

SAS (2) – Climate Finance

# **Mortality Modelling using Temperature Metrics**

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# DRP Context



## Climate Metrics

- Initially for portfolio management
- Pivoted to mortality modelling
- Focus now on Lee-Carter model with temperature data

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# MOTIVATION

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# Climate Metrics

Captures the status of the climate or the impact on the climate. i.e.

Carbon Footprint	ESG Score
Temperature	Natural Disaster Index

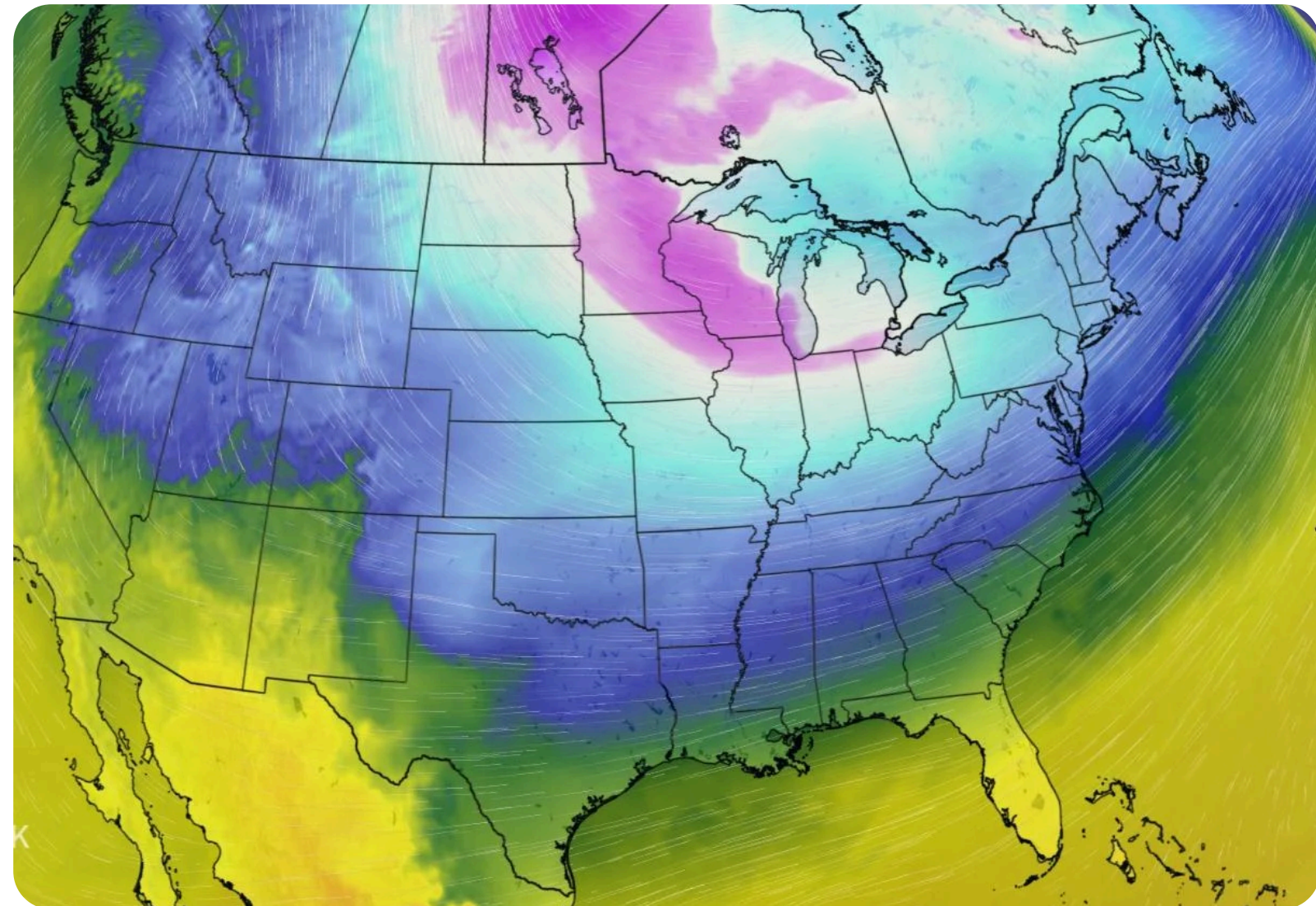
Motivation?  
Pricing Liabilities  
(Life, P&C, Mortgages, etc)

# Core Questions

- **Does temperature affect mortality in Ontario?**
- **How can we model it?**
- **Is it better than traditional methods?**



# Aim of our project



- **Evaluate** the performance of the classic Lee-Carter mortality model using Ontario data (1991–2023) for males and females.
- **Extend** the model by incorporating mean annual temperature derived from daily climate records.
- **Compare** baseline and climate-adjusted models using forecast accuracy metrics (RMSE, MAPE).
- **Analyze** whether temperature trends explain additional variation in mortality rates.
- **Validate** model assumptions through residual analysis.



# Data Sources

## Deaths

Yearly Deaths from 1991–2023 for age bins from 20–89

- Types:
- Male
  - Female

## Exposure

Yearly Exposure from 1991–2023 for age bins from 20–89

- Types:
- Male
  - Female

## Temperature

Mean Annual Temperature for Toronto from 1991–2023

	1991	1992	1993	1994
22	1061978	1051986	1041399	1030001
27	1272329	1230926	1182339	1136797
32	1301807	1310907	1322893	1331842
37	1174389	1204928	1239438	1264078
42	1069653	1074605	1089326	1116243
47	844293	900503	946112	987468
52	675039	698991	730094	760293
57	616719	613492	616800	628825
62	579778	587841	592319	594584
67	498506	502964	510437	515979
72	364432	381693	396877	412194
77	255667	259131	260594	261541
82	142053	147879	153830	160060
87	62040	63705	65832	68117

Year	Average Temperature
1991	8.940109589
1992	7.036448087
1993	7.261534247
1994	7.405068493
1995	7.878383562
1996	7.180081967
1997	7.586821918
1998	10.18816438
1999	9.493369863

# What is Lee-Carter Model

- A statistical model used in demography and actuarial science to project mortality rates over time.
- Separates mortality patterns into **age effects** and **time effects**.

$$\ln m_{x,t} = a_x + b_x k_t$$

- $m_{x,t}$ : Mortality rate at age  $x$ , year  $t$
- $a_x$ : Average mortality pattern by age
- $b_x$ : Sensitivity of each age group to overall mortality changes
- $k_t$ : Overall mortality trend index (captures changes over time)
- Widely adopted by actuaries and statisticians for forecasting because it's simple, interpretable, and performs well with long-term historical data.
- Forms the basis for many national life tables and insurance pricing models.

# Model Framework

We extend the Lee–Carter mortality model from:

$$\ln m_{x,t} = a_x + b_x k_t$$

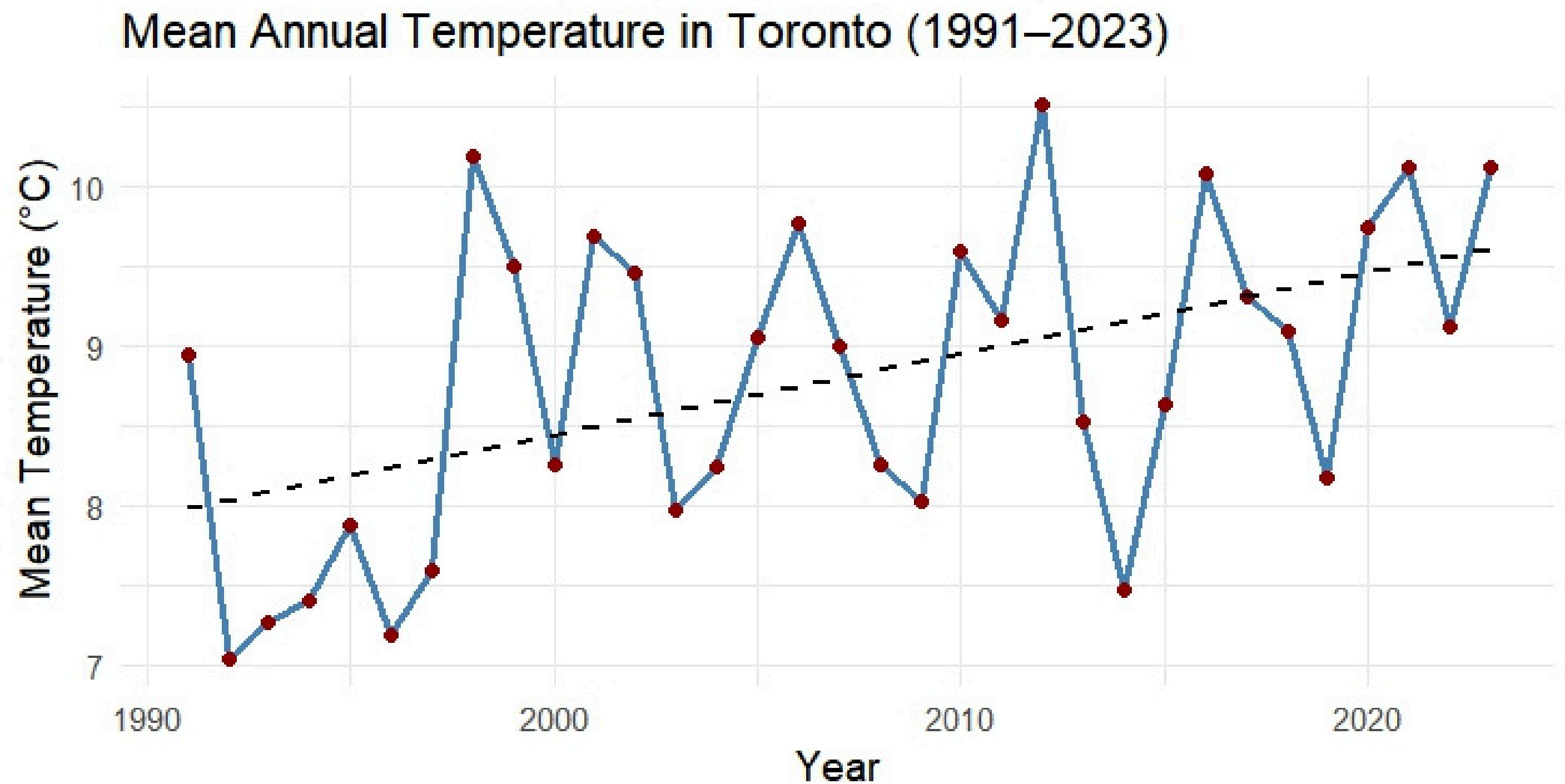
To:

$$\ln m_{x,t} = a_x + b_x k_t + c_x T_t$$

Where:

- $m_{x,t}$ : mortality rate at age  $x$  in year  $t$
- $a_x$ : age-specific average log mortality
- $b_x$ : sensitivity of mortality to the time-varying index  $k_t$
- $k_t$ : time-varying mortality index
- $c_x$ : sensitivity of mortality to mean annual temperature  $T_t$

# Mean Annual Temperature Trend





# ONTARIO MORTALITY RESULTS

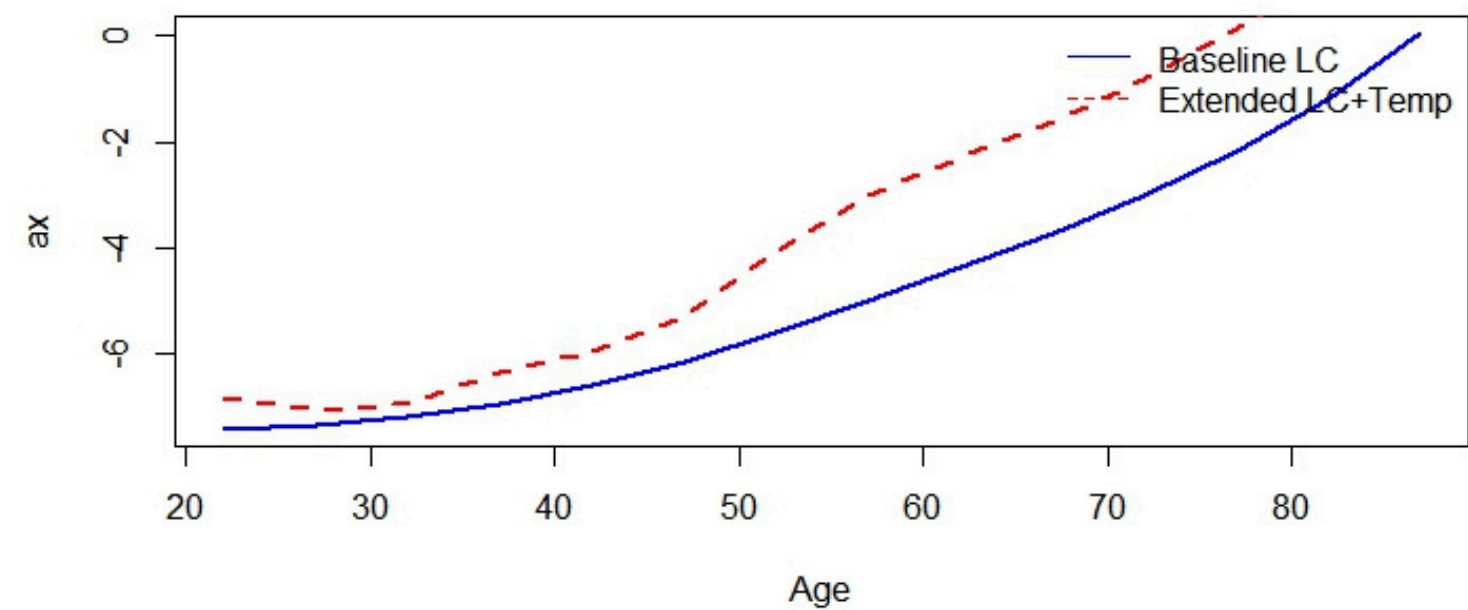
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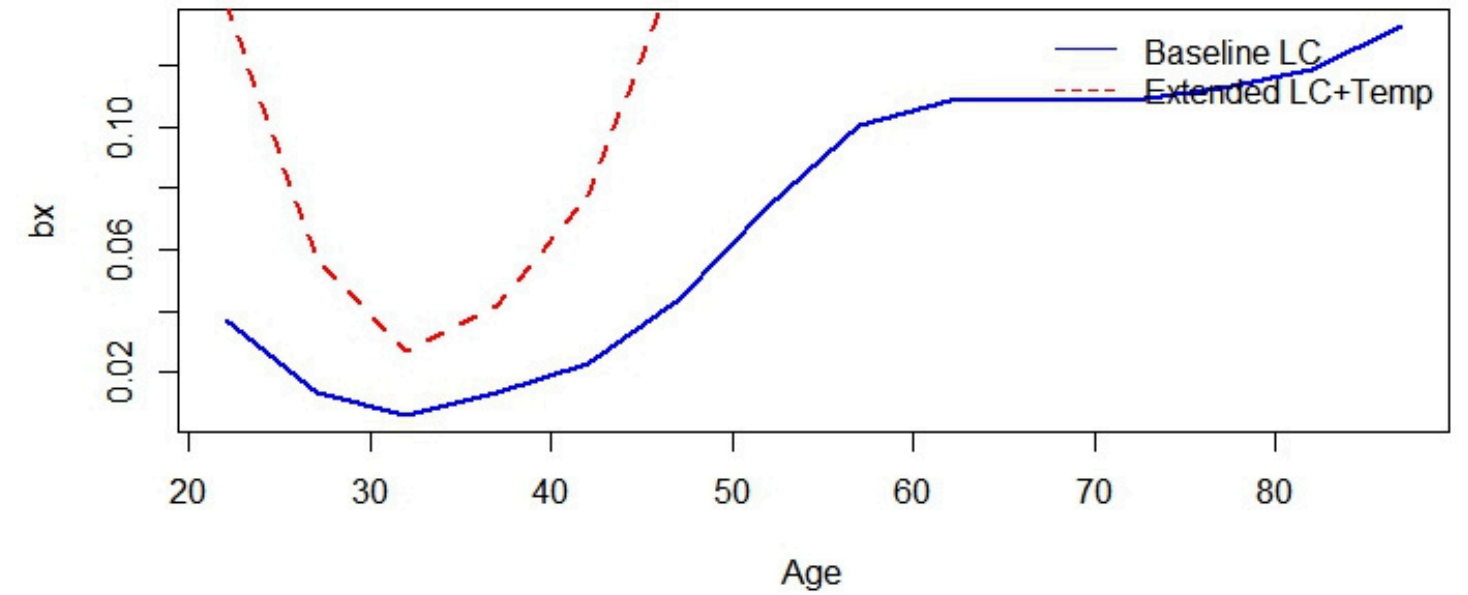
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# Baseline vs Climate Model

