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# **IFoA GIRO Conference 2024**

18–20 November, ICC, Birmingham



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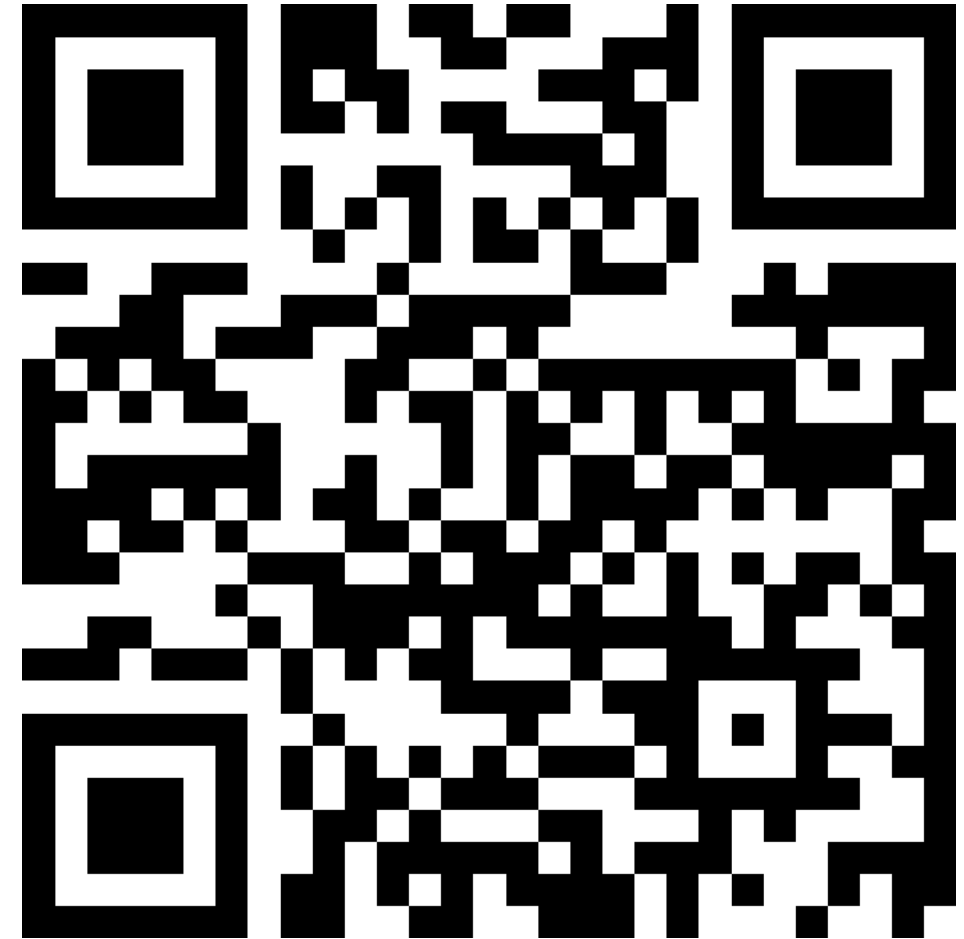
# Boosting sustainability agenda with data science

Dr. Alexey Mashechkin, Debashish Dey FIA  
Assoc. Prof. George Tzougas, Shubham Mehta

**IFoA GIRO Conference 2024**

# Agenda

- **Speakers**
- **Background**
- **Revamping classics**
- **New dimensions and products**
- **Challenges**
- **IFoA Data Science**





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# Speakers



# Speaker introductions



**Dr. Alexey Mashechkin**



**Debashish Dey FIA**



**Assoc. Prof. George Tzougas**



**Shubham Mehta**

	Dr. Alexey Mashechkin	Debashish Dey FIA	Assoc. Prof. George Tzougas	Shubham Mehta
<b>Volunteering with IFoA</b>	✓	✓	✓	✓
<b>- Role</b>	Lifelong Learning Pillar Chair	Sustainability & Climate Change Working Party Co-chair (practitioners)	Sustainability & Climate Change Working Party Co-chair (academics)	Sustainability & Climate Change Working Party Member
<b>Background</b>				
<b>Business</b>	✓	✓		✓
<b>- Actuarial traditional</b>	✓	✓		✓
<b>- Actuarial non-traditional</b>				✓
<b>- Data science</b>	✓			
<b>Academic focus</b>			✓	



