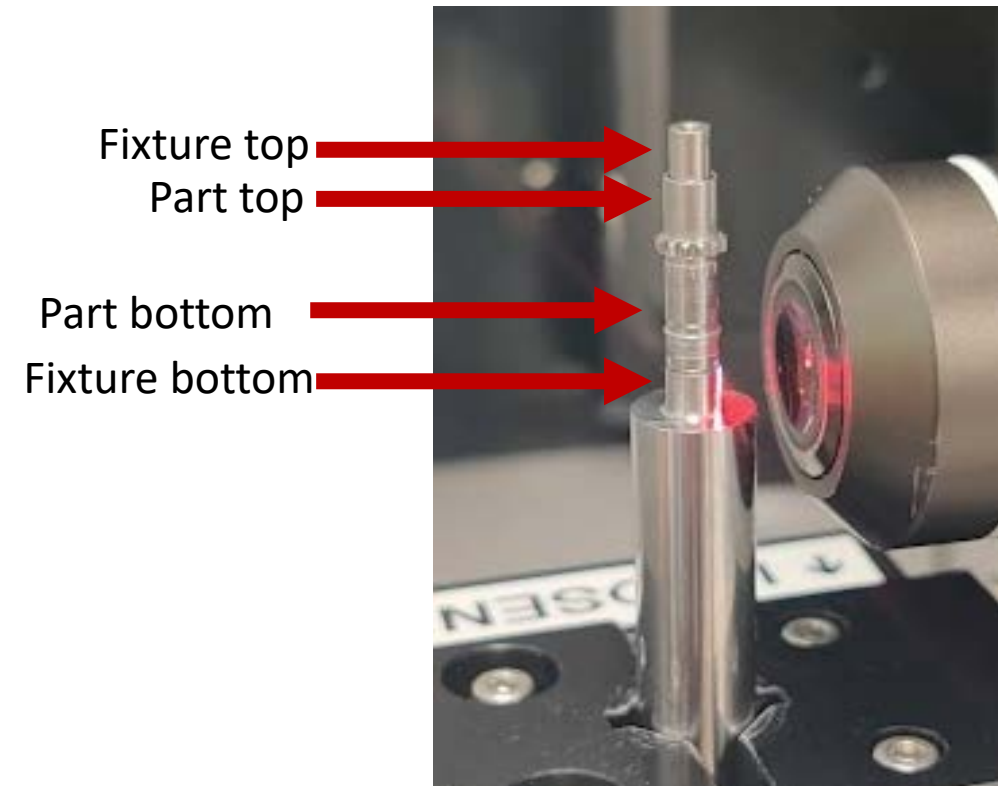
- Validation process:
 - Define a large set of simulated parts with varying levels of runout
 - Obtain measurement datasets using the virtual instrument
 - Feed the measurement data into the model
 - **Compare model predictions to known angles**

```
def FullMeasurement(sensor, part, fixture, measurementID):  
    mitaka.ZShift(0.025)  
    fixtureTop = MeasurementSystem(sensor, fixture)  
    fixtureTop.Capture(rotationAxis)  
    fixtureTop.SavePoints('fixtureSim_top_' + str(measurementID))  
  
    mitaka.ZShift(-0.003)  
    partTop = MeasurementSystem(sensor, part)  
    partTop.Capture(rotationAxis)  
    partTop.SavePoints('sim_top_' + str(measurementID))  
  
    mitaka.ZShift(-0.008)  
    partBottom = MeasurementSystem(sensor, part)  
    partBottom.Capture(rotationAxis)  
    partBottom.SavePoints('sim_bottom_' + str(measurementID))  
  
    mitaka.ZShift(-0.005)  
    fixtureBottom = MeasurementSystem(sensor, fixture)  
    fixtureBottom.Capture(rotationAxis)  
    fixtureBottom.SavePoints('fixtureSim_bottom_' + str(measurementID))  
    mitaka.ZShift(-0.009)
```

Model validation

- A total of 900 permutations were created with a variety of relative and rotation angles
- Each measured four times as per measurement protocol
- Input into model...

```
theta = [0.1, 0.2, 0.5, 1.0, 2.0, 5.0]  
phi = [0, 0.1, 1, 10, 100]  
r = 1.6
```