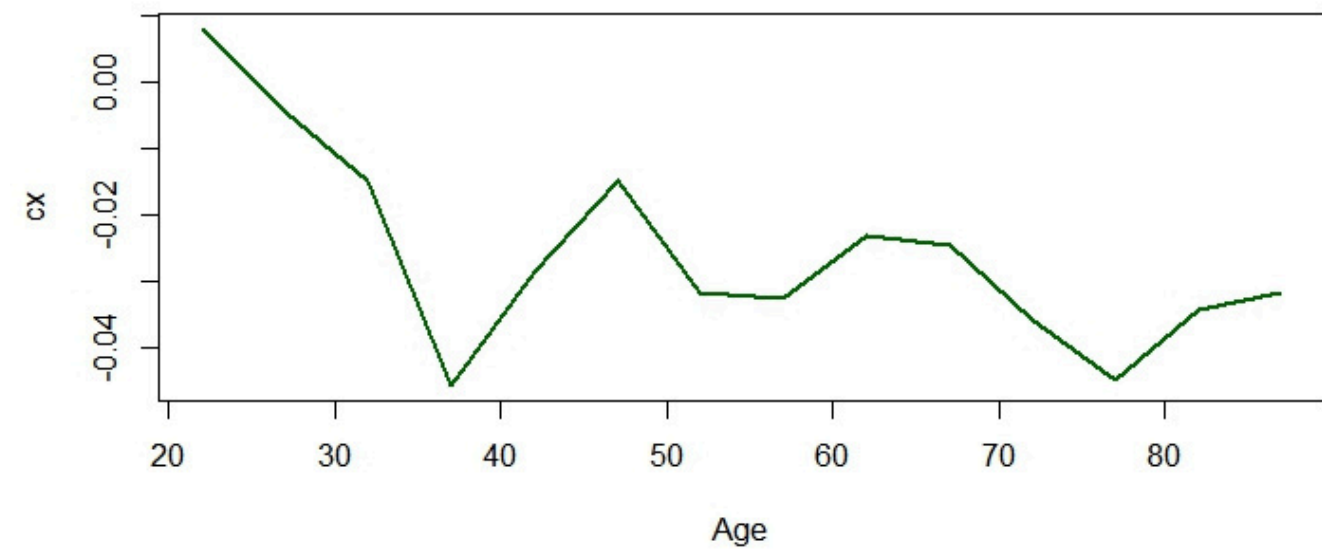
ax (Age Effect) - male



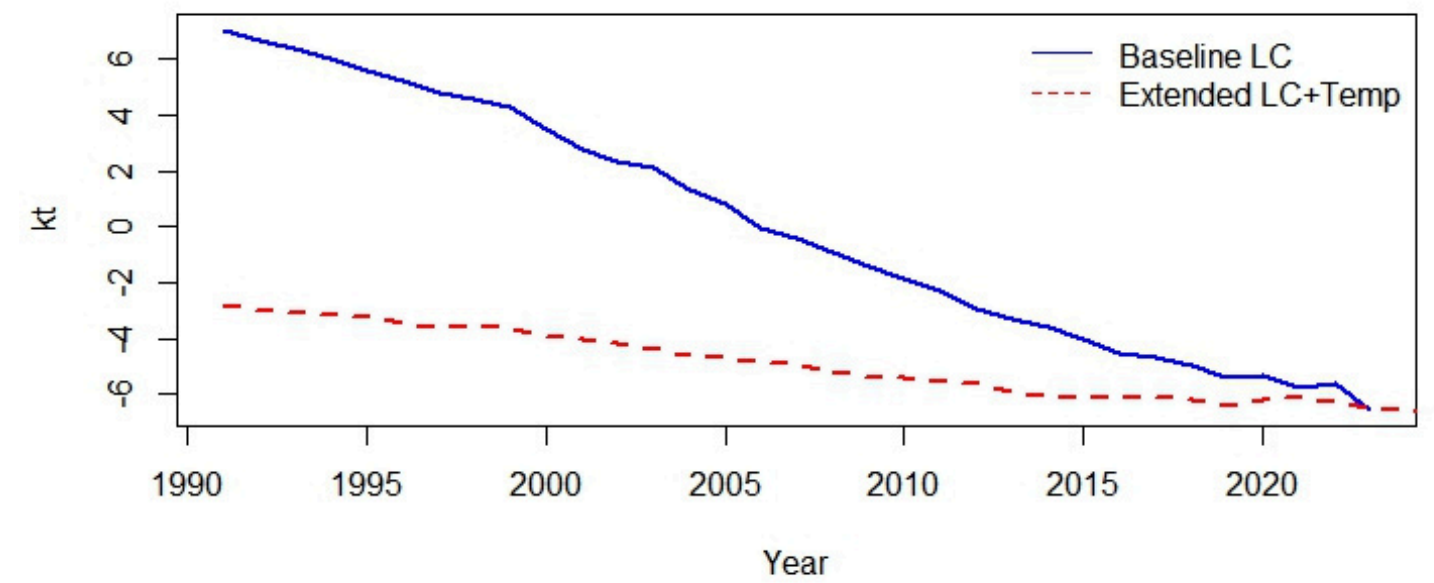
bx (Age Sensitivity) - male



cx (Temp Sensitivity) - male

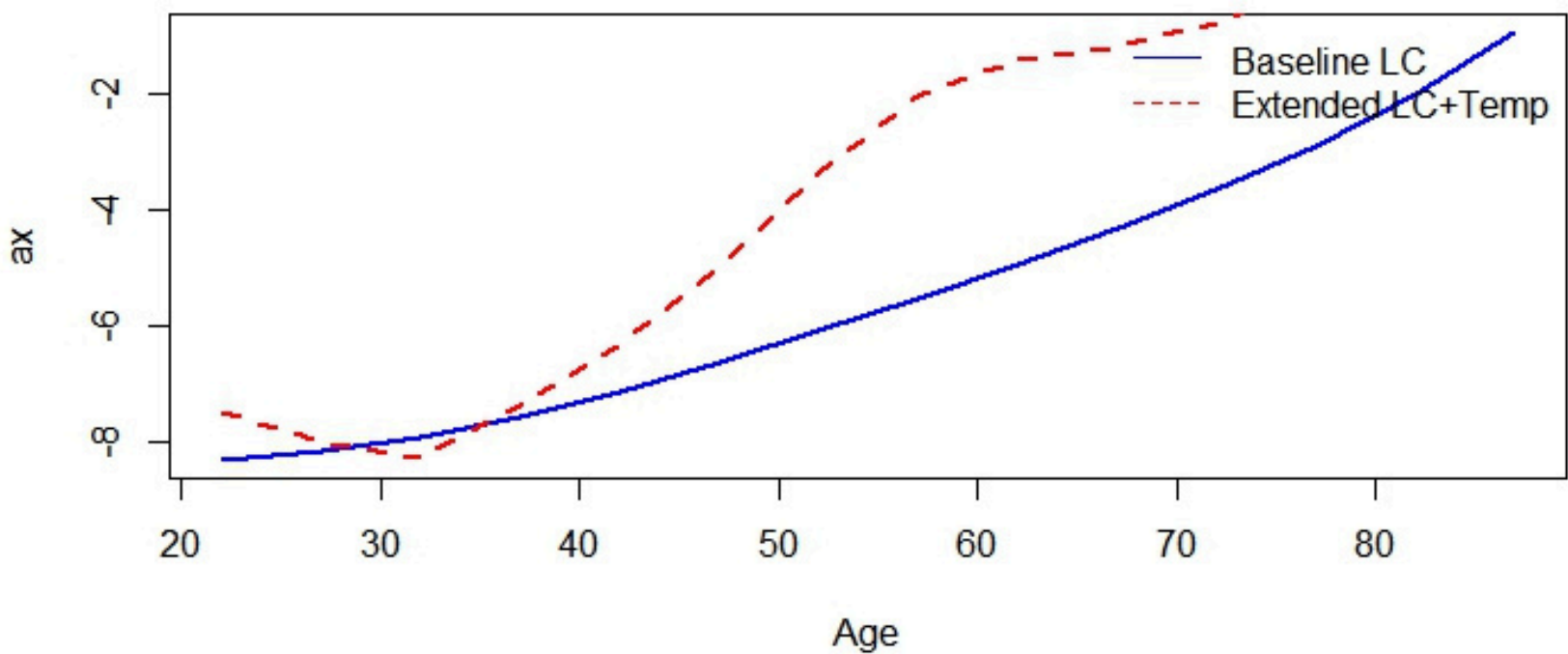


kt (Time Index) - male

