

A day in a sustainable life.

From data to decision-making: decarbonizing electricity demand.

Hussain Kazmi, PhD

Attila Balint

Jolien Despeghel

Meet π



Lives in the UK



Concerned about the environment



Already takes public transport to work, but would like to live even more sustainably

Enter the built environment



During construction phase

Building materials and source
Building type (standalone, apartment, ...)

±50 tons of embodied CO₂

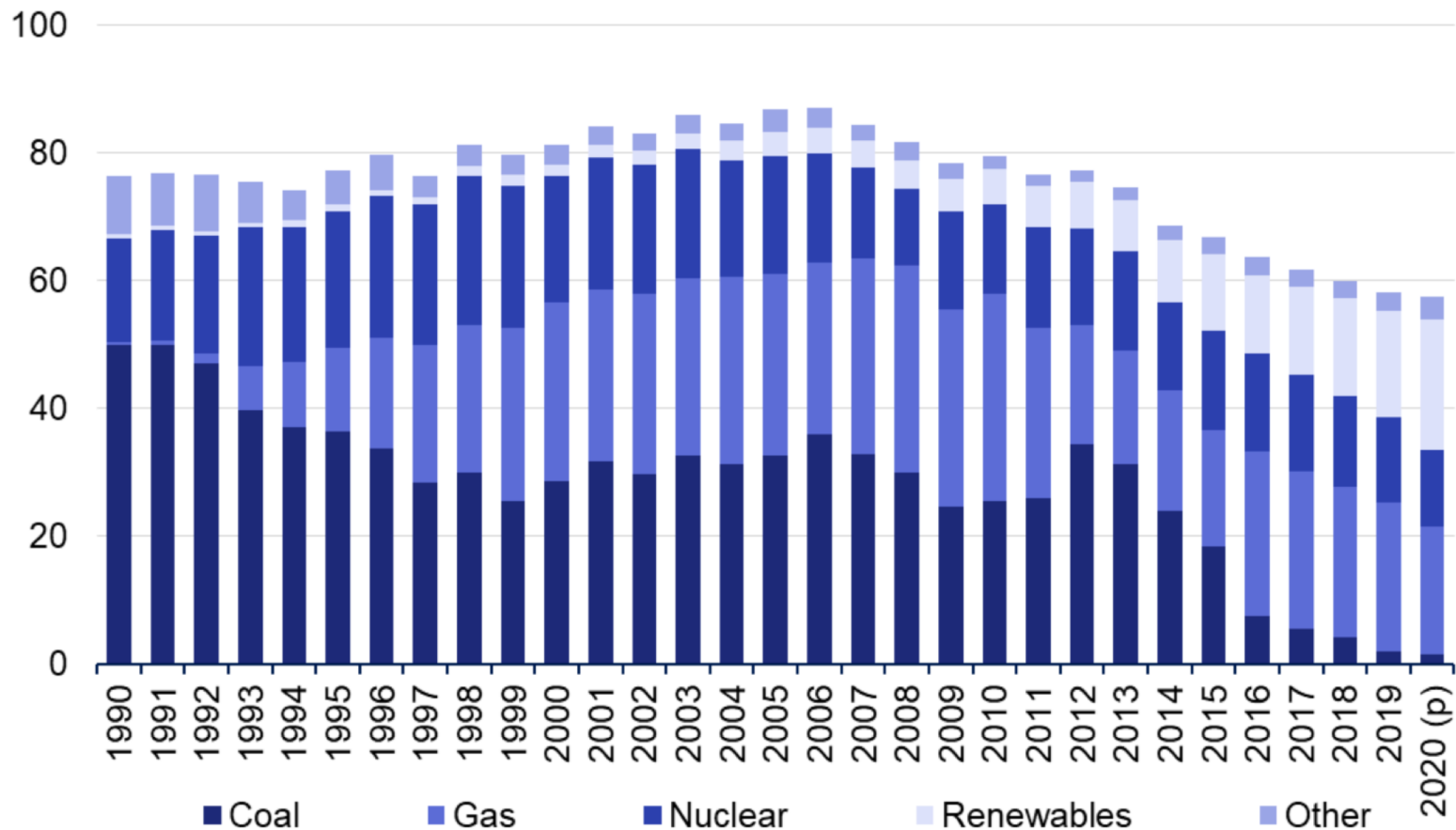


During operation phase

Weather conditions (geography)
Building and installation properties
Occupant behavior