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# Background



# Sustainability agenda

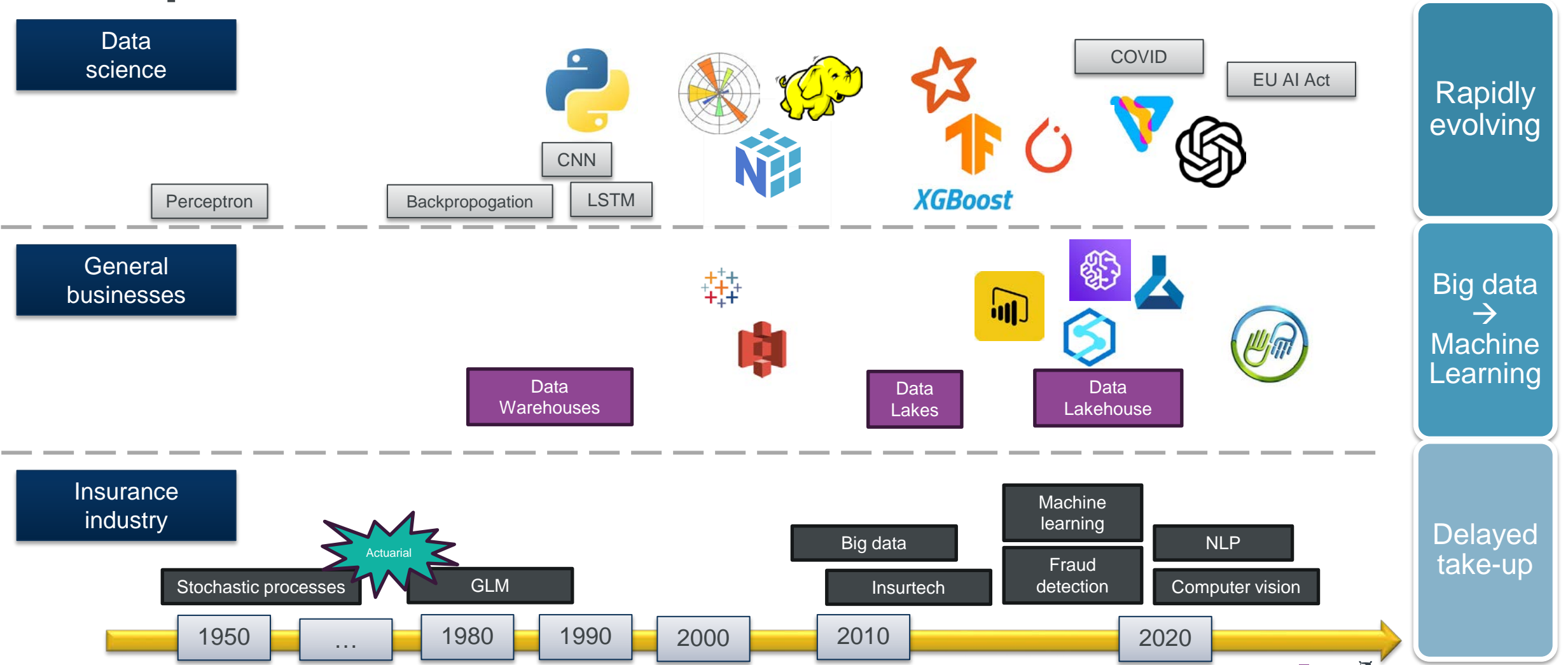


GOVERNMENTS



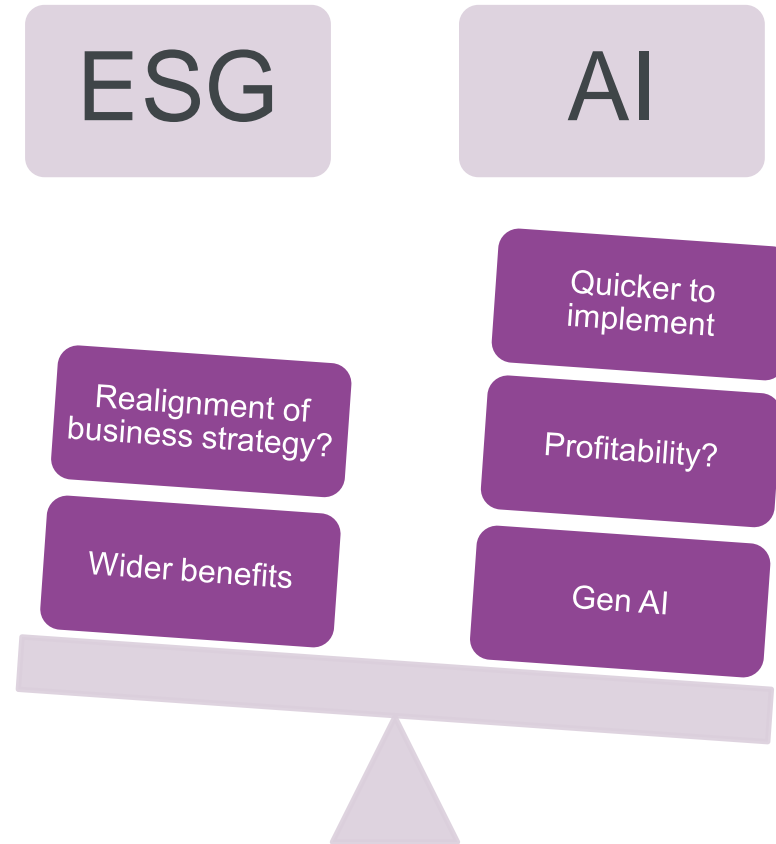
BUSINESSES

# Adoption of data science across industries





# Conflicting goals for businesses?





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# Revamping classics



# Use cases within general insurance industry

Business area	Risk/Opportunities	Solutions/Techniques	Data science-ready?
<b>Pricing</b>	<ul style="list-style-type: none"> <li>• <b>Real-time</b> claims &amp; quotes</li> <li>• <b>Usage-based</b> and more precise tariffication</li> <li>• <b>NatCat</b> risks</li> <li>• <b>ESG-metrics</b></li> <li>• Data <b>bias risk</b></li> </ul>	<ul style="list-style-type: none"> <li>• Utilisation of <b>satellite images</b></li> <li>• Climate-models</li> <li>• Integrated IT solutions</li> <li>• Big and/or <b>external</b> data</li> </ul>	✓
<b>Reserving</b>	<ul style="list-style-type: none"> <li>• Revamped <b>portfolio</b> segmentation</li> <li>• <b>NatCat</b> risks</li> </ul>	<ul style="list-style-type: none"> <li>• <b>Individual</b> reserving</li> <li>• <b>Big</b> data</li> </ul>	✓
<b>Capital modelling</b>	<ul style="list-style-type: none"> <li>• More <b>complex &amp; accurate</b> modelling required</li> </ul>	<ul style="list-style-type: none"> <li>• Enhanced predictive models</li> </ul>	✓
<b>Governance, Infrastructure</b>	<ul style="list-style-type: none"> <li>• ORSA / better internal <b>compliance</b></li> <li>• Streamlined <b>IT services</b> and connected internal <b>databases</b></li> </ul>	<ul style="list-style-type: none"> <li>• Real-time statistics</li> <li>• Cloud solutions</li> <li>• <b>Enhanced models</b></li> <li>• <b>Management statistics</b></li> </ul>	✓
<b>Reporting</b>	<ul style="list-style-type: none"> <li>• <b>ESG</b> reporting</li> <li>• <b>Efficient</b> reporting e.g. Solvency 2 templates</li> </ul>	<ul style="list-style-type: none"> <li>• Data scraping techniques</li> <li>• <b>Large language models</b></li> <li>• <b>Real-time</b> management information</li> </ul>	✓