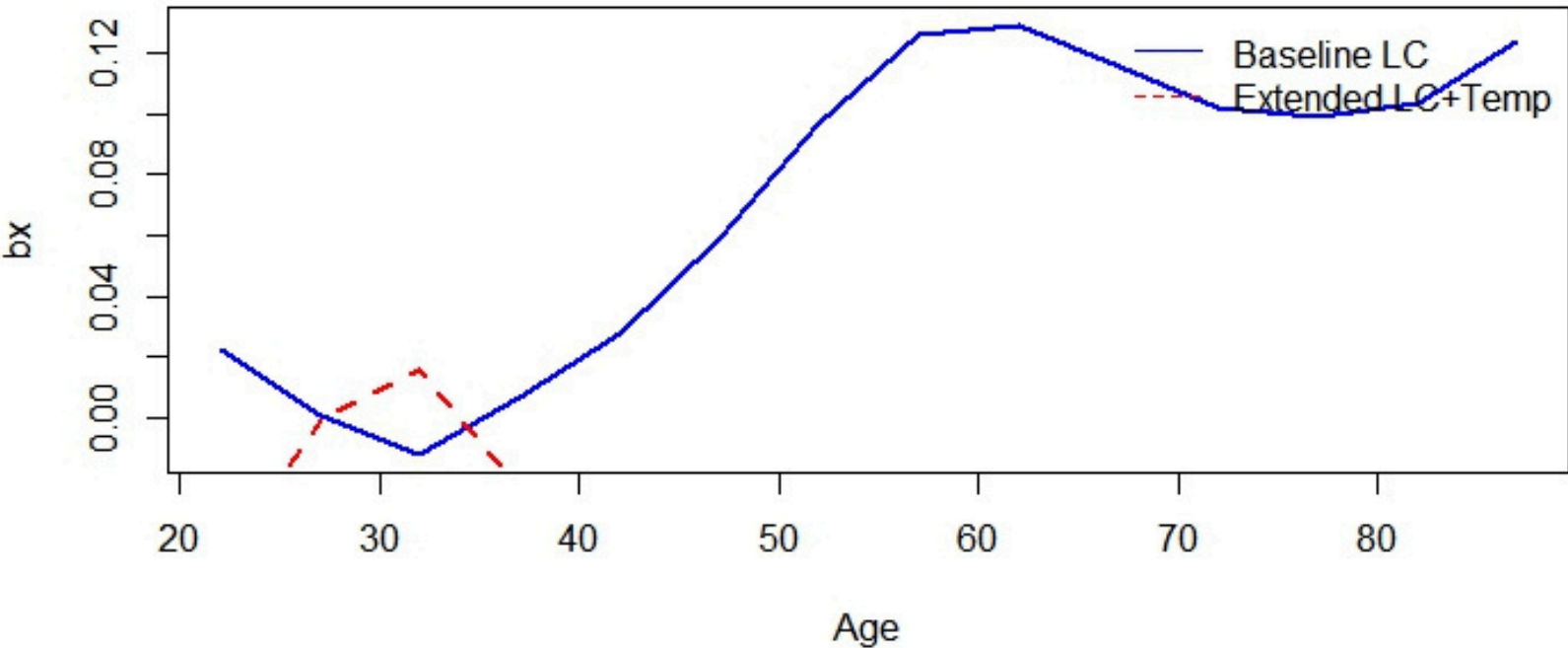


# Baseline vs Climate Model

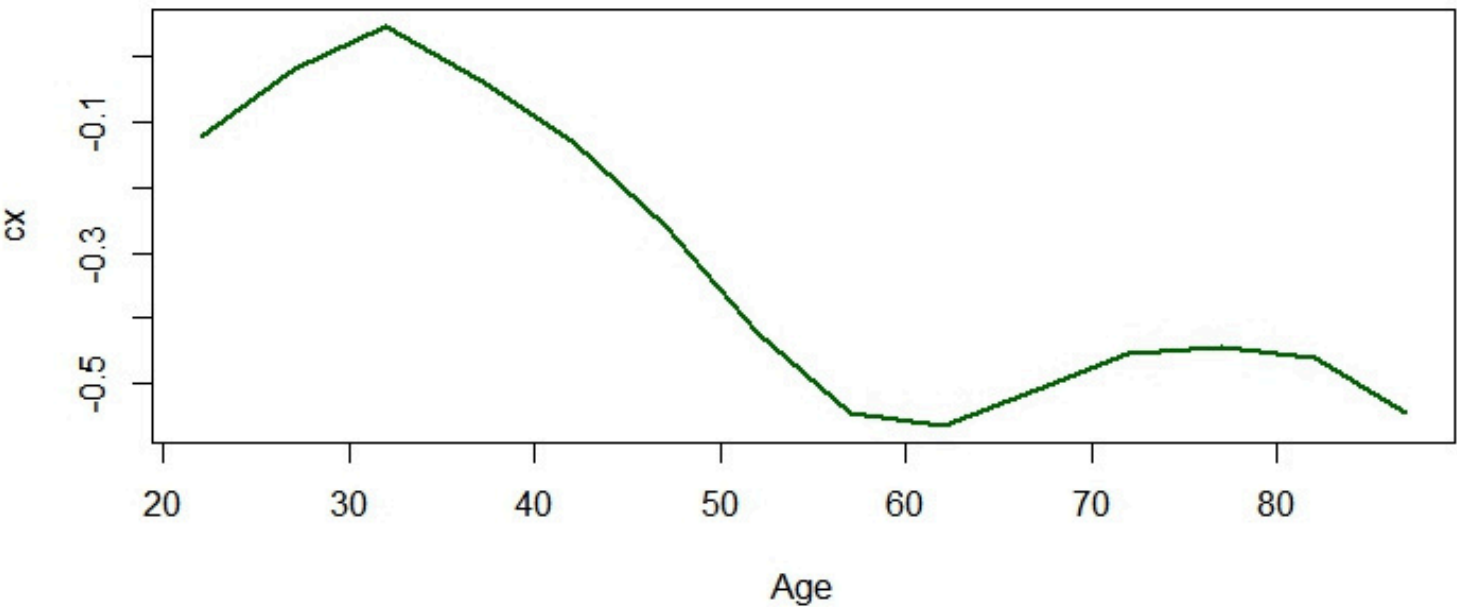
ax (Age Effect) - female



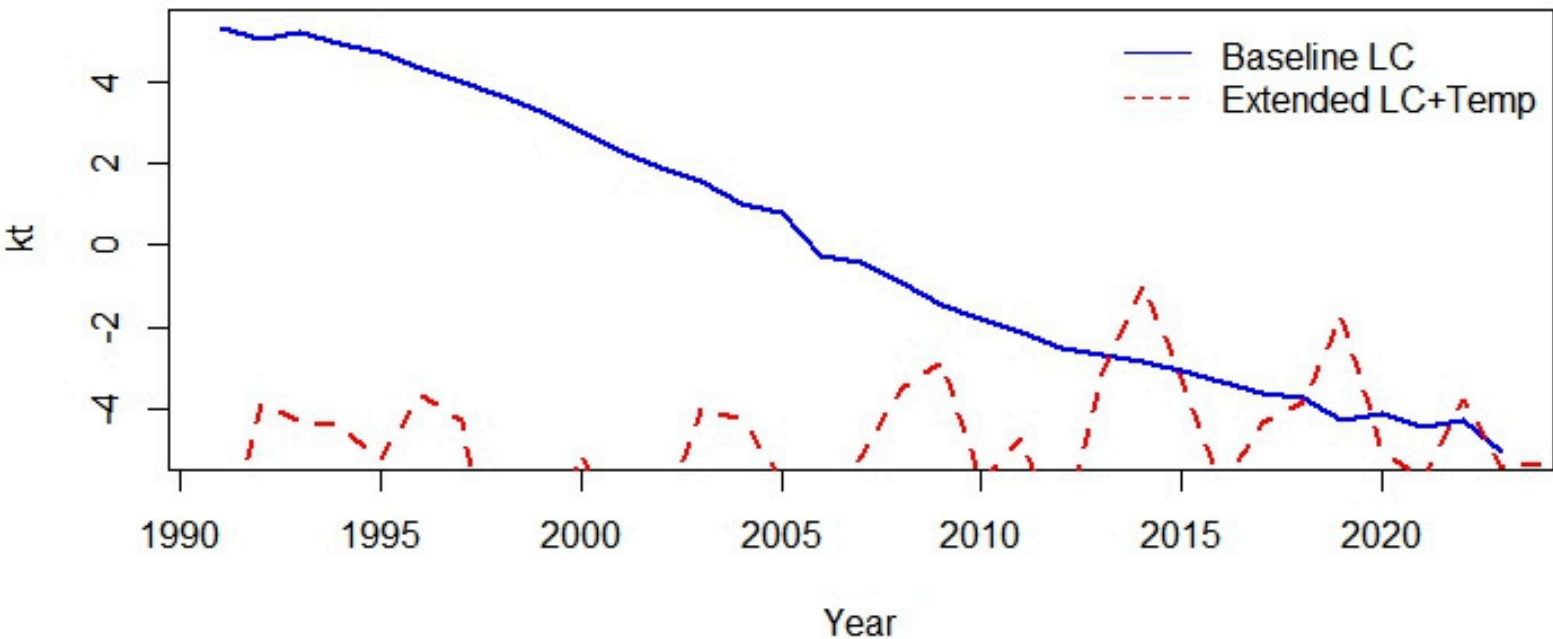
bx (Age Sensitivity) - female



cx (Temp Sensitivity) - female



