



# The impacts of unusual weather on the Gross Domestic Product (GDP)

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Session 2: Measuring climate change adaptation

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# Outline

Econometric model -> Relevant sectors -> Weather variables

Obtained results -> Recommendations for further research



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Report (2014; Pim Ouwehand and Floris van Ruth): to adjust **GDP** also for **unusual weather effects** in order to distinguish their **irregular behavior** from structural economic changes and to estimate the **magnitude** of unusual weather effects.

<https://www.cbs.nl/en-gb/background/2014/12/how-unusual-weather-influences-gdp>

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**Climate change adaptation**: Which sectors are at **risk** and which sectors will **profit** from climate change? -> Update of the former report (2020; Pim Ouwehand), in Dutch:

<https://www.cbs.nl/nl-nl/maatwerk/2020/48/klimaatimpact-op-de-economie>



# The econometric model

**ARIMA**      **AutoRegressive Integrated Moving Average**

Is normally used to calculate a **seasonally adjusted GDP** (quarterly time serie per sector).

To include **unusual weather** some extra variables are added. They are transformed from daily to quarterly variables.

Per sector: several **iterations** towards an **optimal model**.



# Sectors with significant unusual weather effects

**Cold weather** increases the heating needs

B Mining and Quarrying

D Energy (Electricity, gas, steam and air condition supply)

**Severe frost** hinders activities

C Manufacturing

F Construction

**Warm weather** leads to a sunny 'going out' mood

I Accommodation and food service activities

Many unusual weather effects will tend to average out over a quarter.

Retail sales will catch up shortly after a very rainy week with less sales.



# Unusual weather; available indicators

Unusual weather: deviations from the long term average.

Impact on GDP: detectable at sector level on a quarterly basis.

## Available weather indicators

Temperature

daily average, min, max

Sunshine

number of hours, day-time %

Rainfall

daily sum, number of hours

Wind speed

daily average, 1h max

Snow

height



# Unusual weather; regression variable 'cold'

**Cold weather** increases heating needs. Only heating needed when the daily average is below 18 degrees Celsius.

'Degree days' = 18 minus 'daily average', if it is below 18°C  
it is 0, if it is 18°C or higher

Daily average T	...	20	8	-1	-4	-8	...
'Degree days'	...	0	10	19	22	26	...

From daily to quarterly data: take the sum of 'degree days'





# Unusual weather; regression variable 'frost'

**Severe frost** hinders activities. Count the days in a quarter with temperatures below  $0^{\circ}\text{C}$ , below  $-3^{\circ}\text{C}$ , or below  $-7^{\circ}\text{C}$ .

Daily average T	...	20	8	-1	-4	-8	...
Below $0^{\circ}\text{C}$	...	0	0	1	1	1	...
Below $-3^{\circ}\text{C}$	...	0	0	0	1	1	...
Below $-7^{\circ}\text{C}$	...	0	0	0	0	1	...
Below $0^{\circ}\text{C}$	...	0	0	1	4	8	...
Below $-3^{\circ}\text{C}$	...	0	0	0	4	8	...
Below $-7^{\circ}\text{C}$	...	0	0	0	0	8	...



# Unusual weather; regression variable 'warm'

**Warm weather** leads to a sunny 'going out' mood. More drinks, meals, and hotel bookings.

Best proxy      maximum Temperature

## Alternatives

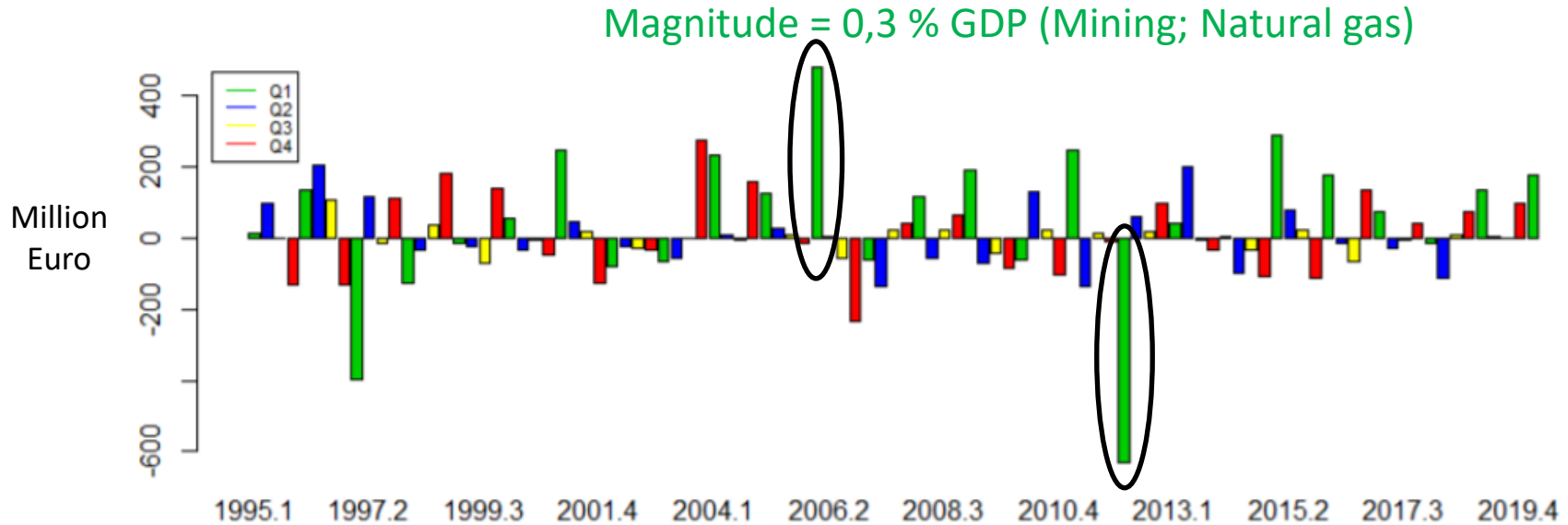
Sunny days      count 1 if 'day-time % sun' > 50%, otherwise 0