±2.5 tons of operational CO₂ / year
100+ tons over lifetime

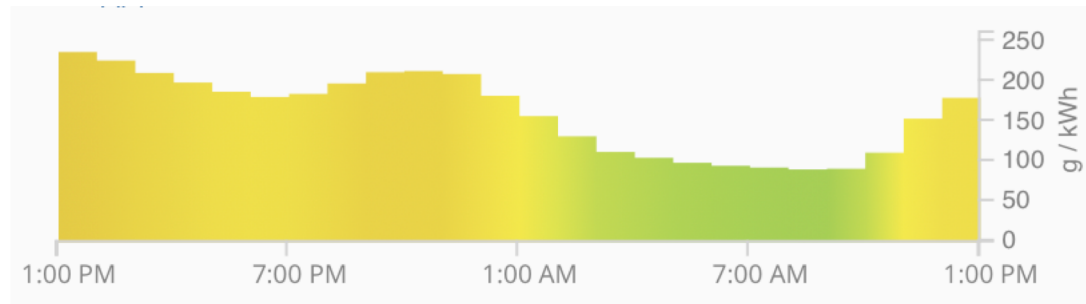
Electricity source mix



2020 UK greenhouse gas emissions, provisional figures, National Statistics

Carbon intensity of electricity mix

Situation on 27th October, 2021

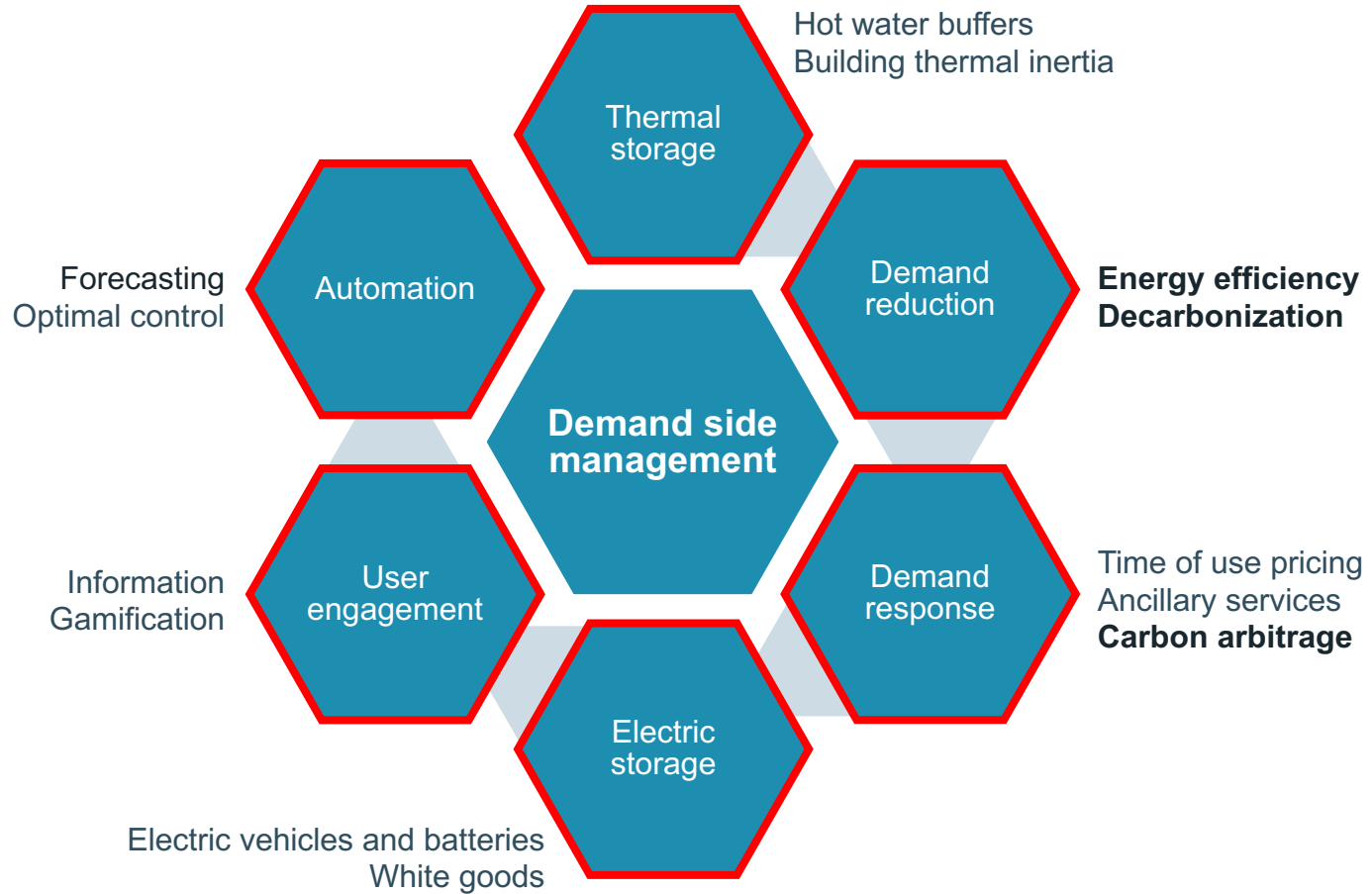


A lot more natural gas
Overall: ± 240 g CO₂

Mostly renewables via wind
Overall: ± 85 g CO₂

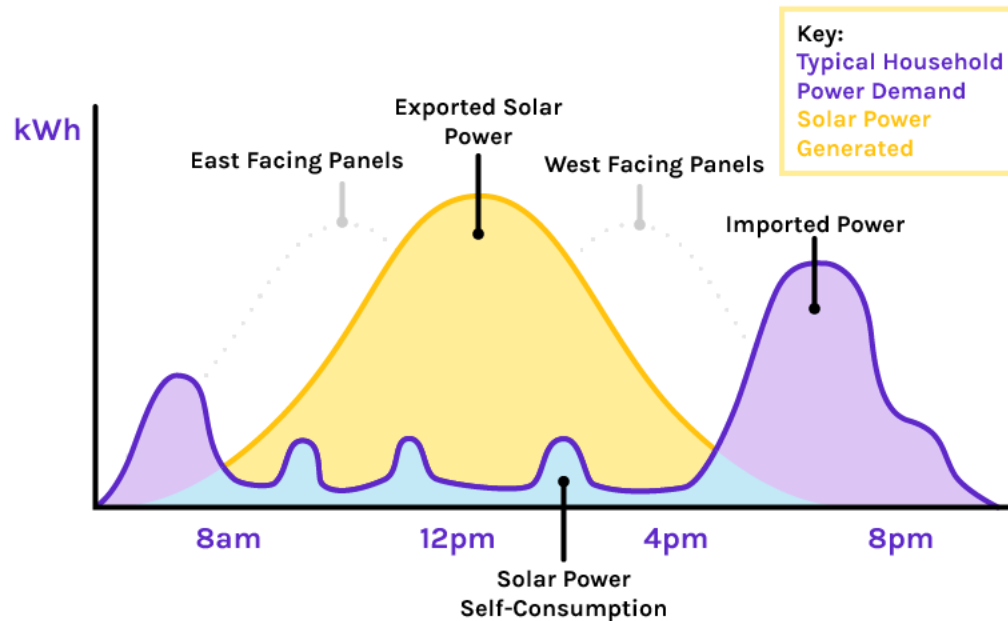
<https://app.electricitymap.org/zone/GB>