# Use case: Climate-related hazards on claim frequency prediction in motor insurance

Existing models are not directly applicable to incomplete records of weather hazards and claims.

## Key contribution to addressing this gap

- ✓ **Novel** class of predictive compound frequency models.
- ✓ Study looks at **Greek motor insurance data set**, documenting climate events and storm-triggered motor claims.
- ✓ The proposed modelling framework uncovers the combined distribution of storms and claims **despite incomplete data**.
- ✓ **Geospatial covariates**

## Key finding

- ✓ **Negative intrinsic dependence** between actual storms and claim frequencies per storm.

## Practical impact

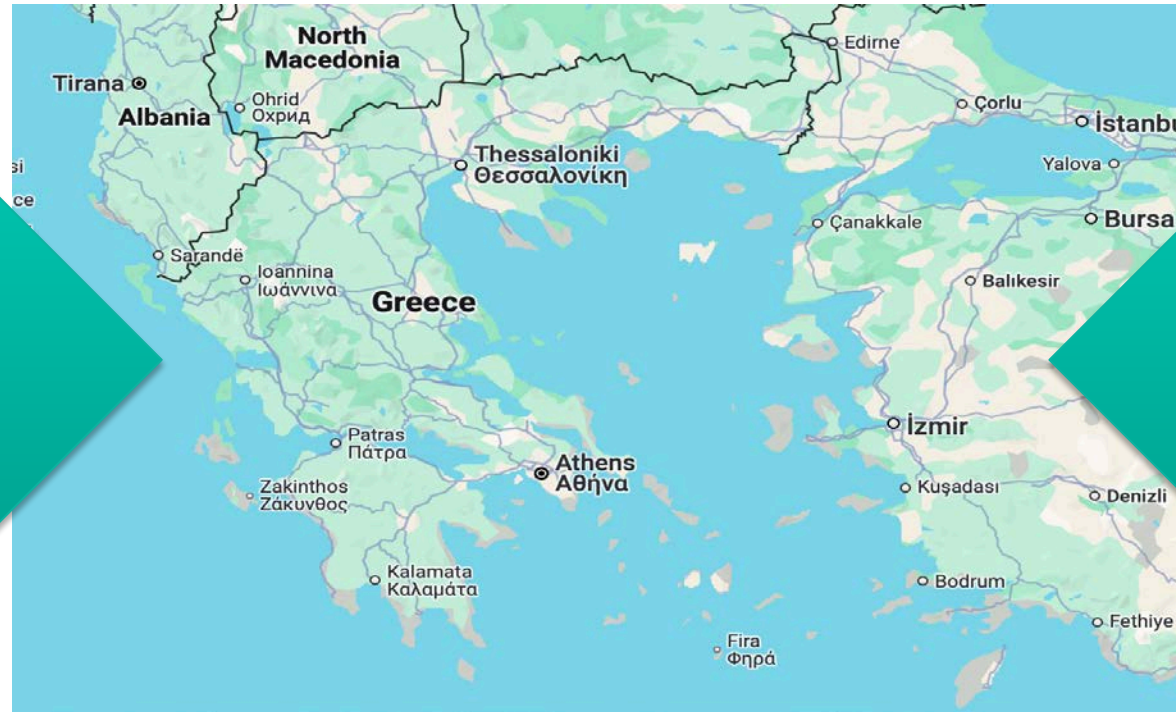
- ✓ **Offers valuable insights** for the insurance sector, especially with Greece's new law mandating natural disaster vehicle insurance from 1 January 2025.

# Use case: Climate-related hazards on claim frequency prediction in motor insurance

The study analyses a **Greek motor insurance dataset**, documenting storm events and storm-triggered claim counts across various regions of Greece over time.

## Geospatial features

- ✓ Vulnerability
- ✓ Coast ratio
- ✓ Log alt mean
- ✓ Residential
- ✓ Industrial
- ✓ Agricultural
- ✓ Wetlands



## Exposure

- ✓ Motorcycle
- ✓ Cars
- ✓ Trucks

# Use case: Climate-related hazards on claim frequency prediction in motor insurance

Goal: Derive the joint distribution of  $(M, N)$ , based on geospatial features  $x$  in a region over a year, using complete data  $(\tilde{M}, \tilde{N})$ .  $\tilde{M}$  is actual number of storms  $\tilde{N} = (\tilde{N}_1, \tilde{N}_2, \dots, \tilde{N}_{\tilde{M}})$  is the number of claims triggered per storm.