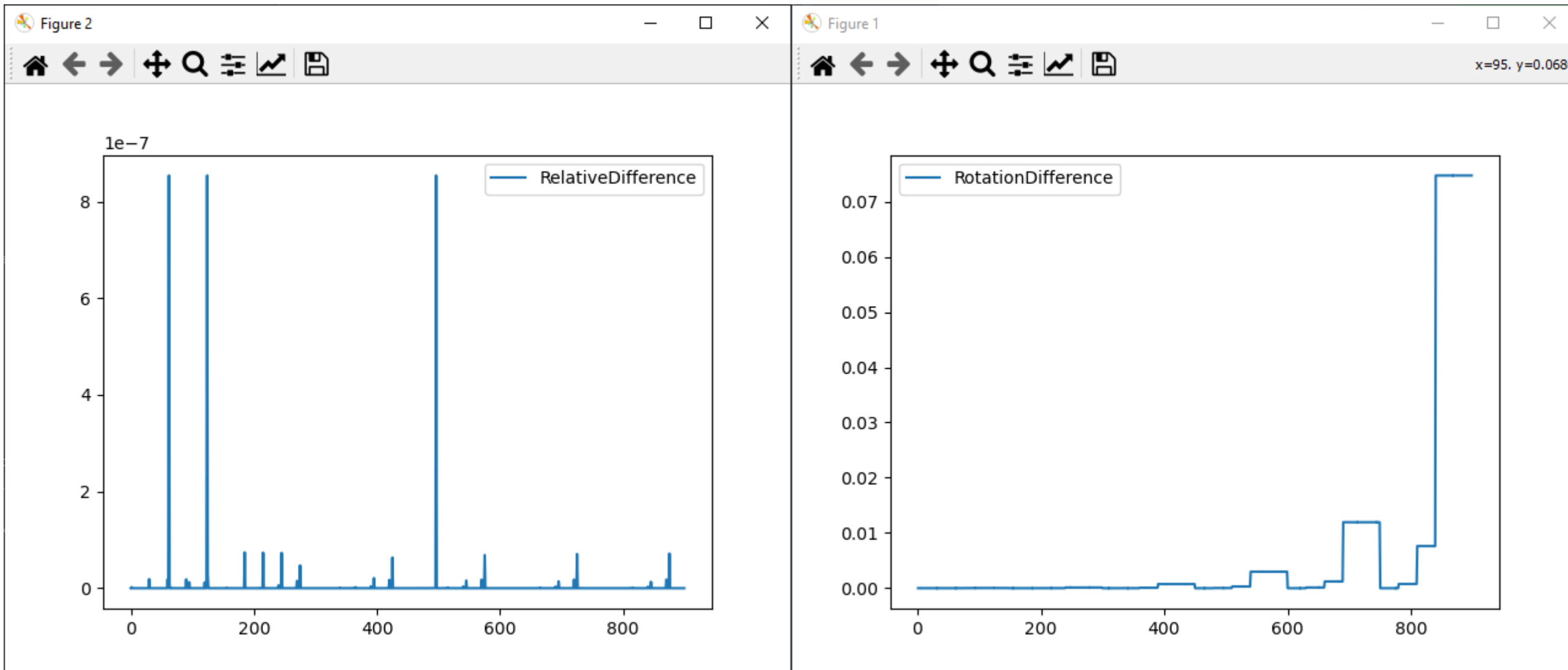
Model validation

A total of 900 permutations were

```
theta = [0.1, 0.2, 0.5, 1.0, 2.0, 5.0]
```

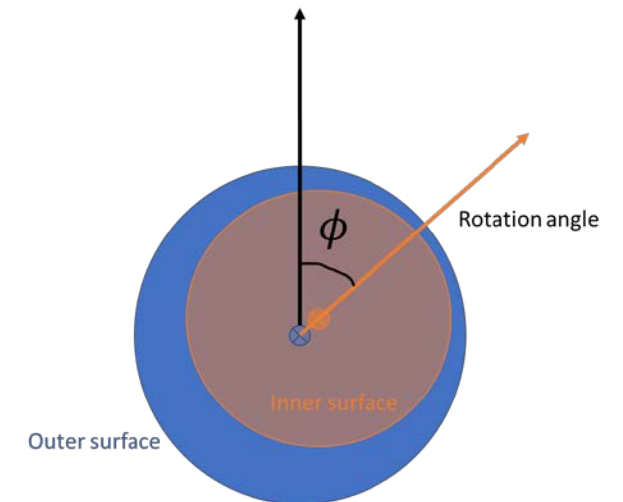
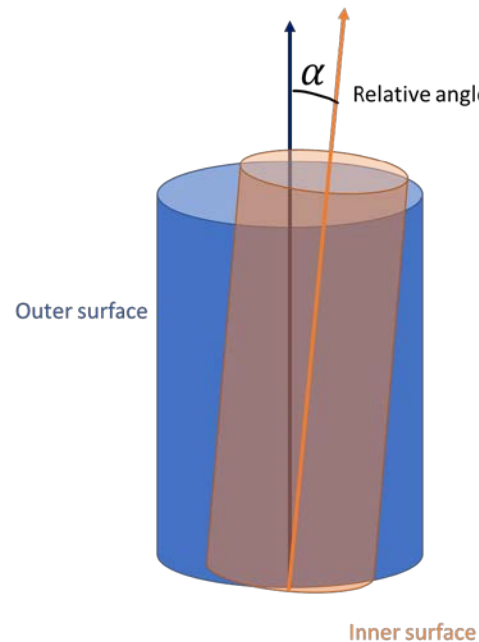
df - DataFrame												
Index	Rotation Angle (Simulation)	Relative Angle (Simulation)	Rotation Angle (Measurement)	Relative Angle (Measurement)	FixtureXRotation	PartXRotation	FixtureZRotation	PartZRotation	RotationDifference	RelativeDifference	Fixture	Part
1	90.05	0.000174534	90.05	0.000174531	0.1	0.1	0	0.1	5.16063e-06	2.08823e-09	0.1	1.1
2	90.5	0.00174531	90.5	0.00174531	0.1	0.1	0	1	2.78807e-07	2.08826e-10	0.1	10.1
3	95	0.0174311	95	0.0174311	0.1	0.1	0	10	3.41815e-08	2.09088e-11	0.1	100...
4	140	0.153209	140	0.153209	0.1	0.1	0	100	3.93067e-09	4.75778e-12	0.1	100...
5	0	0.1	4.20159e-05	0.1	0.1	0.2	0	0	4.20159e-05	3.64467e-12	0.1	0.2
