

EPFL

■ **IPESÉ**
Industrial Process
and Energy Systems
Engineering

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de Lausanne**

ENV-421






Energy Technologies

Jonas SCHNDIRIG

Week 7

This week

Agenda

	Week 2 	Week 4 	Week 7 	Week 9 
Lectures	Energy System Fundamentals	Energy Conversion Technologies	Technologies' Impacts	Climate Impact on Energies
Applications & Exercises	The Swiss Energy System Evolution & Perspectives	Efficiencies & Classification	Conference <i>Is it all about renewable energies?</i> Closing the Balance & Defining Compromises	Powerplay Game
	Project: Addressing Contemporary Challenges to the Swiss system Energy-independent and carbon-neutral Switzerland 2050 			





A G E N D A

Carbon Cycle

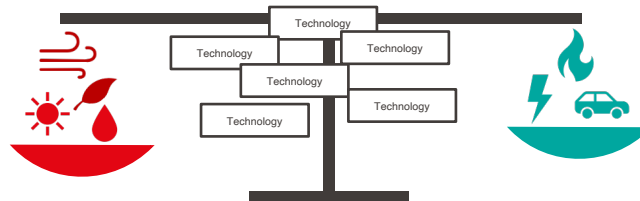
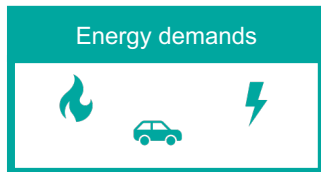
Energy System Modeling