Demand side management

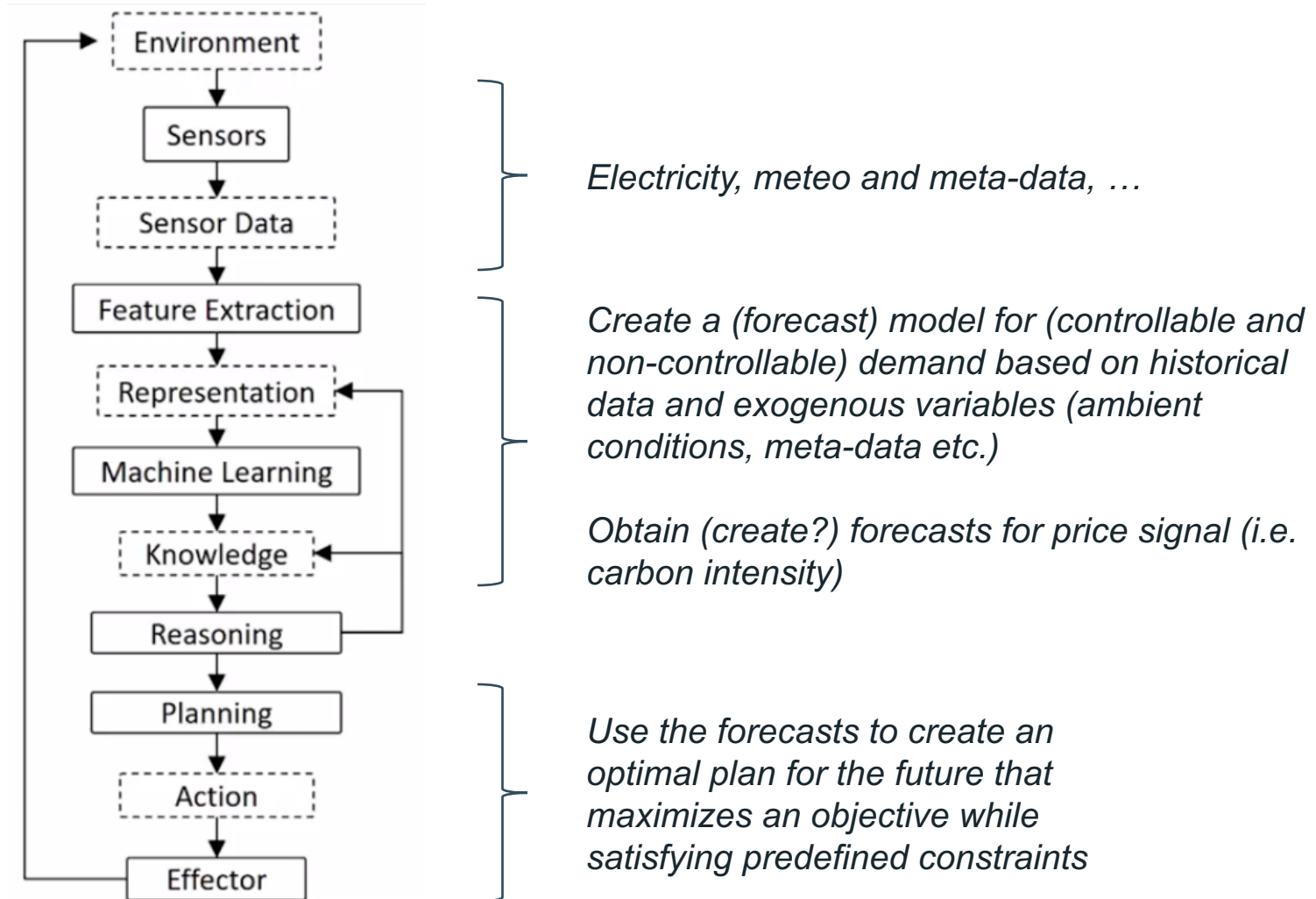


Self-consumption

SOLAR POWER SELF-CONSUMPTION



The data pipeline



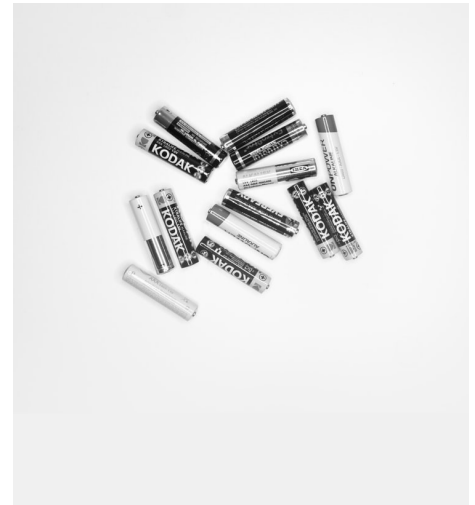
Elements of the optimization



Household electricity demand



Grid electricity supply
(carbon intensity)



Household electric storage (flexibility)

Constrained optimization

Minimize some costs

subject to some constraints

by changing variable, x

How to decarbonize

Objective

Minimize

$$\sum_{N_h} (p_{el} \cdot P_g)$$

Control variable



Constraints

subject to

$$\frac{dE_b}{dt} = P_b$$

$$E_b < E_{max}$$

$$-P_{max} \leq P_b \leq P_{max}$$

$$P_g = P_b + P_c$$

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- Random guessing / grid search
 - Pros: ???
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- Derivative-free optimization
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- Convex optimization
 - Pros: Accurate and fast
 - Cons: May be infeasible with complex models or require convexification

Other considerations

- Forecasts for electricity demand and carbon intensity?
- Actual control of the battery
- Dimensioning the battery
- Local generation with solar PV
- ...