6	0.199999	0.1	0.199999	0.1	0.1	0.2	0	0.1	1.33656e-10	7.28928e-12	0.1	1.2
7	1.99969	0.10003	1.99969	0.10003	0.1	0.2	0	1	1.83187e-09	7.28707e-12	0.1	10.2
8	19.7065	0.102994	19.7065	0.102994	0.1	0.2	0	10	2.58318e-11	7.07742e-12	0.1	100...
9	124.374	0.238633	124.374	0.238633	0.1	0.2	0	100	7.28804e-10	1.52733e-12	0.1	100...
10	0	0.4	4.12456e-10	0.4	0.1	0.5	0	0	4.12456e-10	1.82232e-12	0.1	0.5
11	0.124999	0.4	0.124999	0.4	0.1	0.5	0	0.1	4.86245e-10	9.11171e-12	0.1	1.5
12	1.24997	0.400019	1.24997	0.400019	0.1	0.5	0	1	3.91987e-10	3.64453e-12	0.1	10.5
13	12.4763	0.401895	12.4763	0.401895	0.1	0.5	0	10	3.61686e-10	7.25503e-12	0.1	100...
14	110.777	0.526654	110.777	0.526654	0.1	0.5	0	100	3.75934e-10	3.46023e-12	0.1	100...
16	0.11111	0.9	0.11111	0.9	0.1	1	0	0.1	3.81539e-11	1.37692e-11	0.1	1.1
17	1.11109	0.900017	1.11109	0.900017	0.1	1	0	1	2.28024e-11	6.88449e-12	0.1	10.1