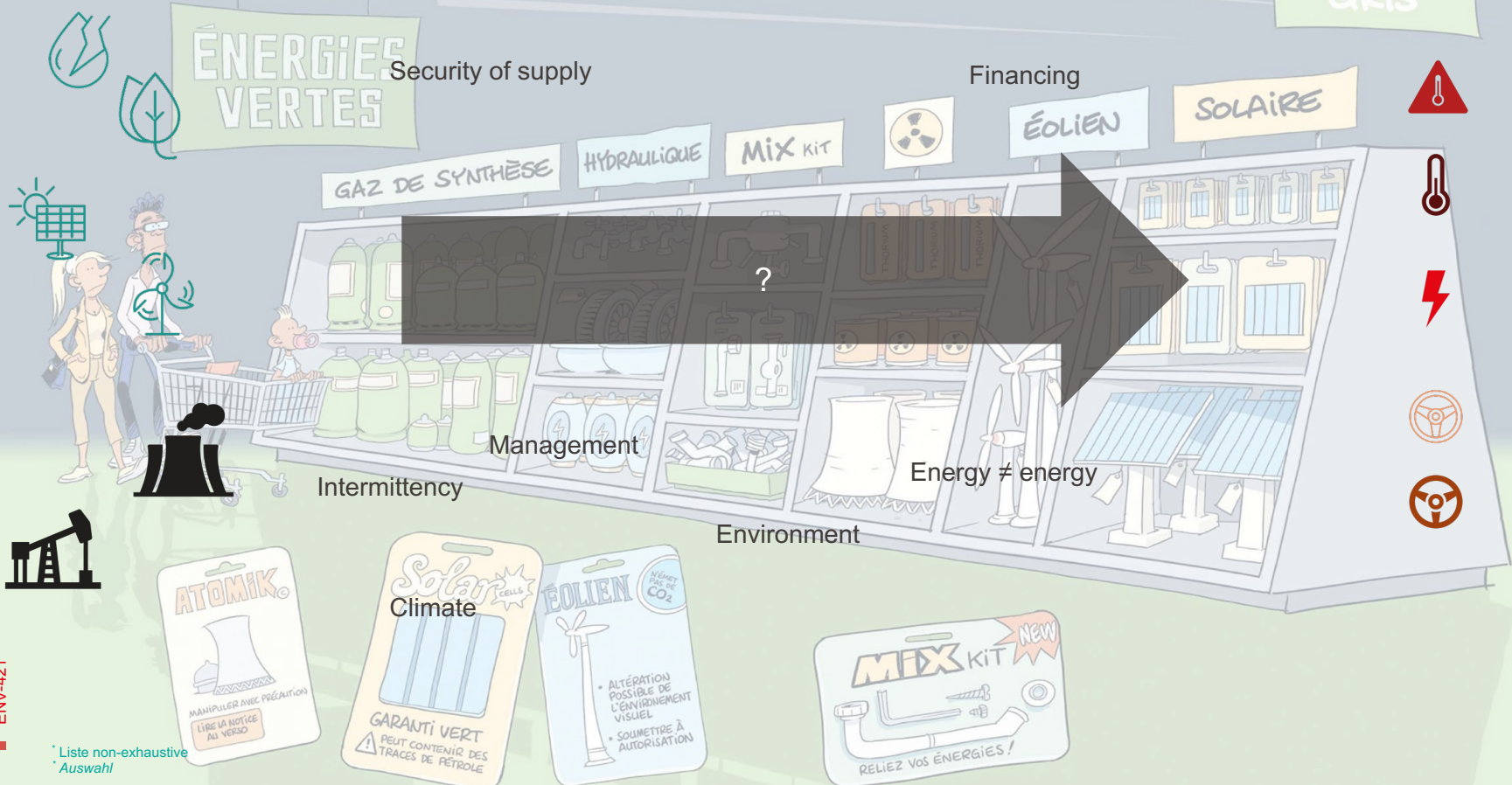
Technologies' Impacts

Optimization

Multi-Objective Optimization







Resources

Cost & emissions
Yearly availability

Technology example: *Combined Cycle gas turbine*

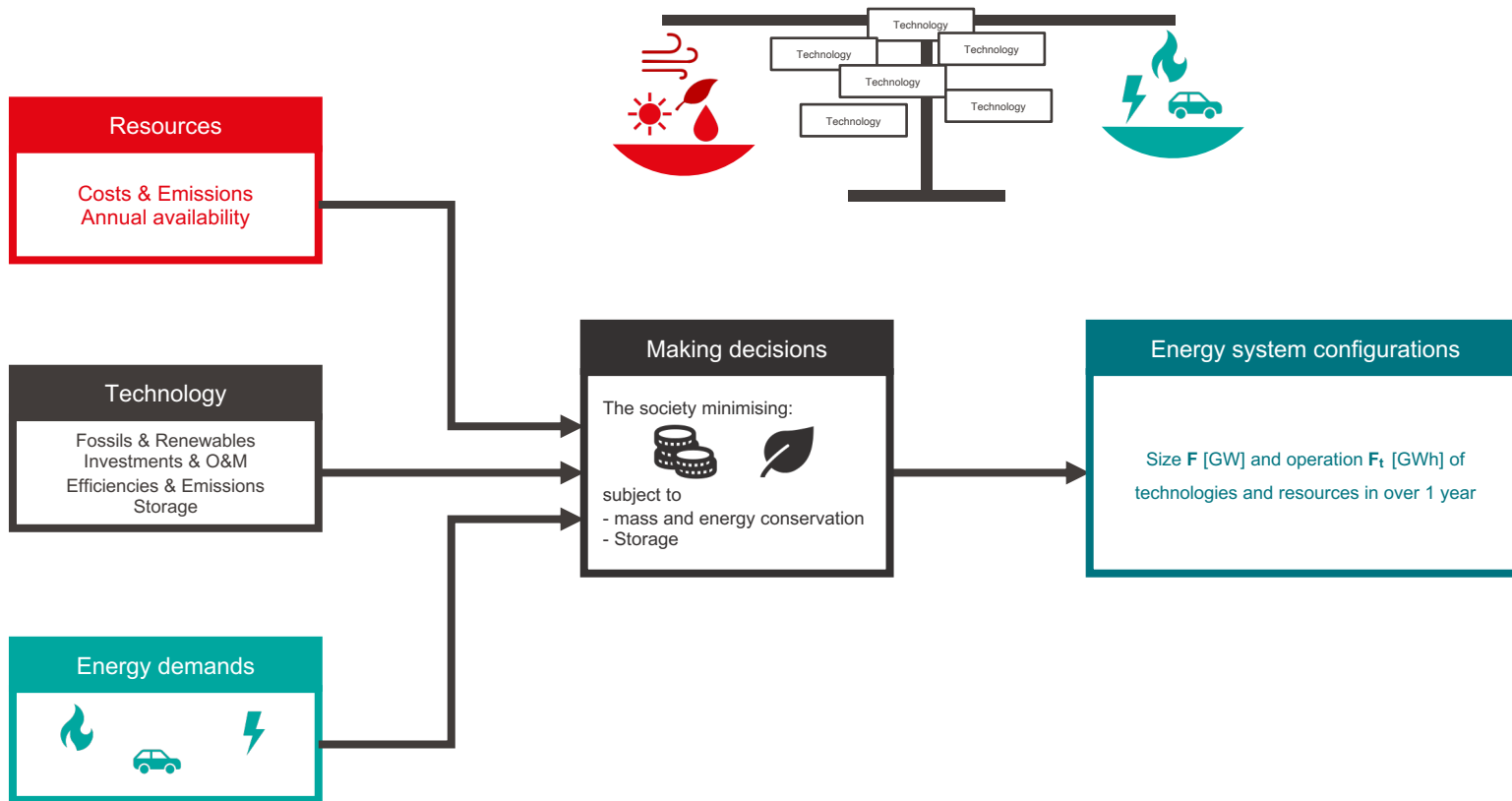
Technologies

Fossils and renewables
Investment and O&M cost
Efficiencies, emissions
Storage

Energy demand



Energy system modeling - EnergyScope



Energy system modeling

The life of a technology

Installation F [GW] and operation F_t [GWh] of technologies and resources in over 1 year



Investment
25 kCHF

Annual Service
1000 CHF/year
Fuel
1.82 CHF/l

Disposal (Transport)
100 CHF



Construction Cost

Maintenance Cost
Exploitation Cost

Demolition and Disposal Cost

Construction

Use

EoL



Construction Impact

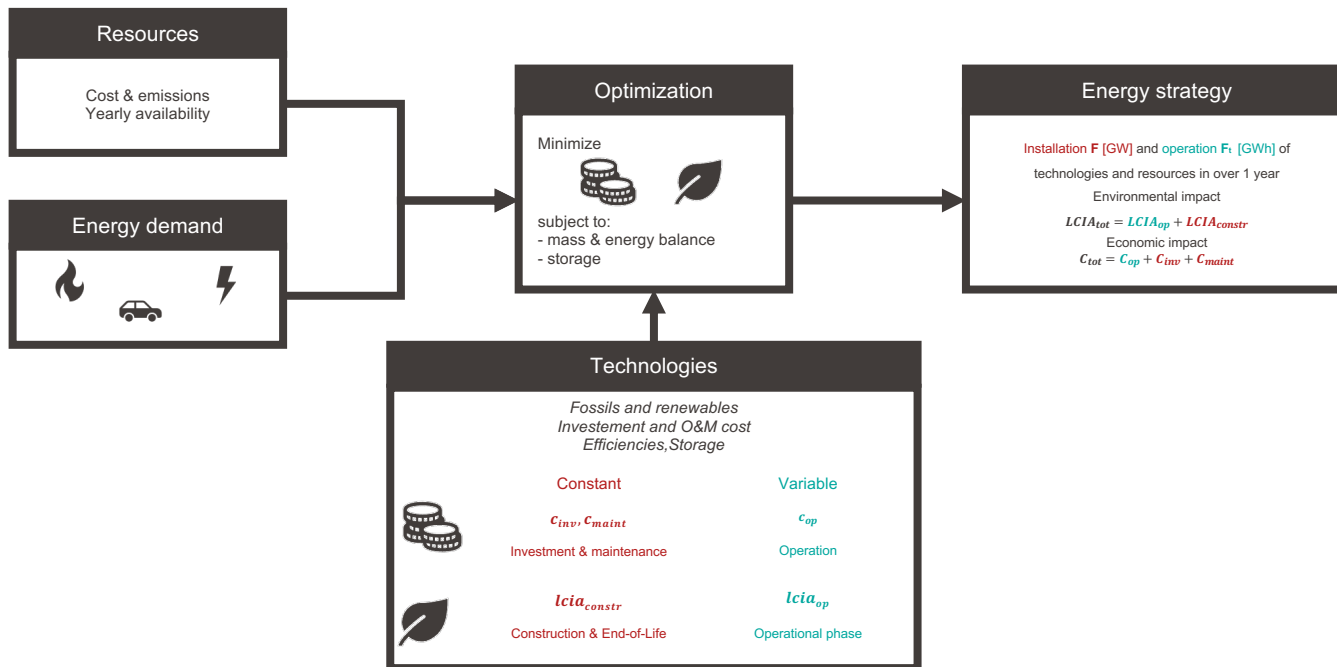
Exploitation Impact

Demolition and Disposal impact



Energy system modeling

Technologies' impacts within the energy system



Graphical representation of the methodology followed integrating LCIA indicators into ES. The green steps at the bottom of the figure illustrate the adaptation of the Life Cycle Inventory (LCI) to the ES technologies to be split into variable and constant impact to allow the optimization of economic (red) and environmental (green) variables.