## GENERAL MODEL

$$f_{\tilde{M}, \tilde{N}}(\tilde{m}, \tilde{n}; x) = f_{\tilde{M}}(\tilde{m}; x) \prod_{t=1}^{\tilde{m}} f_{\tilde{N}_t | \tilde{M}}(\tilde{n}_t; \tilde{m}, x)$$

## CASE 1: INDEPENDENT

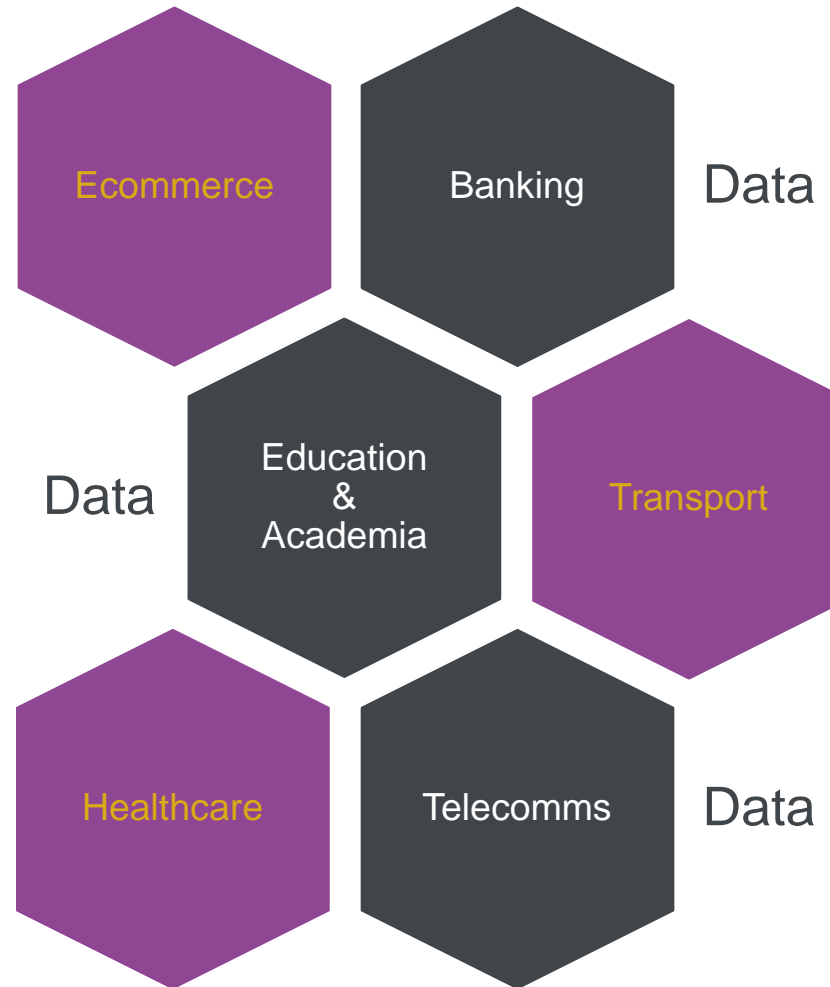
$$f_{M,N}(m, n; x) = \sum_{\tilde{m} \geq m} \binom{\tilde{m}}{m} f_{\tilde{M}}(\tilde{m}, x) \left( f_{\tilde{N}_t}(0; x) \right)^{\tilde{m}-m} \left( 1 - f_{\tilde{N}_t}(0; x) \right)^m \times \sum_{\tilde{n} \in \mathcal{C}_n} \left( \frac{f_{\tilde{N}_t}(\tilde{n}_t; x)}{1 - f_{\tilde{N}_t}(0; x)} \right)^m$$

## CASE 2: DEPENDENT

$$f_{mix}(m, n; x, \Phi_{mix,k}) = \sum_{b=1}^B \pi_b(x) f_{M,N}(m, n; x)$$

**Outcome:** Fitted compound Poisson-Poisson (PP), Negative Binomial-Negative Binomial (NN) and Negative Binomial-Poisson (NP) models. Mixture models with components were also fitted leading to 42 total models. Cross-validation used to assess predictive performance and computational burden - best-performing model being 5NN.

# Take-up in other industries





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# New dimensions and products





# Global view - positive impact of data science/AI on SDGs

## General use cases around the world



Geospatial



Smart cities



Education



Healthcare



Banking

## Predicted quantifiable impacts

97m new jobs in  
26 countries

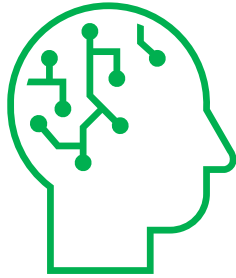
Africa +\$400m  
GDP by 2030

+40% healthcare  
delivery

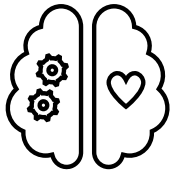
Global +\$15.7tr  
GDP by 2030

Sources: [The Economist](#), [Medium](#), [ORF](#), [PWC](#), [WEF](#)

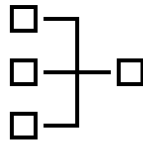
# Sustainable data science solutions



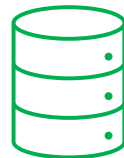
Green AI (vs Red AI)