Model validation

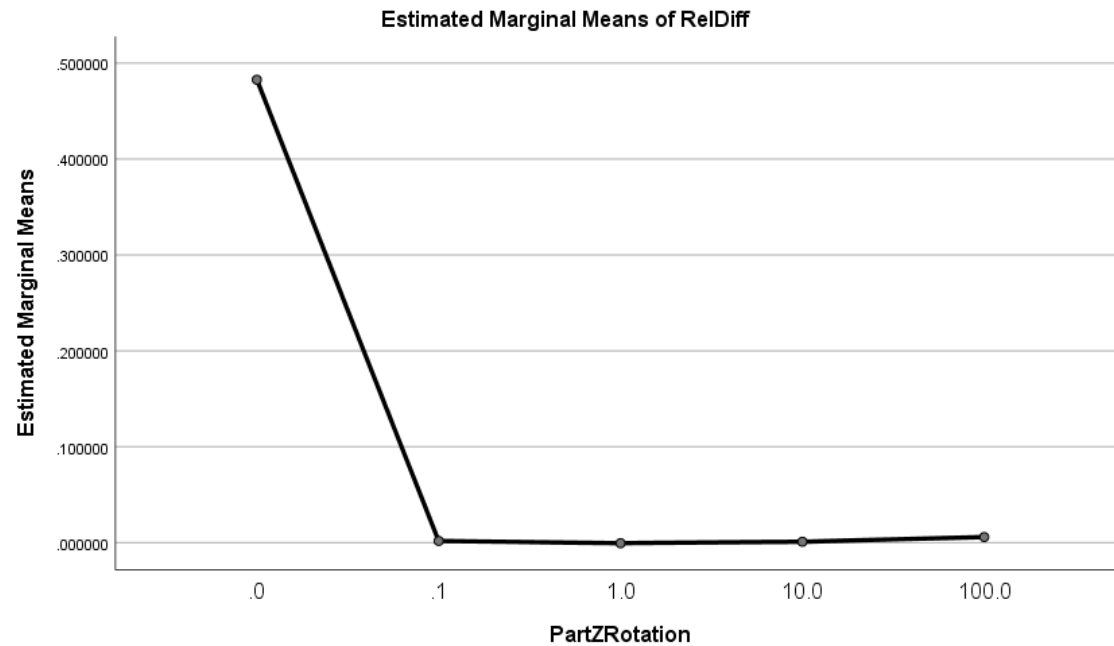


Model validation

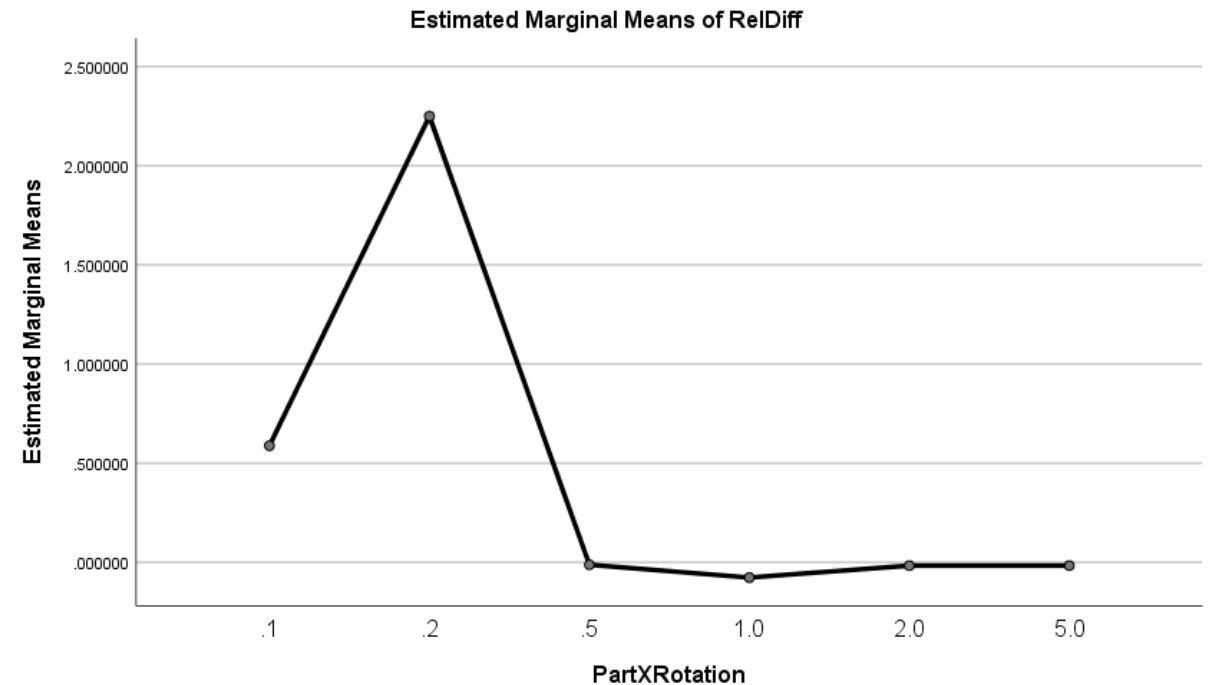
- Analysis of Variance (ANOVA) was used to identify significant factors contributing to model error for the two critical metrics:
- Relative angle (runout magnitude)
- Rotation angle (runout direction)