Resources

Cost & emissions
Yearly availability

Technology example: *Combined Cycle gas turbine*

Technologies

Fossils and renewables
Investment and O&M cost
Efficiencies, emissions
Storage

Energy demand



Investment costs:	1339.67 [kCHF/MW]
Maintenance costs:	40.08 [kCHF/MW]
Construction emissions:	490.88 [t CO ₂ /MW]
Lifetime:	24 [y]
Annual operation factor:	85 [%]



Energy system modeling

Economic characterization



Investment

25 kCHF

Disposal (Transport)

100 CHF

Annual Service

1000 CHF/year

Fuel

1.82 CHF/l

Investment + EoL

$$C_{inv} \left[\frac{\text{CHF}}{\text{year}} \right] = c_{inv} \left[\frac{\text{CHF}}{\text{kW}} \right] \cdot F [\text{kW}] \cdot \tau \left[\frac{1}{\text{year}} \right]$$

- c_{inv} : specific investment cost
- F : installed size
- $\tau = \frac{i(1+i)^n}{(1+i)^n - 1}$ annualization factor
- i : interest rate
- n : lifetime

Maintenance

$$C_{maint} \left[\frac{\text{CHF}}{\text{year}} \right] = c_{maint} \left[\frac{\text{CHF}}{\text{kW} \cdot \text{year}} \right] \cdot F [\text{kW}]$$

- c_{maint} : specific maintenance cost
- F : installed size

Operation

$$C_{op} \left[\frac{\text{CHF}}{\text{year}} \right] = c_{op} \left[\frac{\text{CHF}}{\text{kWh}} \right] \cdot F_t [\text{kW}] \cdot t_{op} [h]$$

- c_{op} : specific operational cost
- F_t : technology use during period t
- t_{op} : period duration



Energy system modeling

Economic characterization



Investment

25 kCHF

Disposal (Transport)

100 CHF

Annual Service

1000 CHF/year

Fuel

1.82 CHF/l



Investment + EoL

$$C_{inv} [\text{CHF}/\text{year}] = c_{inv} \left[\frac{\text{CHF}}{\text{kW}} \right] \cdot F [\text{kW}] \cdot \tau \left[\frac{1}{\text{year}} \right]$$

- c_{inv} : specific investment cost
- F : installed size
- $\tau = \frac{i(1+i)^n}{(1+i)^n - 1}$ annualization factor
- i : interest rate
- n : lifetime

Maintenance

$$C_{maint} \left[\frac{\text{CHF}}{\text{year}} \right] = c_{maint} \left[\frac{\text{CHF}}{\text{kW} \cdot \text{year}} \right] \cdot F [\text{kW}]$$

- c_{maint} : specific maintenance cost
- F : installed size

Operation

$$C_{op} \left[\frac{\text{CHF}}{\text{year}} \right] = \sum_t c_{op} \left[\frac{\text{CHF}}{\text{kWh}} \right] \cdot F_t [\text{kWh}]$$

- c_{op} : specific operational cost
- F_t : technology use during period t

Total Cost

$$\sum_{RES} C_{op} [res] \quad C_{tot} = \sum_{TEC} (C_{inv}[tec] + C_{maint}[tec]) +$$

Calculate the annual cost of the car given in example, assuming an annual usage 20'000 km at a consumption of 5l/100km. The car is leased for 4 years at 0.9% .



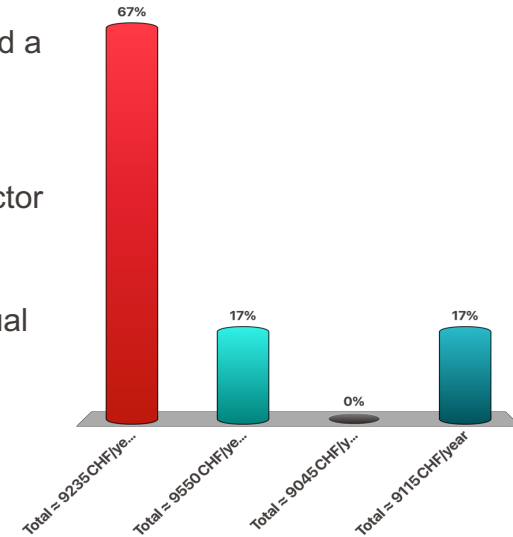
Annual Cost of a Fossil-Fuel Car

A gasoline car travels 20,000 km/year with a consumption of 5 L/100 km. The fuel price is 1.82 CHF/L. The car's purchase price is 25,000 CHF, financed over 4 years at 0.9% interest, with an annual service cost of 1,000 CHF and a one-time disposal fee of 100 CHF (annualized over 4 years).

- (a) Calculate the annual fuel cost.
 (b) Compute the annualized lease (capital) cost using the annualization factor

$$\tau = \frac{i(1+i)^n}{(1+i)^n - 1}, \text{ with } P = 25'000 \text{ CHF, } i = 0.9\%, n = 4.$$

- (c) Sum the fuel, lease, service, and disposal costs to obtain the total annual cost.



- ✓ A. Total ≈ 9235 CHF/year
 B. Total ≈ 9550 CHF/year
 C. Total ≈ 9045 CHF/year
 D. Total ≈ 9115 CHF/year



1. Step (a): Fuel Cost

- $Fuel\ consumption = 20'000 \times \frac{5}{100} = 1000L/year$
- $Fuel\ cost = 1000L \times 1.82CHF/L = 1820CHF/year$

2. Step (b): Lease Payment

- Calculate using the annuity formula:
 - $(1+i)^n = (1.009)^4 \approx 1.0366$
 - $Payment \approx 25'000 \times \frac{(0.009 \times 1.0366)}{(1.0366-1)} \approx 25'000 \times \frac{0.00933}{0.0366} \approx 25'000 \times 0.255 \approx 6375CHF/year$

3. Step (c): Service and Disposal

- Service = 1000 CHF/year; Disposal fee = 100 CHF over 4 years = 25 CHF/year.