kt (Time Index) - female



# FIT ANALYSIS

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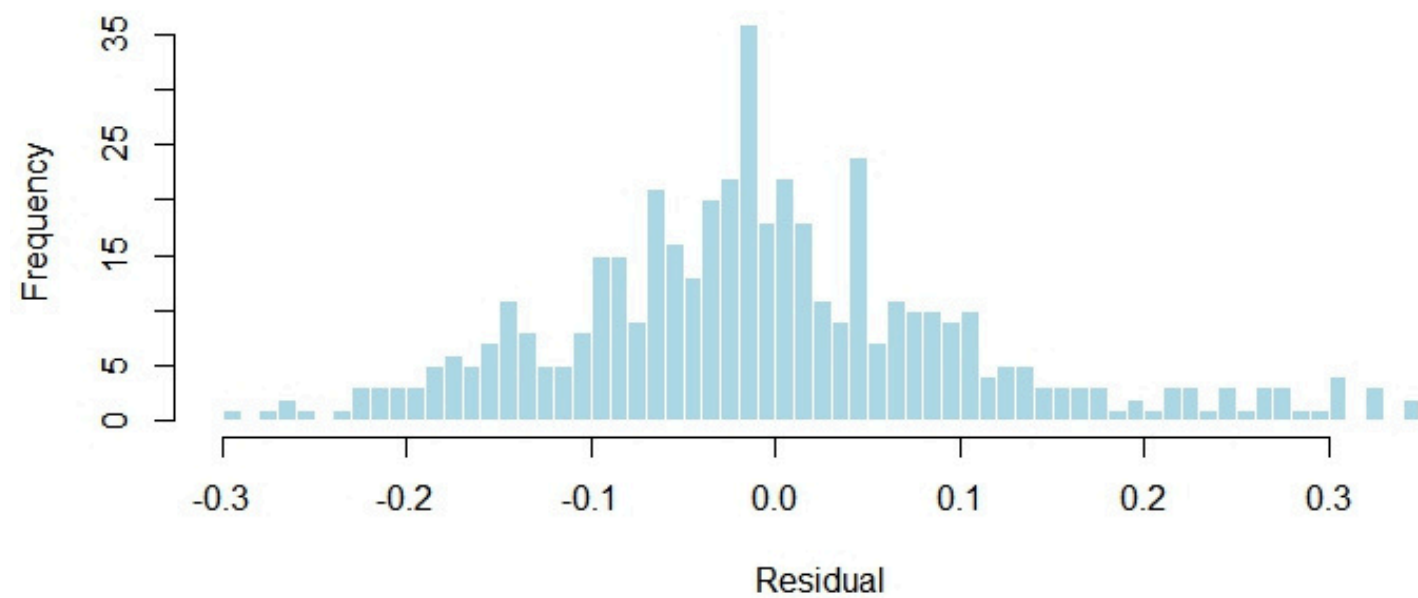
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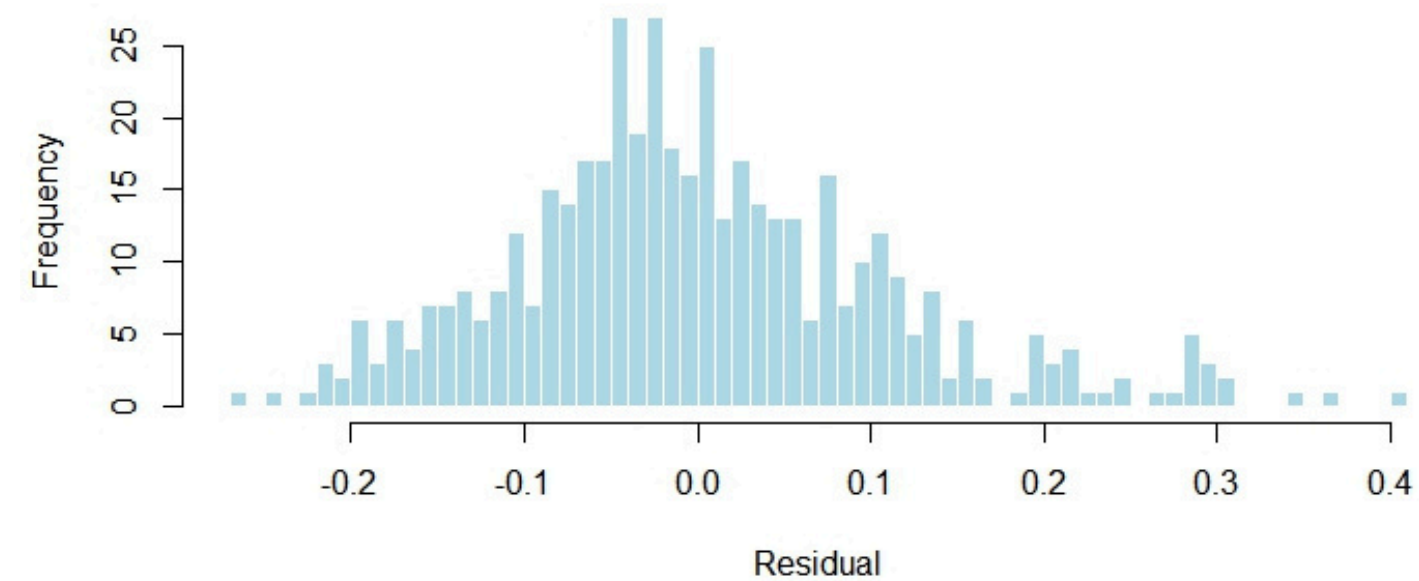