Relative Angle

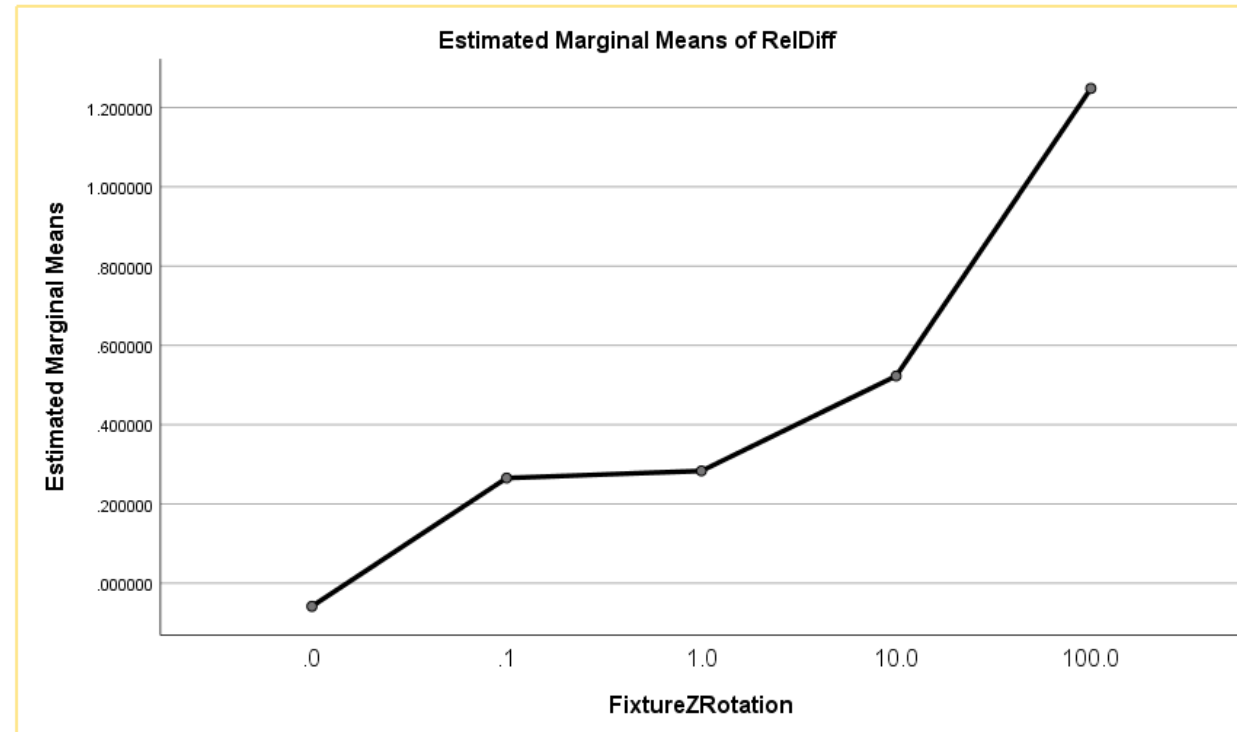


Errors occur when the part is well aligned to the rotation axis...