4. Total Annual Cost:

- $1820 + 6375 + 1000 + 25 \approx 9220CHF/year$ (≈ 9235 CHF when rounded)}

Correct answer: Option A



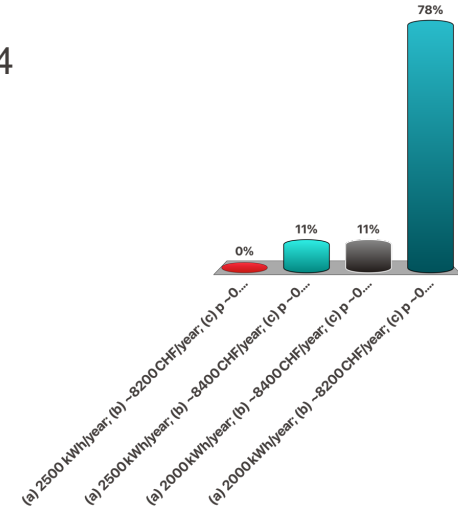
Car Switching to BEV: Break-even Electricity Price

A BEV is considered with the following parameters:

- **Usage:** 20,000 km/year; consumption: 10 kWh/100 km
- **Investment:** 35,000 CHF financed over 4 years at 0.9%
- **Annual Service:** 500 CHF; **Disposal Fee:** 250 CHF (annualized over 4 years)
- Let p be the electricity price (in CHF/kWh). Determine p such that the BEV's total annual cost equals the fossil car's 9235 CHF/year.

- (a) Compute the BEV's annual energy consumption.
 (b) Calculate the annualized lease cost for 30,000 CHF.
 (c) Write and solve the equation for p given that total annual cost equals fixed costs plus electricity cost.

- A. (a) 2500 kWh/year; (b) ~8200 CHF/year; (c) $p \sim 0.20$ CHF/kWh
 B. (a) 2500 kWh/year; (b) ~8400 CHF/year; (c) $p \sim 0.00$ CHF/kWh
 C. (a) 2000 kWh/year; (b) ~8400 CHF/year; (c) $p \sim 0.80$ CHF/kWh
 ✓ D. (a) 2000 kWh/year; (b) ~8200 CHF/year; (c) $p \sim 0.50$ CHF/kWh



Car Switching to BEV: Break-even Electricity Price

1. Step (a): Annual Energy Consumption

- $Energy = 20'000 \times \frac{10}{100} = 2000 \text{ kWh/year}$

2. Step (b): Lease Payment for 30,000 CHF

- Using the same annuity method, note that increasing the principal from 25,000 CHF to 30,000 CHF scales the lease by 1.2:

- $6375 \frac{\text{CHF}}{\text{year}} \times 1.2 \approx 7650 \frac{\text{CHF}}{\text{year}}$

3. Step (c): Fixed Costs

- Fixed costs = Lease + Service + Disposal:

- $7650 + 500 + \frac{250}{4} = 7650 + 500 + 62.5 = 8212.5 \text{ CHF/year}$

- Total cost equation for BEV:

- $8212.5 + 2000p = 9235$

- Solve for p:

- $2000p = 9235 - 8212.5 = 1022.5$

- $p - \frac{1022.5}{2000} \approx 0.511 \frac{\text{CHF}}{\text{kWh}}$

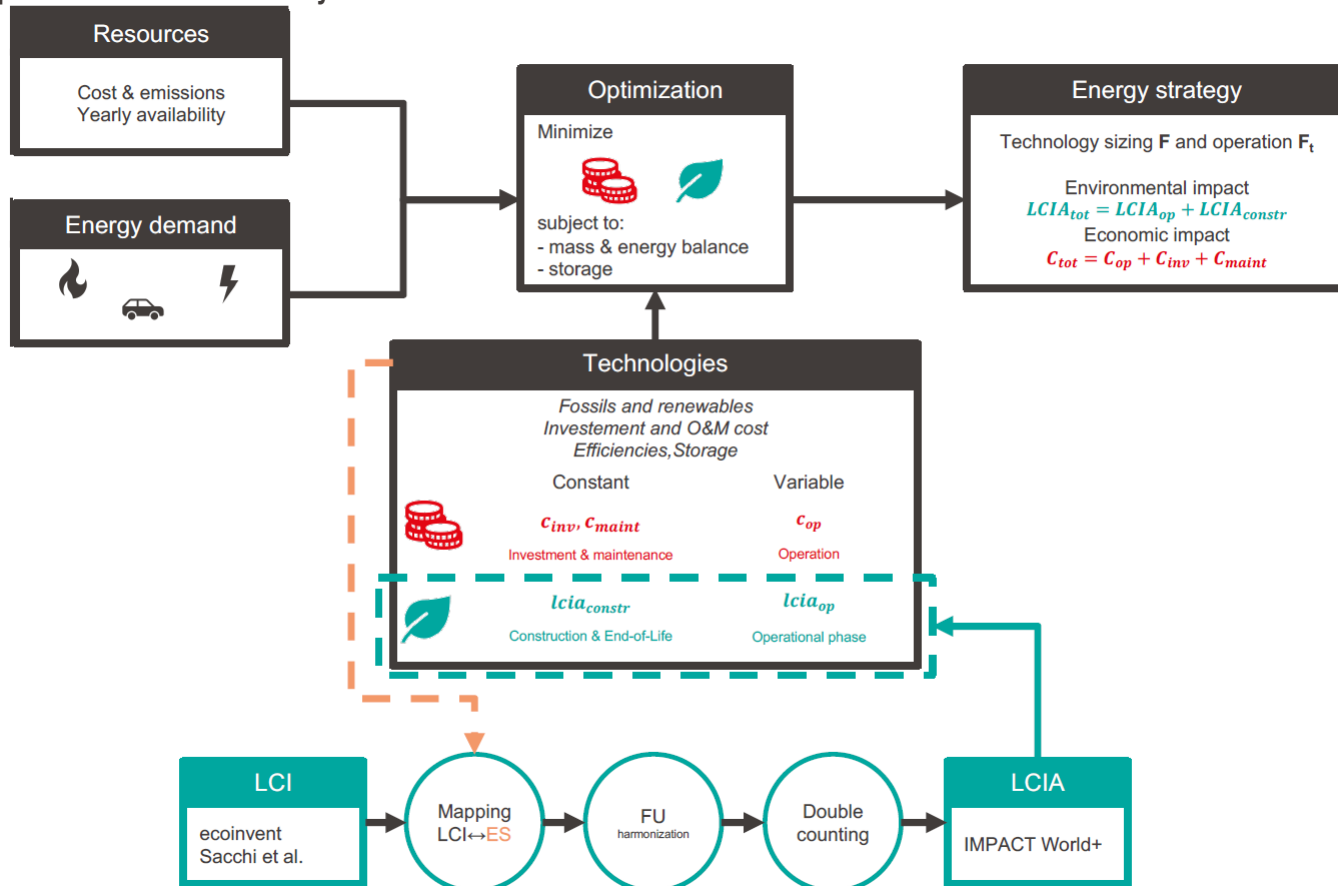
- Thus, $p \sim 0.5 \text{ CHF/kWh}$, indicating that the BEV will be less expensive, as long as the electricity price is $\leq 0.51 \text{ CHF/kWh}$