# Residual Analysis

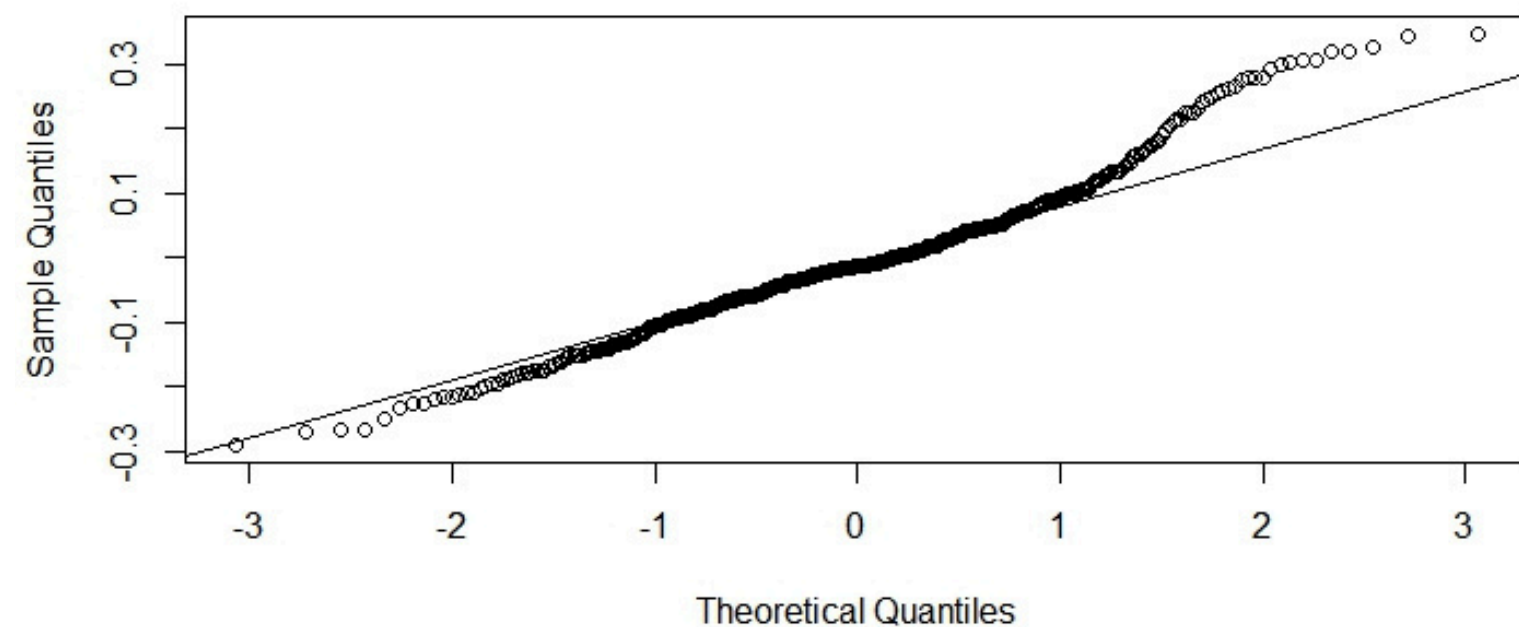
Residual Histogram - Baseline LC - male



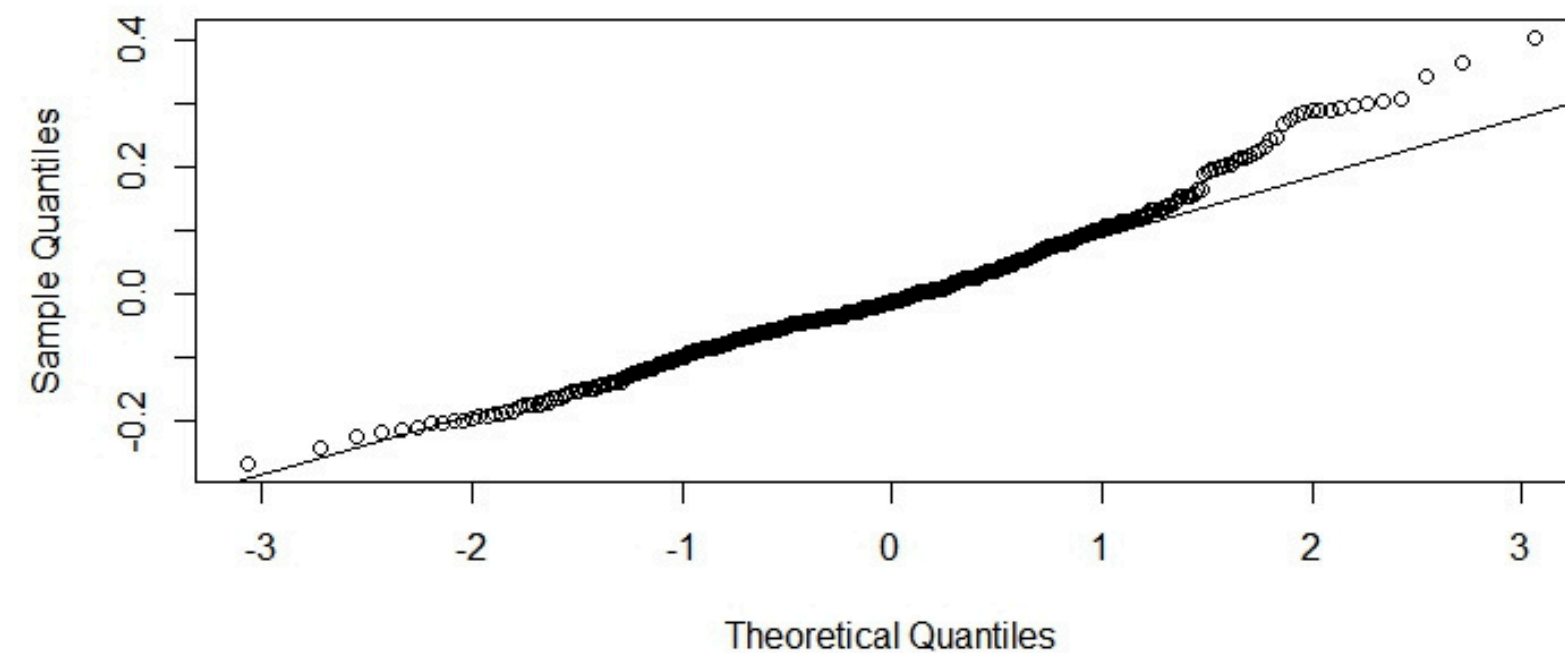
Residual Histogram - LC+Temp - male



Q-Q Plot - Baseline LC - male

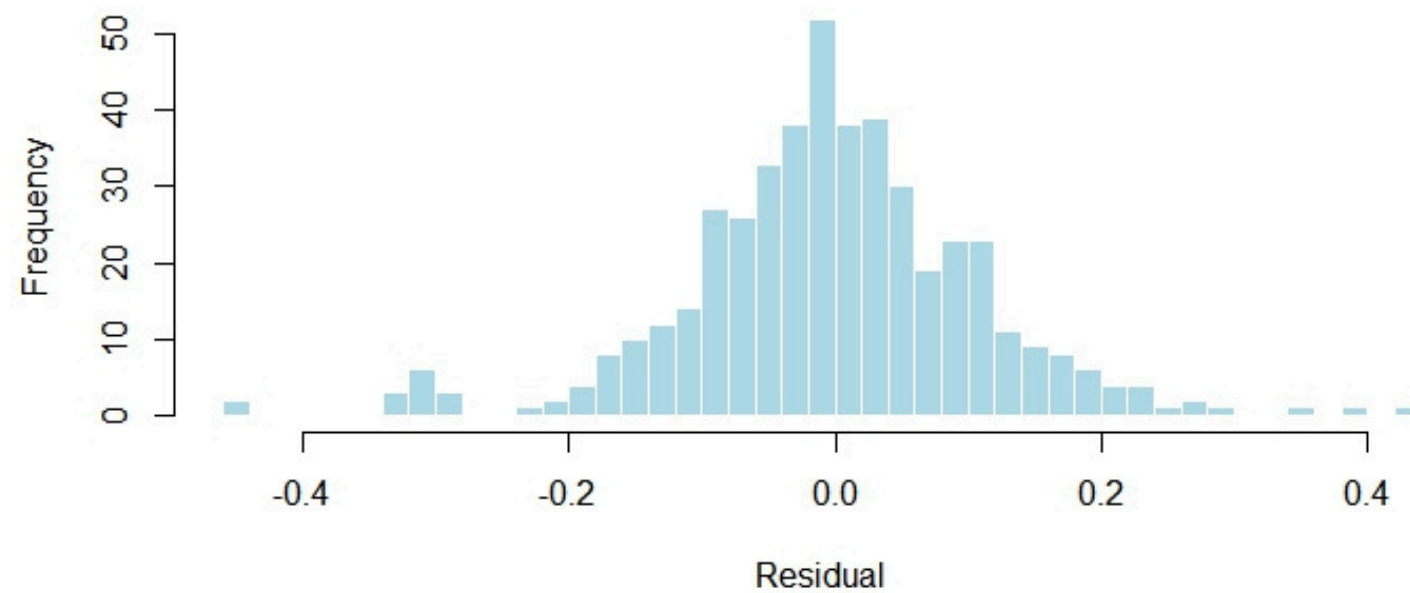


Q-Q Plot - LC+Temp - male

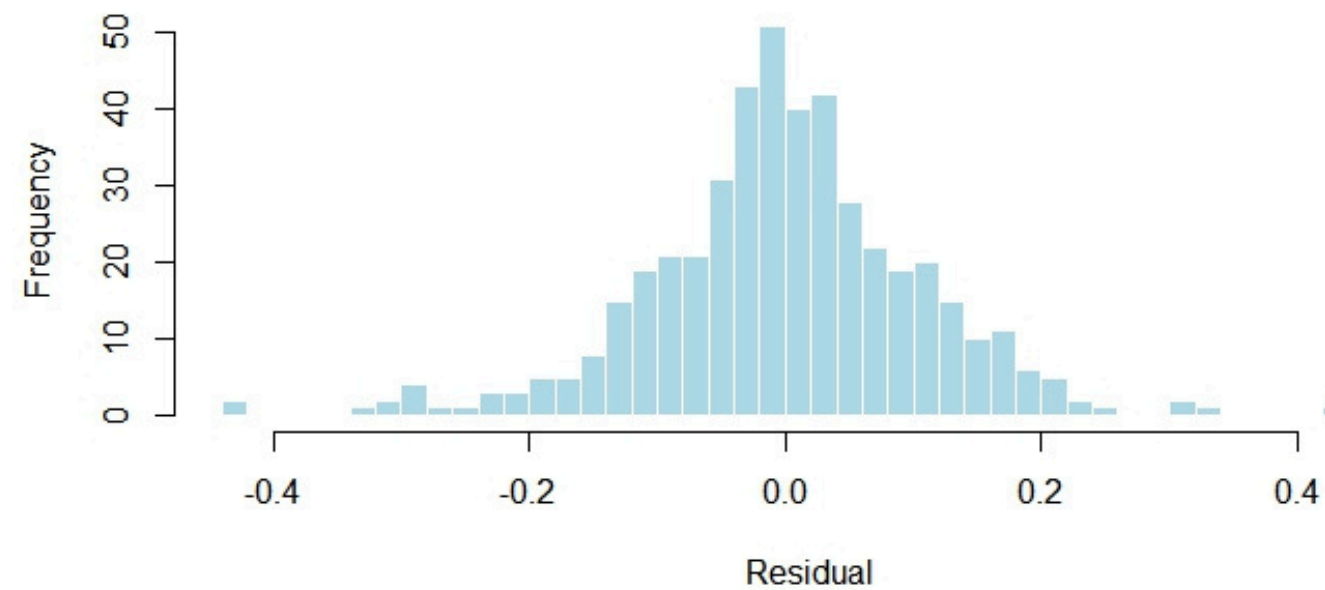


# Residual Analysis

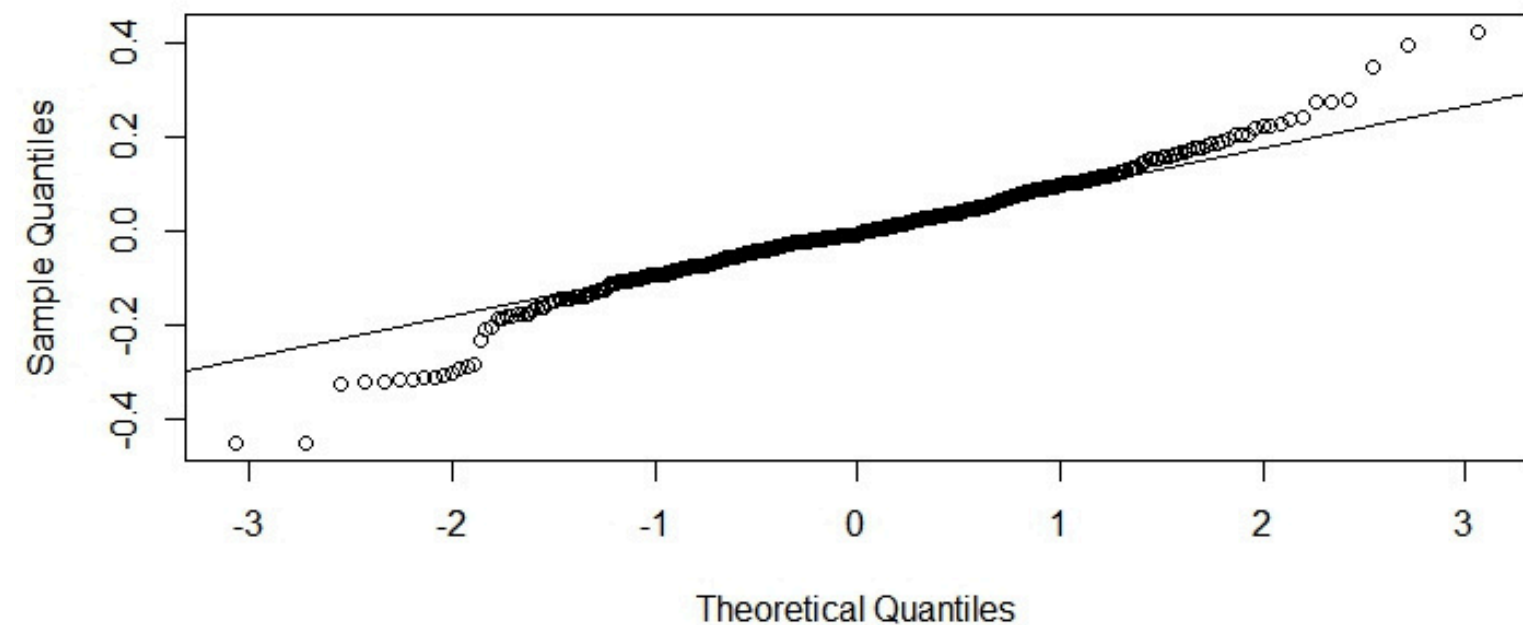
Residual Histogram - Baseline LC - female



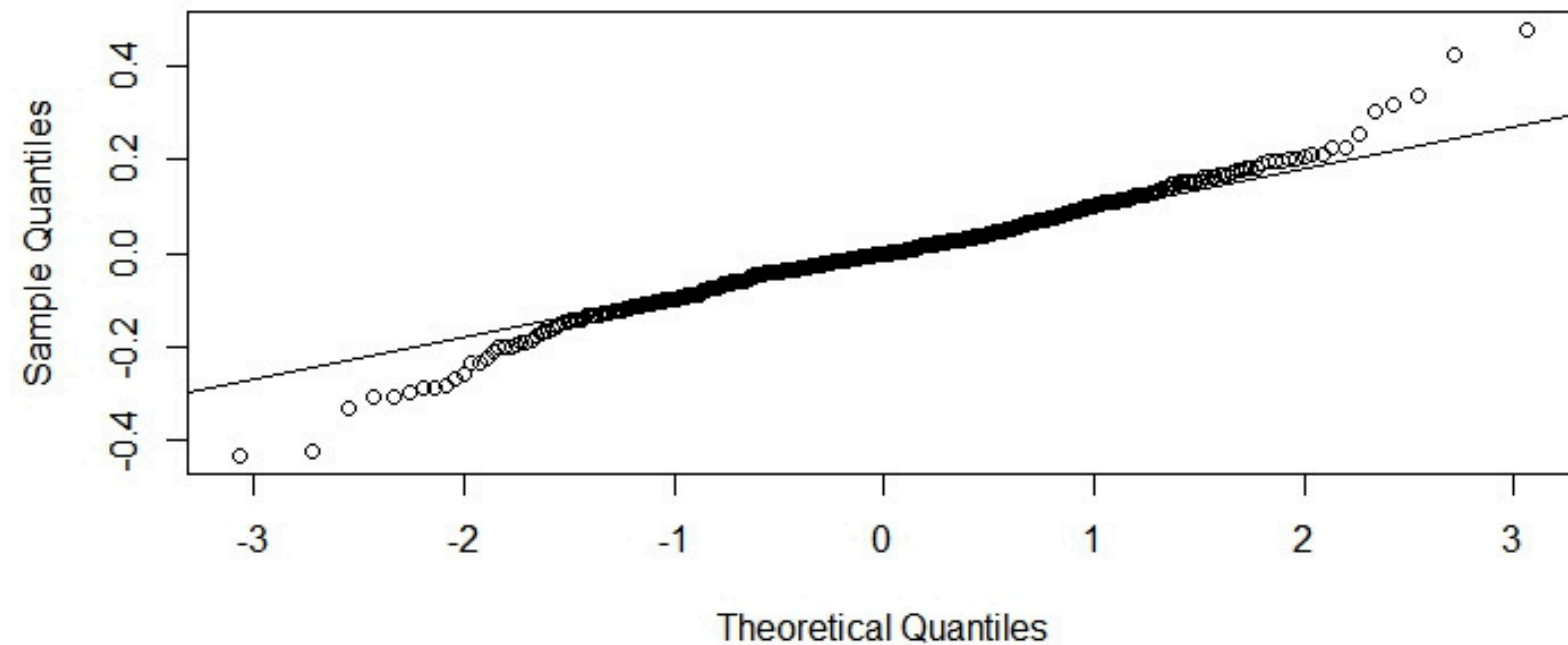
Residual Histogram - LC+Temp - female



Q-Q Plot - Baseline LC - female



Q-Q Plot - LC+Temp - female



# Fit Error Metrics

Male

Baseline	Climate
RMSE: 11.37%	RMSE: 10.94%
MAPE: 8.53%	MAPE: 8.29%

Female

Baseline	Climate
RMSE: 11.16%	RMSE: 10.92%
MAPE: 8.24%	MAPE: 8.04%

## Key Insights

- 1.Consistent Improvement
- 2.Modest but real gains
- 3.Slight gender differences
- 4.Overall model quality

# TAKEAWAYS

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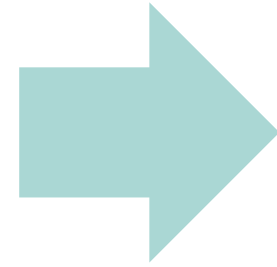
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# The Apparent Paradox