Relative angle

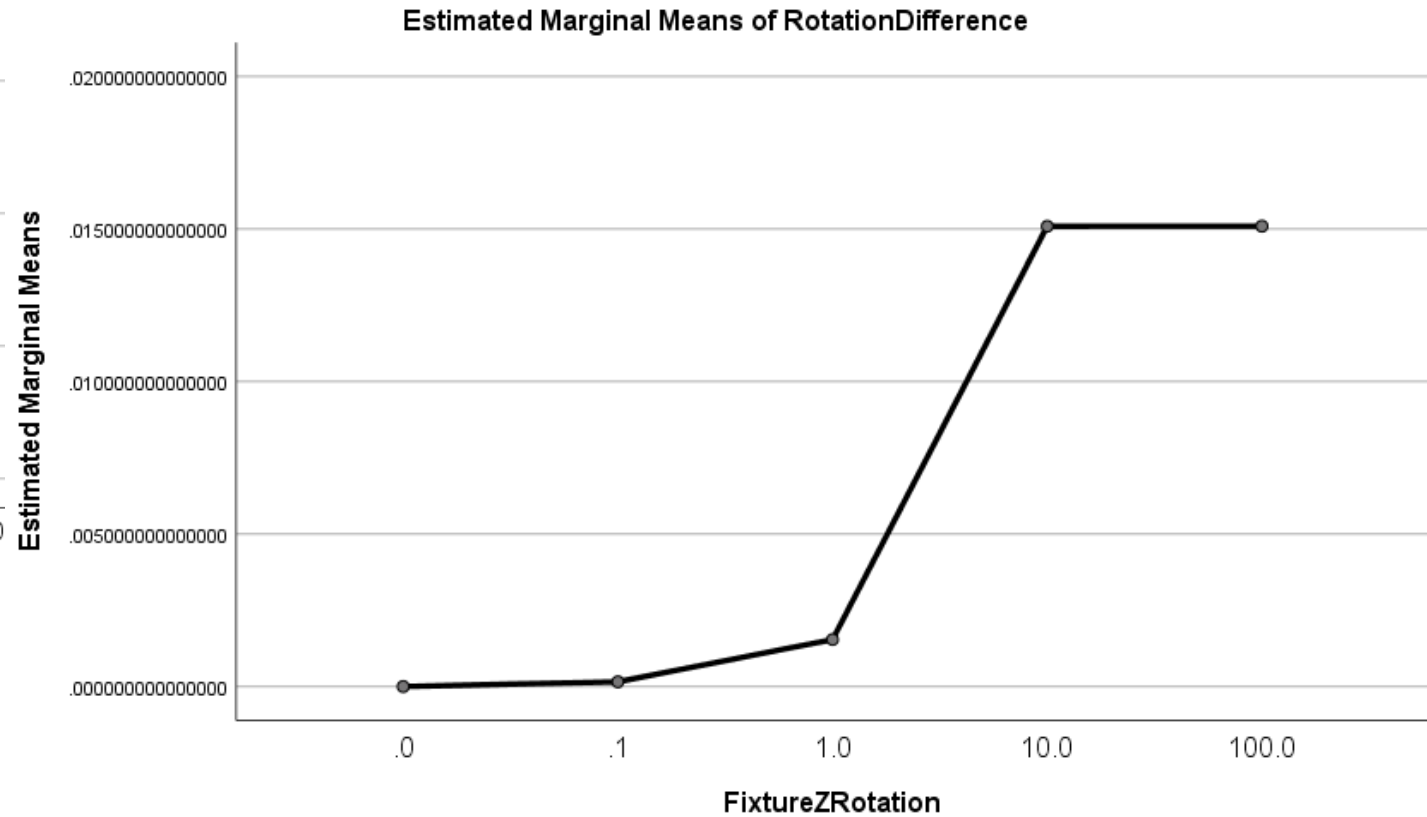
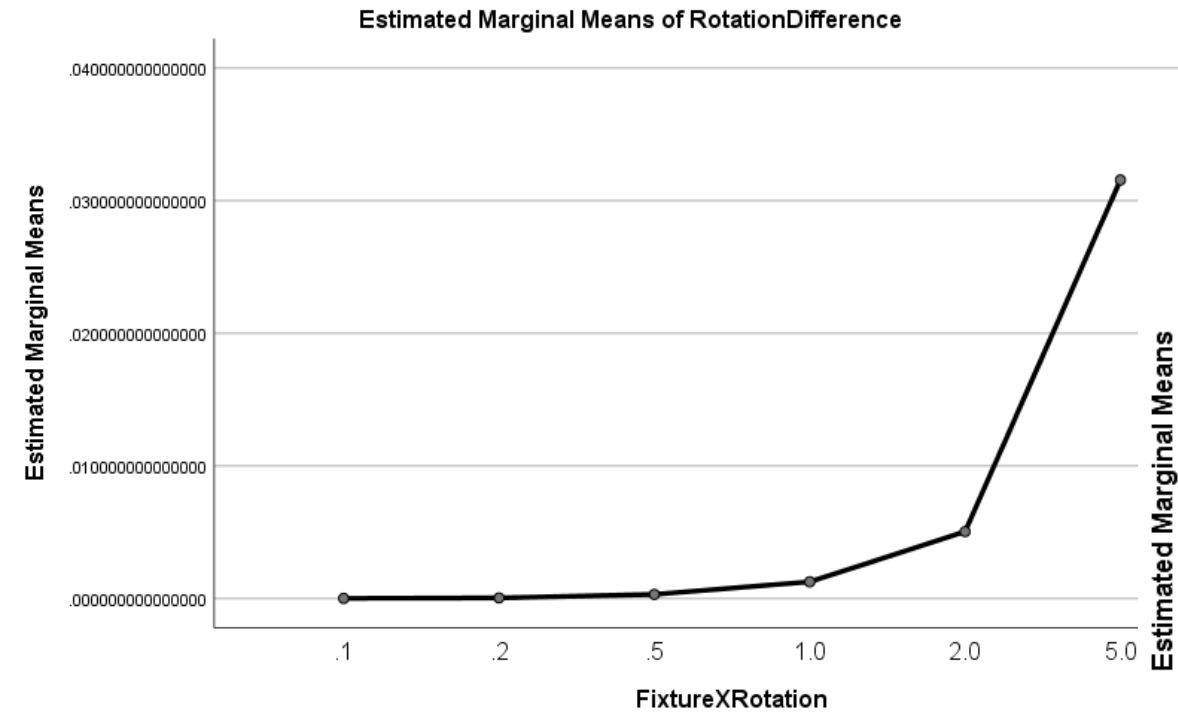
... and the fixture is poorly aligned!