Correct answer: Option D



Integration of life cycle inventory

Compromises linked to sustainability



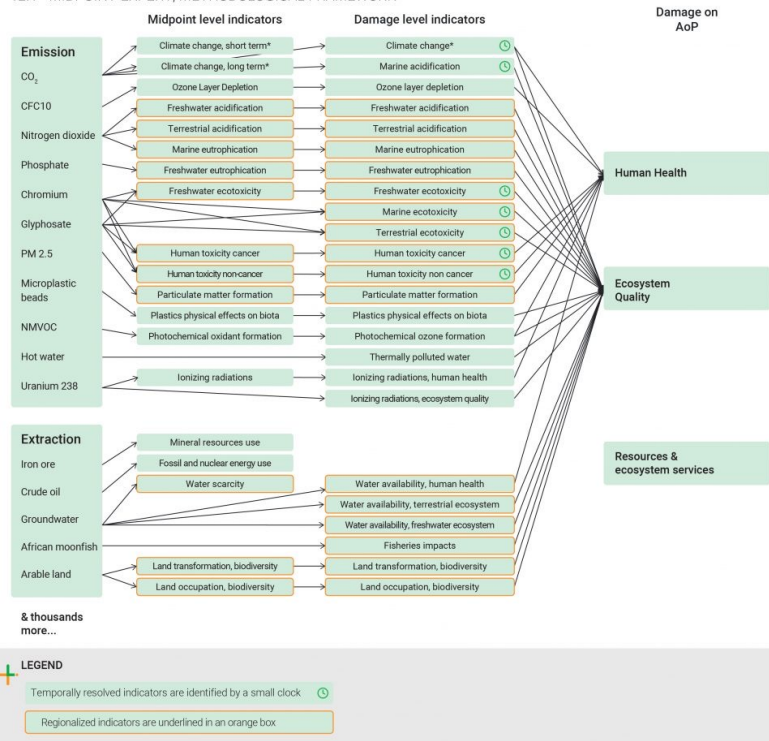
Environmental indicator



Impact profile	AoP	Acronym	Indicator	Unit
		CF	Carbon footprint	[kg · CO ₂ ^{eq} (short)]
		FNEU	Fossil and nuclear energy use	[MJ deprived]
		REQD	Remaining Ecosystem quality damage	[DALY]
		RHHD	Remaining Human health damage	[PDF · m ² · yr]
		WSF	Water scarcity footprint	[m ³ world – eq]
Impact categories	Human Health	CCHHL	Climate change, human health, long term	[DALY]
		CCHHS	Climate change, human health, short term	[DALY]
		HTXCL	Human toxicity cancer, long term	[DALY]
		HTXCS	Human toxicity cancer, short term	[DALY]
		HTXNCL	Human toxicity non-cancer, long term	[DALY]
		HTXNCS	Human toxicity non-cancer, short term	[DALY]
		IRHH	Ionizing radiation, human health	[DALY]
		OLD	Ozone layer depletion	[DALY]
		PMF	Particulate matter formation	[DALY]
		PCOX	Photochemical oxidant formation	[DALY]
	Ecosystem Quality	TTHH	Total human health	[DALY]
		WAVHH	Water availability, human health	[DALY]
		CCEQL	Climate change, ecosystem quality, long term	[PDF · m ² · yr]
		CCEQS	Climate change, ecosystem quality, short term	[PDF · m ² · yr]
		FWA	Freshwater acidification	[PDF · m ² · yr]
		FWEXL	Freshwater ecotoxicity, long term	[PDF · m ² · yr]
		FWEXS	Freshwater ecotoxicity, short term	[PDF · m ² · yr]
		FWEU	Freshwater eutrophication	[PDF · m ² · yr]
		IREQ	Ionizing radiation, ecosystem quality	[PDF · m ² · yr]
		LOBDV	Land occupation, biodiversity	[PDF · m ² · yr]