20W



Spiking Neural  
Networks



Green  
data centres

Sustainable  
Solutions



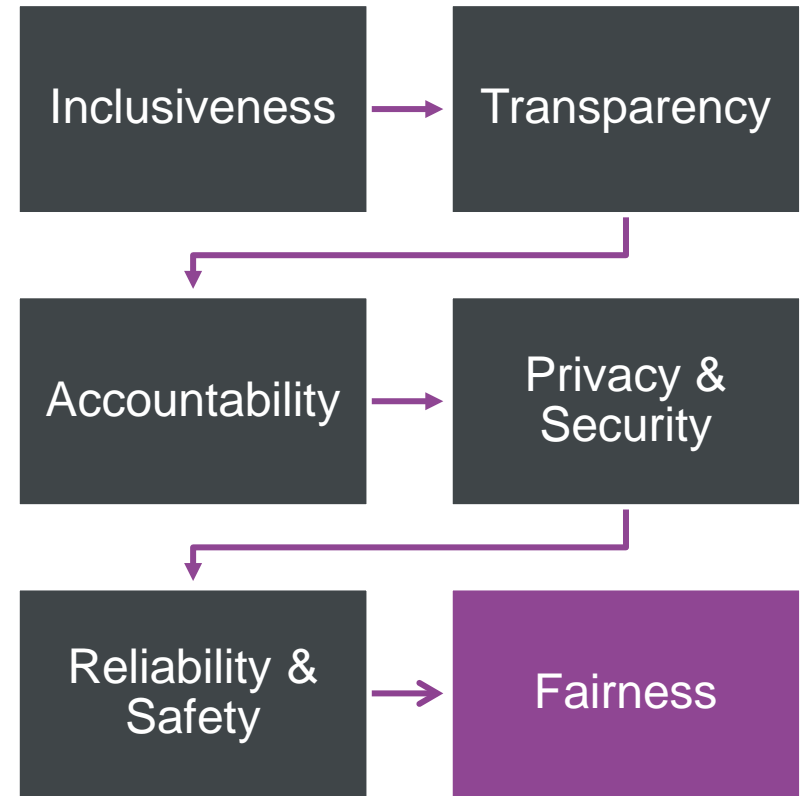
Green cloud  
/ Cloud sustainability solutions



# Responsible data science solutions

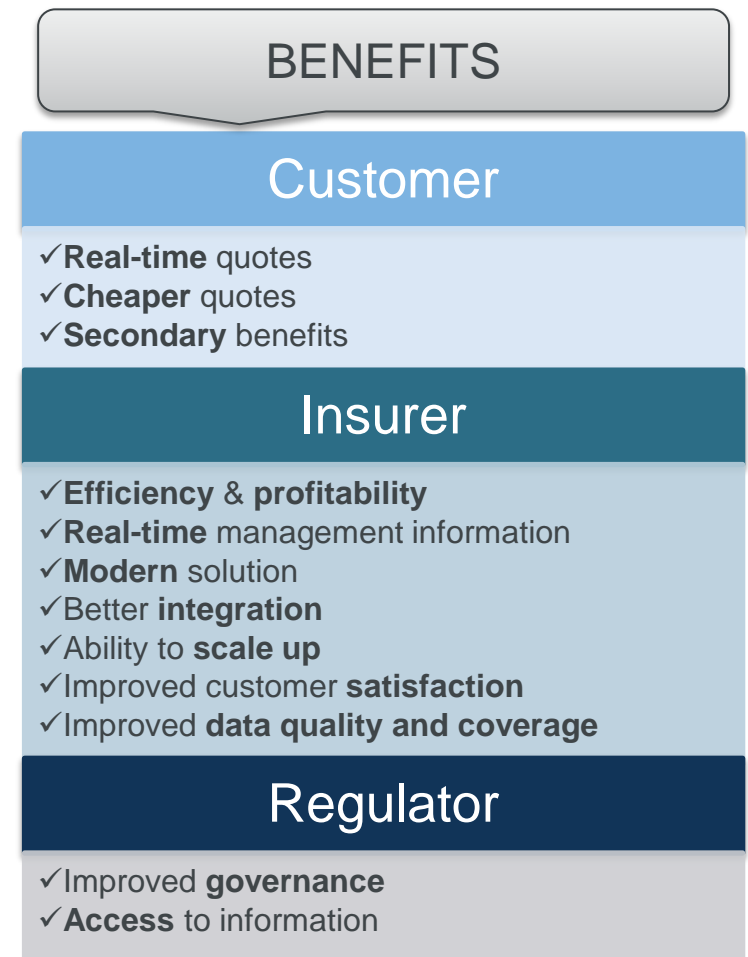
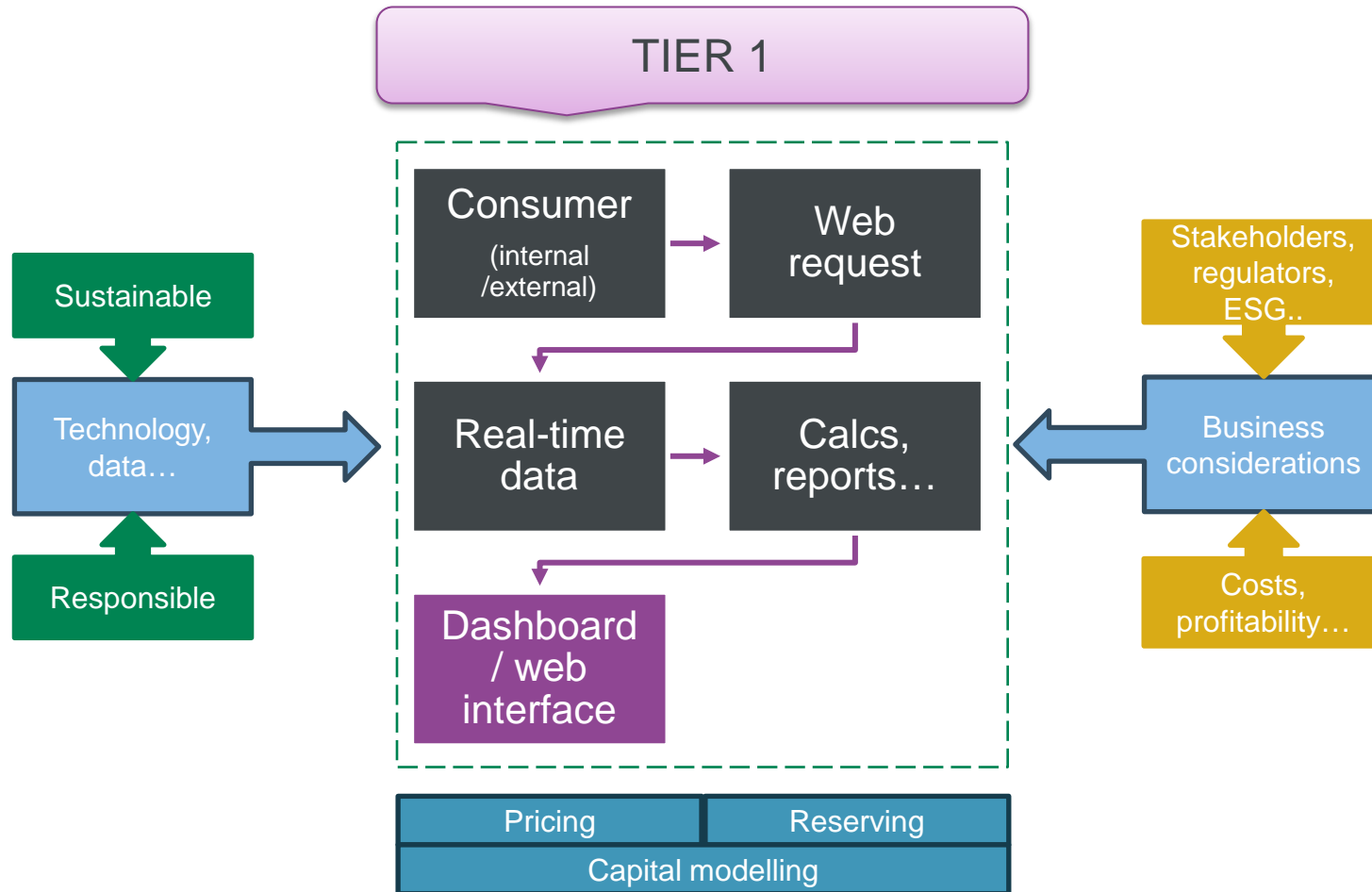