But, magnitudes are tiny ($< 1e-6^\circ$), so not overwhelming measurement accuracy requirements

Rotation angle

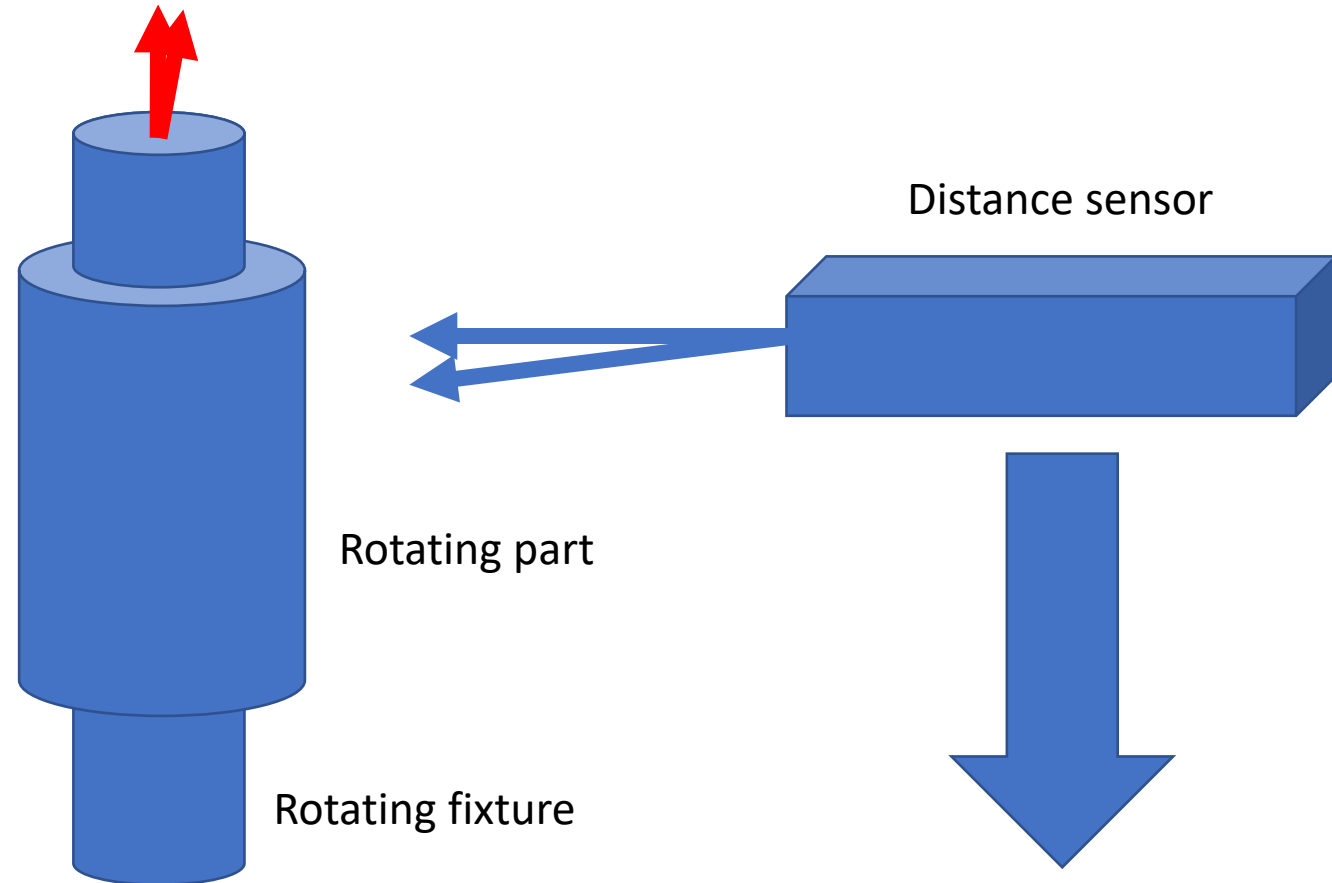
Large misalignment of the fixture leads to large rotation angle errors (0.07°)