v2.1 - MIDPOINT-EXPERT, METHODOLOGICAL FRAMEWORK



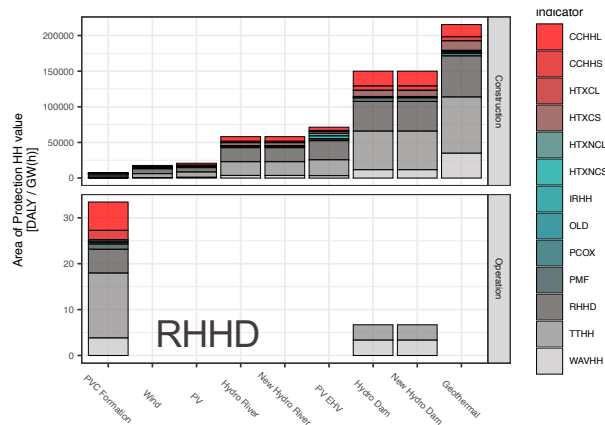
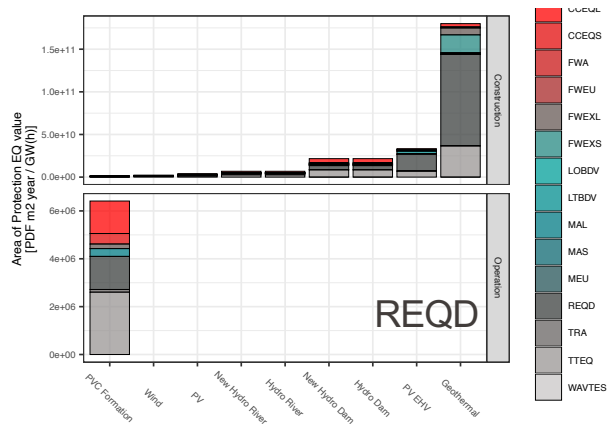
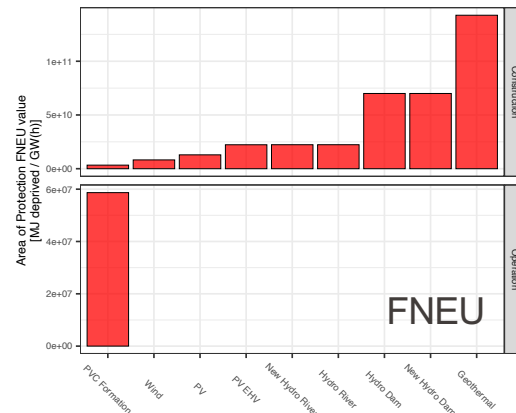
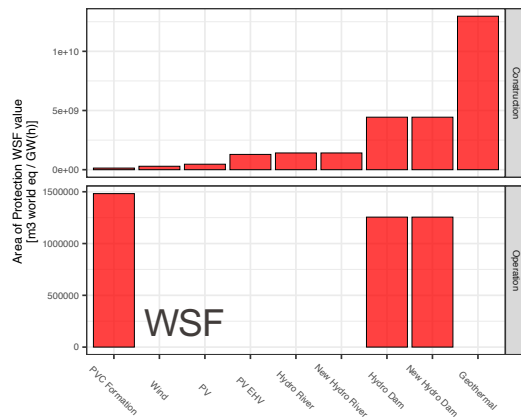
DALY:

- Disability-adjusted life years
- One DALY represents the loss of the equivalent of one year of full health. DALYs for a disease or health condition are the sum of the years of life lost due to premature mortality and the years lived with a disability due to prevalent cases of the disease or health condition in a population.

PDF m² year:

- Potentially Disappeared Fraction
- The unit for overall biodiversity impact using ecosystem quality and species density to describe biodiversity loss