Responsible solutions



Source: [Microsoft](#)

# Insurance as a service (IAAS)



# Insurance as a service (IAAS)

## TIER 2

### Insurer-driven solutions

- ✓ Customer-centric
  - ✓ Health trackers
- ✓ ESG-metrics
- ✓ Consulting as a part of insurance products
- ✓ Protection gap reduction

### Market-driven solutions

- ✓ Direct/Indirect sustainability-aligned
  - ✓ Coral reef insurance
  - ✓ Solar panel minimum annual power output insurance
- ✓ Increasing ESG regs
- ✓ Competitor pressures
- ✓ Customer and employee needs
  - ✓ Generational challenges

## BENEFITS

### Individual

- ✓ Better and healthier **day-to-day**
- ✓ **Inclusive** work environment

### Society

- ✓ **Healthier behaviours** for both individuals and corporates
- ✓ **Sustainable** society
- ✓ Promotion of **ESG-aligned** practices and projects



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# Challenges



# Challenges of ESG integration & looking ahead



**Alignment**  
with  
business strategy?



**Increased regulations**



Rapidly  
evolving space  
– **risk & opportunity**



Increased **complexity** and  
lack of **understanding** of  
models

## ADOPTION

**75%**

More **underwriters**  
considering ESG  
factors ([Marsh](#))

**25%**

**Group-wide approach**  
for ESG integration for  
underwriters ([Marsh](#))

## GOVERNANCE

**58%**

**Chief sustainability officer** retains  
responsibility ([Deloitte](#))

**39%**

Different teams  
working as  
**internal silos** ([KPMG](#))

## CHALLENGES

**45%**

**Insufficient resources**  
and capacity ([KPMG](#))

**45%**

Pressure from  
**ESG rating agencies** ([Deloitte](#))