Model validation - conclusion

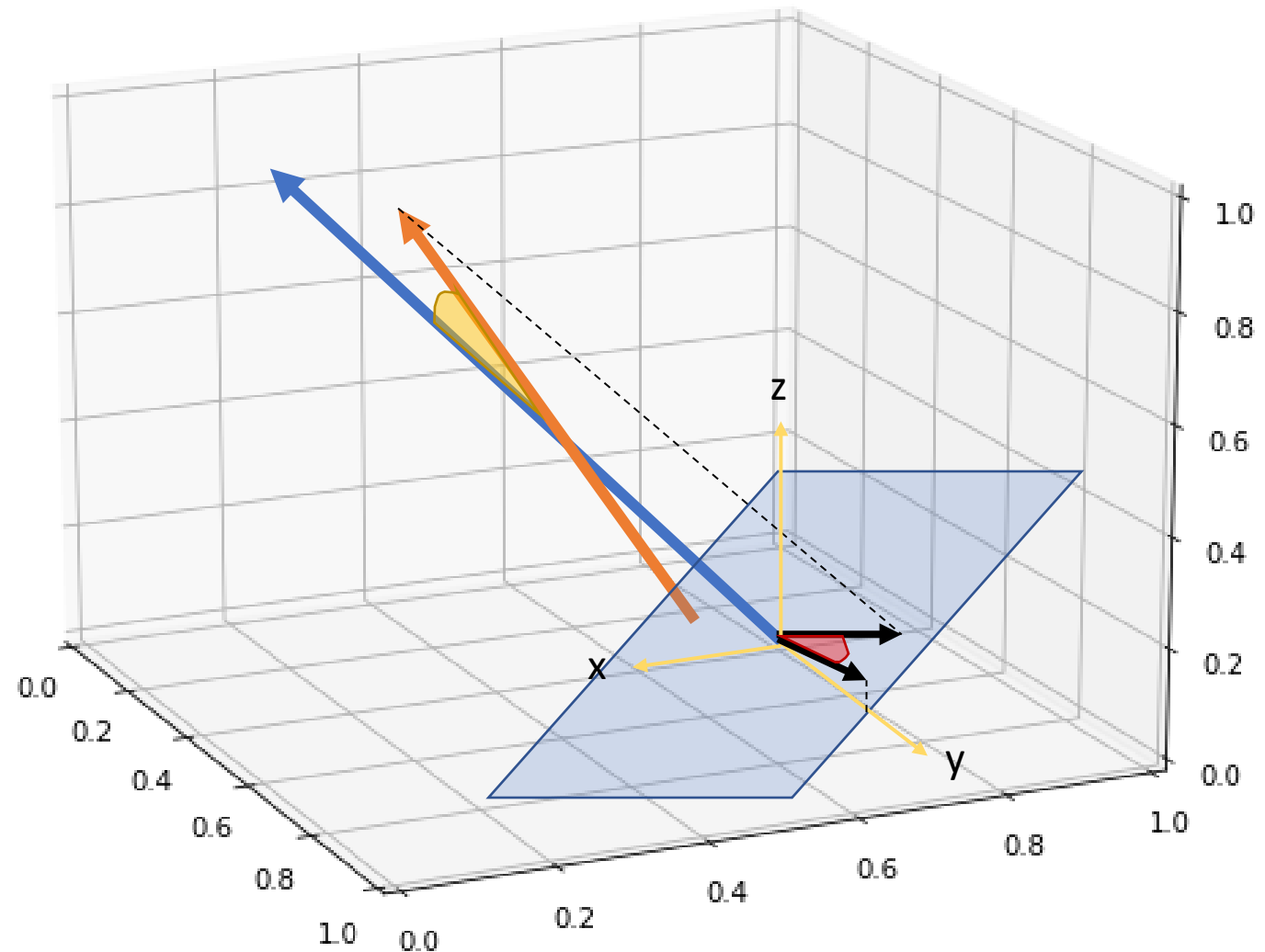
- For the model to be accurate for purpose:
 - the fixture/sensor pair should be perpendicularly aligned (90°),
 - And the fixture should be aligned with the rotation axis,

With total angular errors **smaller than 5°**



Conclusion

- Runout Vector (Magnitude and direction) characterisation has been developed
- Validation with virtual instrument





Thank you



DAT4.Zero has received funding from the European Union's Horizon 2020 Research and Innovation programme under Grant Agreement No. **958363**

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