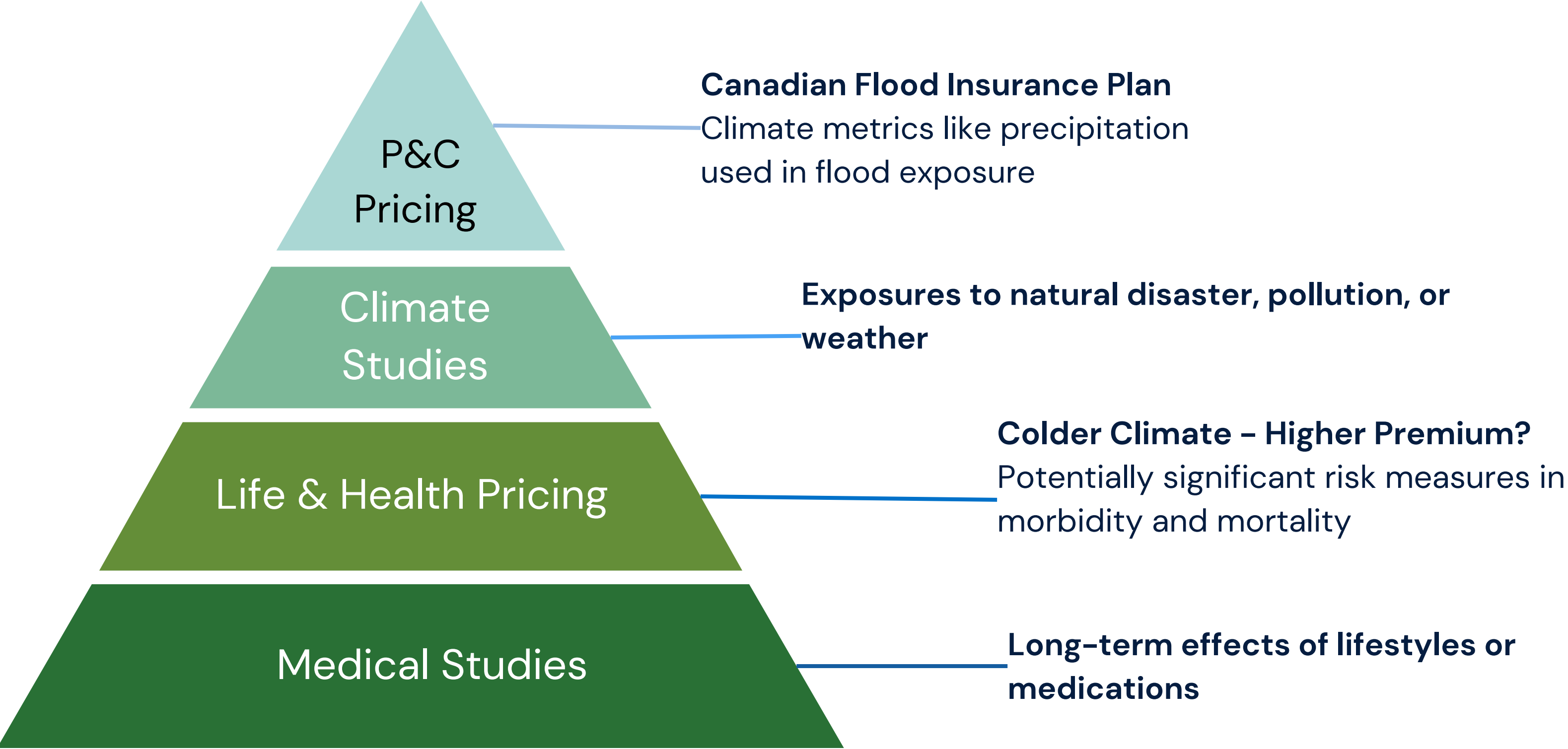
What we know:

- **cx**: Shows warmer temperatures are generally protective
- **Temperature trend**: Ontario has been warming from 1991–2023
- **kt**: The extended model shows less mortality improvement than the baseline model



If warming temperatures are protective AND Ontario is warming, why isn't there MORE mortality improvement in the extended model?

# CLIMATE METRIC APPLICATIONS

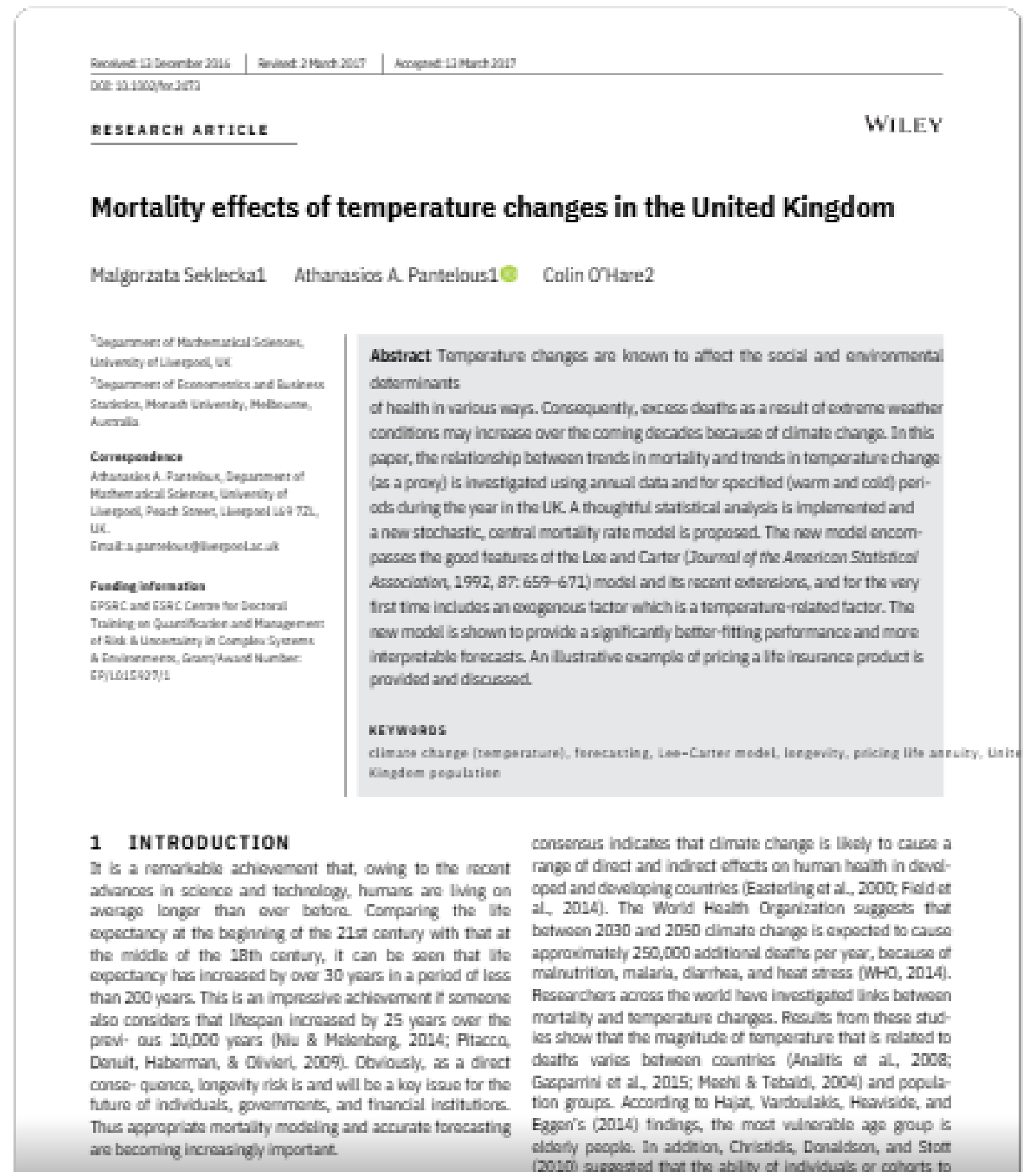


# FURTHER READINGS

SEKLECKA, PANTELOUS, AND O'HARE (2017)

## Mortality effects of temperature changes in the United Kingdom

- Extended LC Model with temperature metric
- Performed better out-of-sample forecasts than base LC (Improves BIC by over 8%)
- Most significant effects found in elderly populations



# FURTHER READINGS

## Effect of Climate Metrics on Insurance Pricing

### Academic Papers - Climate Metrics in Insurance Pricing

#### ESG

Carannante, A., D'Amato, V., & Staffa, M. S. (2024). How ESG corporate reputation affects sustainability premiums in the Insurance Industry. *Frontiers in Applied Mathematics and Statistics*, 10. <https://www.frontiersin.org/journals/applied-mathematics-and-statistics/articles/10.3389/fams.2024.1474505/full#frontiersin>

Balley, M., & Wittenberg, D. (2025). Why Insurers Price Carbon Low: An Analysis of Financed Emissions and Investment Decisions. *SSRN Working Paper*. (No direct link found; usually available at SSRN.com)

Flores, R. M., Lee, S., & Fischer, A. (2023). The role of insurance status in the association between temperature and MI hospitalizations. *Environmental Health*, 22(3), 101-112. <https://pmc.ncbi.nlm.nih.gov/articles/PMC10403039/> and <https://pubmed.ncbi.nlm.nih.gov/37545806/pmc.ncbi.nlm.nih+1>

#### Natural Disaster Risk Index

Keys, B. J., & Mulder, T. (2024). Disaster Risk and Rising Home Insurance Premiums. *NBER Working Paper No. 32579*. <https://www.nber.org/papers/w32579> and summary: <https://www.nber.org/digest/202410/disaster-risk-and-rising-home-insurance-premiums#nber+1>

Montero, E. (2024). Natural disasters, stock price volatility in the property insurance sector. *Journal of Insurance and Risk Management*, 19(2), 132-153. (No direct link found)

Q, Z., Wang, H., & Yang, S. (2022). Exposure to Abnormally Hot Temperature and the Demand for Commercial Health Insurance. *Frontiers in Public Health*, 10, 842005. <https://www.frontiersin.org/journals/public-health/articles/10.3389/fpubh.2022.842005/full> and <https://pubmed.ncbi.nlm.nih.gov/39145844/pubmed.ncbi.nlm.nih+1>

- ESG
- Carbon Footprint
- Temperature
- Natural Disaster Index

# Conclusion

**Replicated UK  
Mortality study  
with Ontario  
population**

**Extension of  
Lee-Carter with  
Temperature**



# Thanks!

ANY QUESTIONS?