# Insurance and actuarial perspective





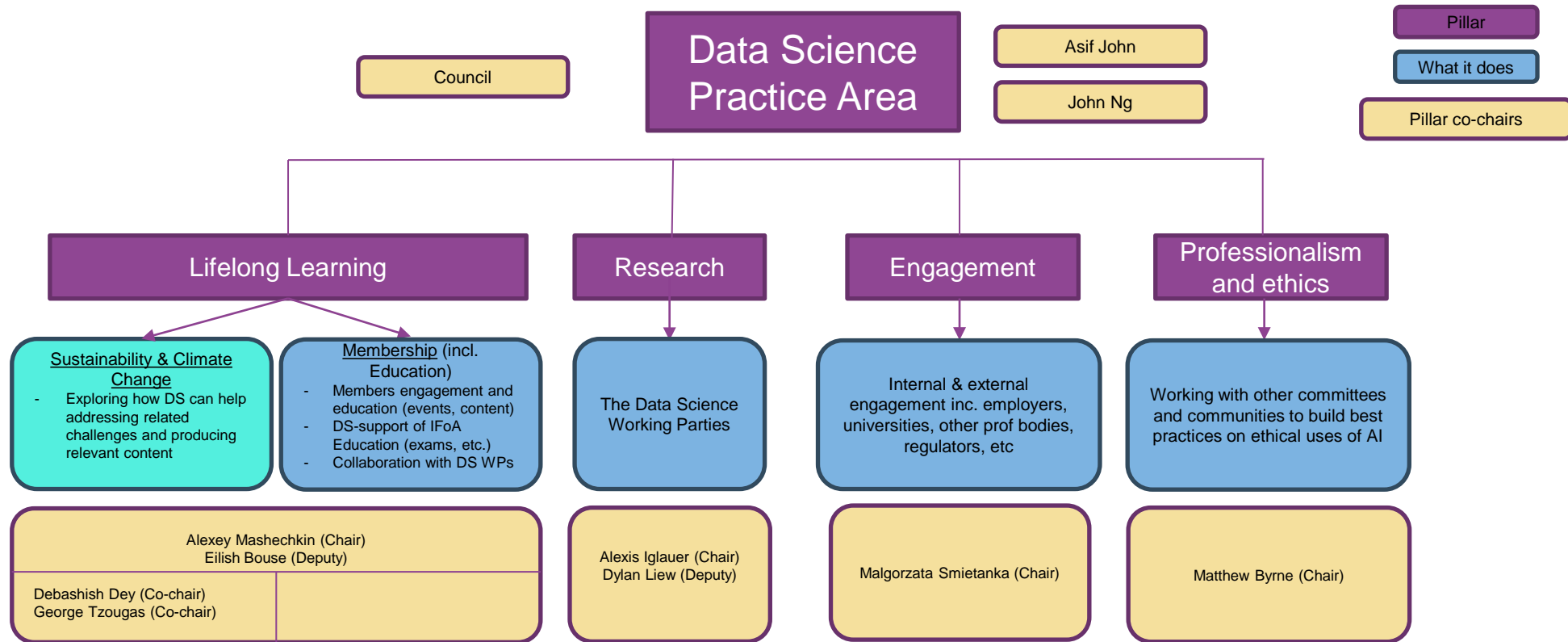


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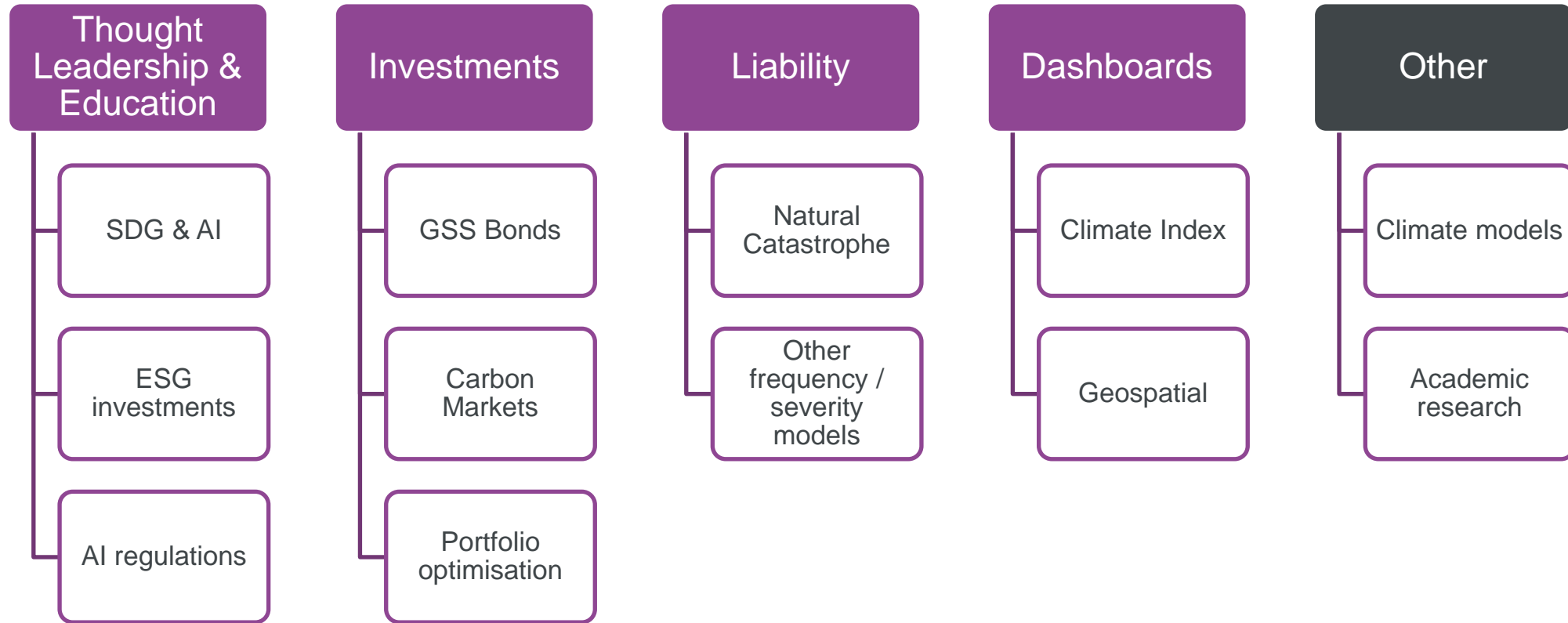
# IFoA Data Science



# Overview of IFoA Data Science structure



# Overview of Working Party projects



# Questions

# Comments

Expressions of individual views by members of the Institute and Faculty of Actuaries and its staff are encouraged.

The views expressed in this presentation are those of the presenter(s).