A G E N D A

Carbon Cycle

Energy System Modeling

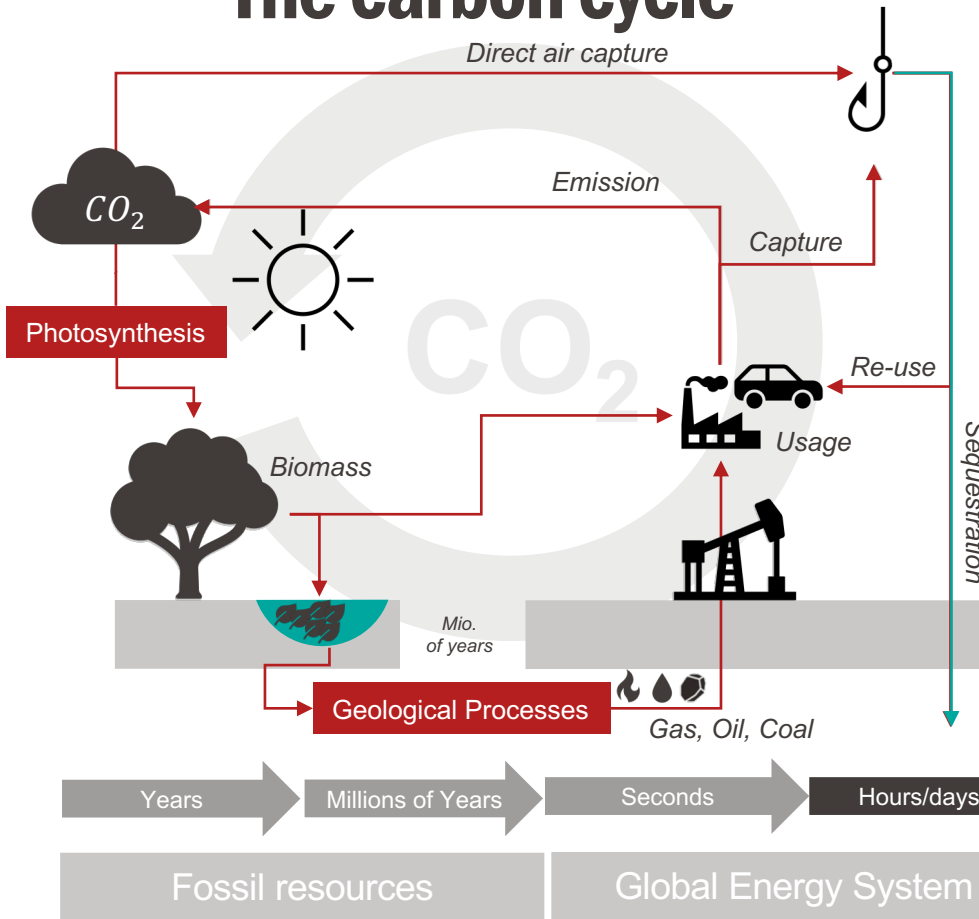
Technologies' Impacts

Optimization

Multi-Objective Optimization



The carbon cycle

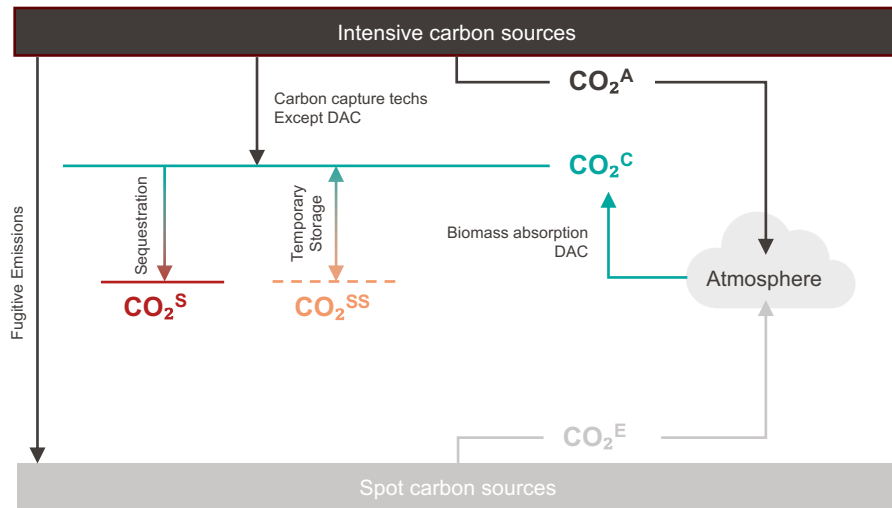


- **Definition:**
The carbon cycle in energy systems represents the movement of carbon atoms through sources, conversions, and sinks, spanning biogenic and non-biogenic origins.
- **Key Components:**
 - Carbon Sources:
 - **Biogenic:** Biomass (wood, wet biomass, plants).
 - **Non-biogenic:** Fossil fuels (natural gas, petroleum products, cement manufacturing emissions).
 - Carbon Sinks:
 - **Natural:** Oceans, Forests, other vegetation etc.
 - **Artificial:** Carbon capture, utilization, and storage (CCUS).
 - Intermediate Carbon Flows:
 - CO_2 emissions from energy conversion and utilization technologies.
 - CO_2 utilization for fuel and chemical synthesis.



The Carbon Cycle

Modeling the Carbon Cycle in Energy Systems



- Models carbon flows as interconnected “layers” balancing inputs and outputs.
 - CO₂ Emission Layer (**CO₂^A**): From **concentrated** sources (e.g., power plants).
 - CO₂ Emission Layer (**CO₂^E**): From **decentralized** sources (e.g., transport).
 - CO₂ Capture Layer (**CO₂^C**): **Captured** emissions stored or utilized.
 - CO₂ Storage Layer (**CO₂^S**): **Sequestered** (long-term)
 - CO₂ Storage Layer (**CO₂^{SS}**): **Storage** (short-term)
- Mathematical Representation:
 - Carbon Content:

$$\text{Carbon Content}_r = \frac{m(c)}{m(r) \cdot LHV_r}$$

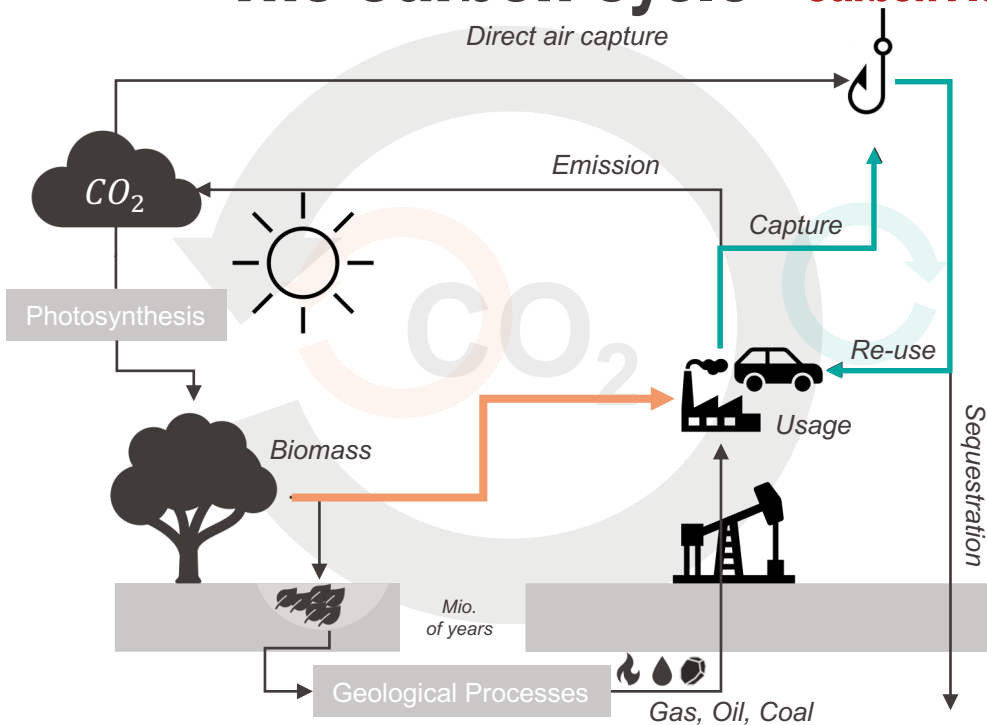
- Emission Balances:

$$\text{Emission}(t) = \sum_{j \in E, c \in C} F_t(j) \cdot \eta(j, c)$$

- $F_t(j)$: Flow from technology j in period t .
- $\eta(j, c)$: Emission factor for technology j in carbon layer c .



The Carbon Cycle - Carbon Flow and Conversion



Carbon Conversion Technologies:

Biomass Utilization:

- **Gasification:** Converts biomass to syngas (CO , H_2).
- **Fischer-Tropsch Synthesis:** Converts syngas to liquid fuels.
- **Anaerobic Digestion:** Produces biogas (CH_4 , CO_2).

Carbon Capture and Utilization (CCU):

- **Methanation:** Produces synthetic CH_4 .
- **CO₂-to-Fuels:** Diesel, jet fuel, and methanol synthesis.

Circular Carbon Flows:

- Carbon emitted by processes is captured and reused or stored.
- Example: CO_2 from biomass combustion used in synthetic fuel production.