Nice days      count 1 if sunny day AND average T > 18°C

Rainy days      count 1 if 'number of hours rain' > 6



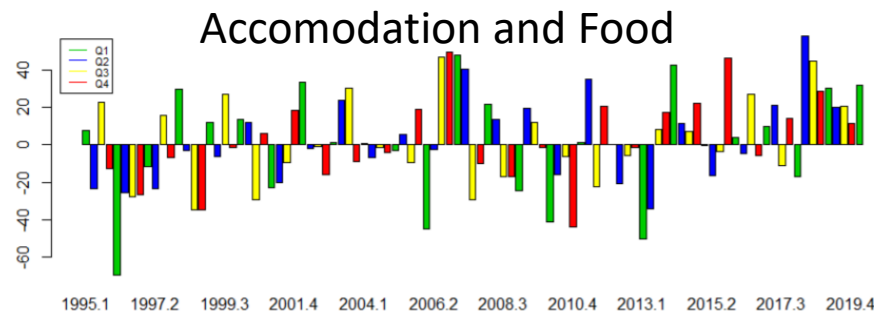
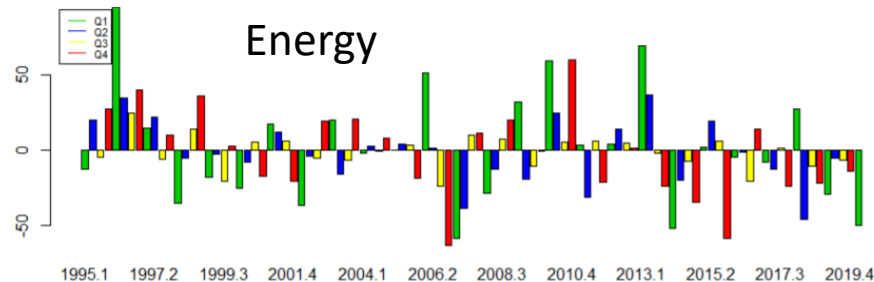
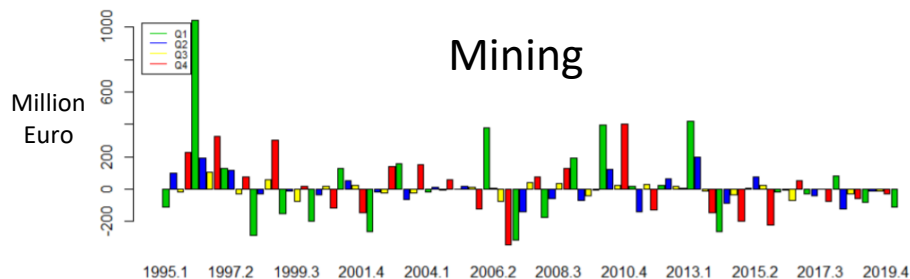
# Impact unusual weather on GDP



Magnitude = -0,4 % GDP  
(Construction and Manufacturing)



# Impact unusual weather on GDP (sectors)



## Different y-axis values

Mining	-200	1000	Energy	-50	50
Construction	-600	0	Acc. & Food	-60	40
Manufacturing	-250	50	<b>GDP</b>	<b>-600</b>	<b>400</b>



# Recommendations

- The transformation from weather indicators (daily) to regression variables (quarterly) is **not trivial**. Also the iteration process towards an optimal model needs **more standardisation**. Future models should also take into account the **Covid-19** effects.
- **More sectors** should be investigated, like **Agriculture** (drought effects). For the already investigated sectors the modeling could also be done for **lower aggregates**, where unusual weather effects might be more prominent (bottom-up approach).
- Unusual weather has both **positive** and **negative** effects on economic growth (GDP). Adaptation measures should not only be focused on sectors at **risk** (minimizing damages), but also on sectors that **profit** from unusual weather (maximizing economic opportunities).
- **More budget** is needed for additional (academic) research. The obtained results are not mature and should be seen as an **inspiring** first exercise.