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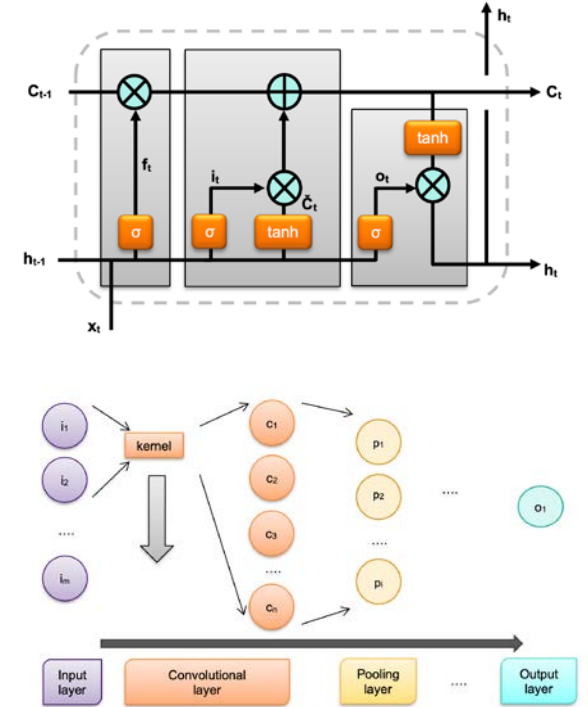
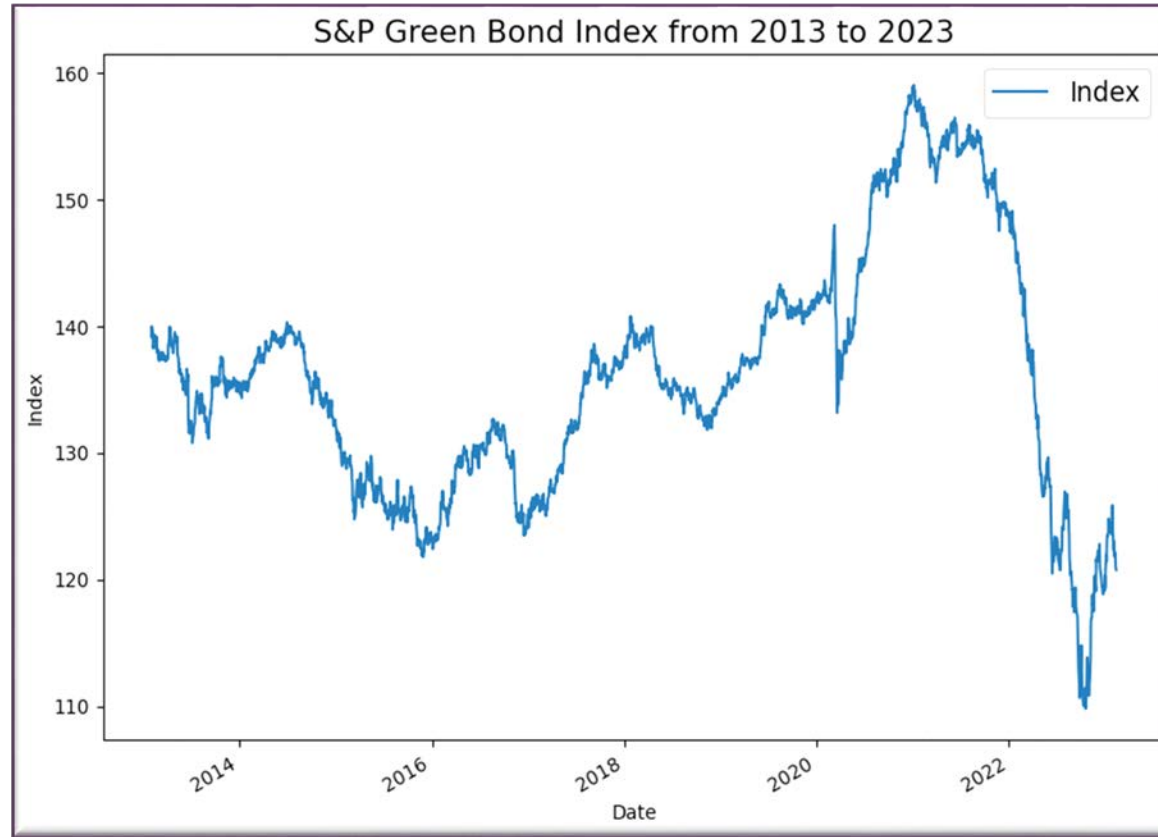
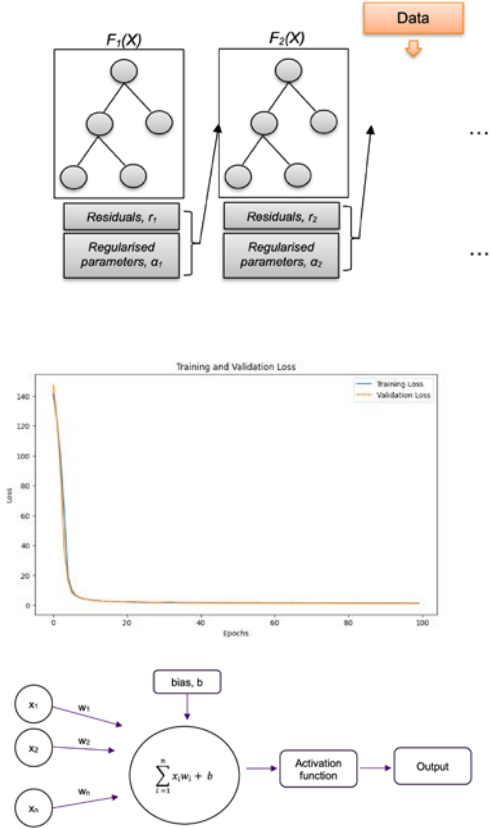
# Appendix



# Appendix: IFoA Data Science – useful links

- IFoA VLE homepage
  - <https://vle.actuaries.org.uk/course/view.php?id=1867>
- IFoA blog posts
  - <https://blog.actuaries.org.uk>
- IFoA life-long learning homepage
  - <https://actuaries.org.uk/learn/lifelong-learning/data-science/>
- Other publications
  - <https://researchportal.hw.ac.uk/en/persons/george-tzougas>

# Appendix: Green bonds



# Appendix: Actuarial index