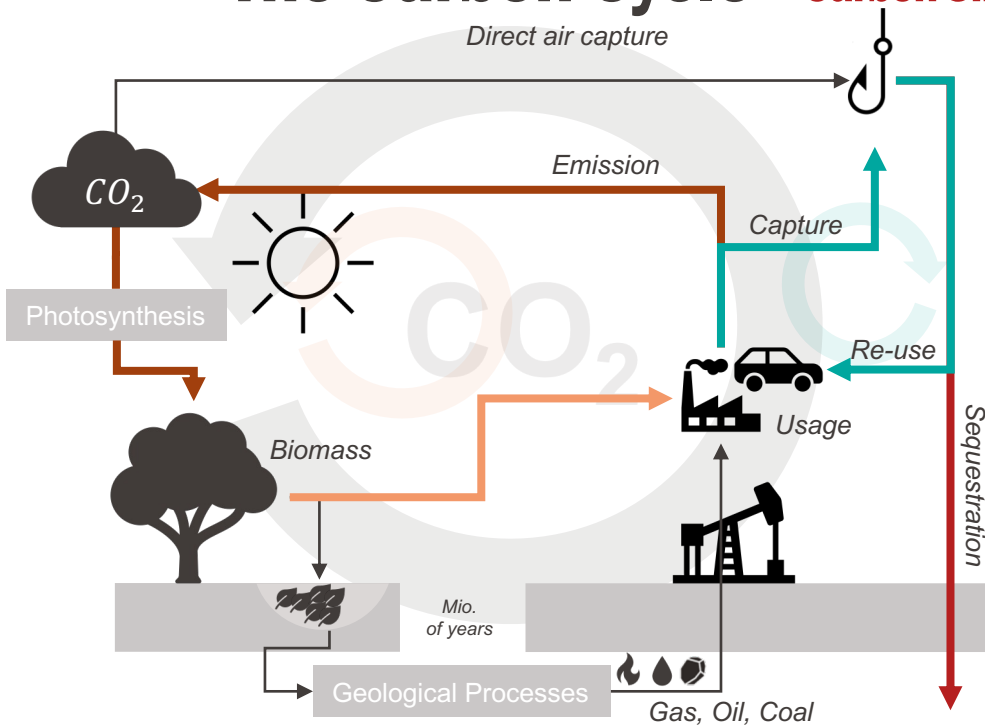
Fossil resources

Global Energy System

CO₂ & Climate Change

CCS

The Carbon Cycle - Carbon Sinks and Utilization



- Natural Carbon Sinks:
 - Reforestation and afforestation.
 - Biomass acting as a temporary carbon storage.
- Artificial Carbon Sinks:
 - CCS: Long-term storage in geological formations.
 - CCU:
 - Produces fuels (methane, diesel) and chemicals.
 - Reduces dependency on fossil fuels but does not eliminate emissions.
- Role of Negative Emissions:
 - Bioenergy with Carbon Capture and Sequestration, Combines biomass combustion with CCS for **net-negative emissions**.



Fossil resources

Global Energy System

CO₂ & Climate Change

CCS

The Carbon Cycle

Decarbonizing Energy Systems

■ Pathways:

- Increase renewable energy use and reduce fossil fuel dependency.
- Maximize carbon recycling through CCU technologies.
- Implement large-scale CCS for unavoidable emissions.

■ Challenges:

- Balancing economic feasibility and technological limitations.
- Managing the complexity of interconnected carbon flows.

■ Take-home:

- The integration of carbon cycle modeling is essential for planning decarbonized energy systems.
- Optimizing carbon sources, sinks, and flows ensures minimal environmental impact while meeting energy demands.

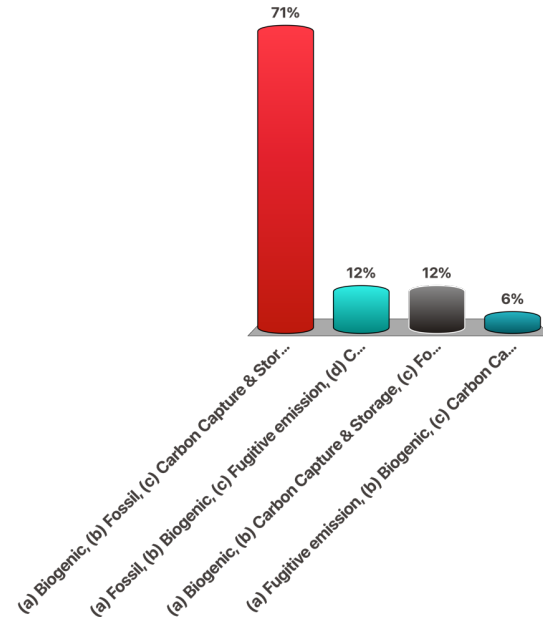


For each scenario below, select the combination that correctly classifies the carbon flow in the context of the carbon cycle:

Scenarios:

- (a) CO₂ emitted from burning wood in a biomass power plant.
- (b) CO₂ emitted from a natural gas power plant.
- (c) CO₂ captured from an industrial facility and injected underground.
- (d) Methane leaking from a natural gas pipeline.

- ✓ A. (a) Biogenic, (b) Fossil, (c) Carbon Capture & Storage, (d) Fugitive emission
- B. (a) Fossil, (b) Biogenic, (c) Fugitive emission, (d) Carbon Capture & Storage
- C. (a) Biogenic, (b) Carbon Capture & Storage, (c) Fossil, (d) Fugitive emission
- D. (a) Fugitive emission, (b) Biogenic, (c) Carbon Capture & Storage, (d) Fossil



- (a) Wood is a renewable, biogenic resource; thus, CO₂ from burning wood is **biogenic**.
- (b) Natural gas is of geological origin; its combustion releases **fossil** carbon.
- (c) Captured CO₂ that is injected underground is classified as **Carbon Capture & Storage (CCS)**.
- (d) Methane leaking from pipelines is an unintentional, uncontrolled release—i.e., a **fugitive emission**.

Correct answer: Option A (again 😊)





A G E N D A

Energy System Modeling

Carbon Cycle

The EnergyScope Framework

Optimization

Multi-Objective Optimization



$$\min f^{obj}(F, F_t) \quad \text{s.t.}:$$

Mass and Energy Balances
 Resource Constraints
 Technology Constraints
 Demand Constraints
 Policy Constraints

- Objective Function f^{obj} :
 - Minimize impacts subject to constraints:
 - Energy balance
 - Resource availability
 - Technological capacities
- Optimization Outputs:
 - Technology sizing F and operation F_t
 - Energy dispatch across scenarios (e.g., future years, renewable penetration).



min
Cost



min
Remaining Ecosystem
Quality Damage



min
Remaining Human Health
Damage



min
Carbon Footprint



min
Fossil & Nuclear Energy Use



min
Water Scarcity Footprint

■ **Objective Function** $f^{obj} = f_{size}^{obj} + f_{operation}^{obj}$

• Minimize impacts:

- Cost: $f_{cost}^{obj} = C_{tot} = C_{inv}^{tot} + C_{maint}^{tot} + C_{op}^{tot}$

$$= \sum_{TEC} (C_{inv}[tec] + C_{maint}[tec]) + \sum_{RES} C_{op}[res]$$

$$= \sum_{TEC} (c_{inv}[tec] \cdot F[tec] \cdot \tau[tec] + c_{maint}[tec] \cdot F[tec])$$

$$+ \sum_{RES} \sum_T (c_{op}[res] \cdot F_t[res] \cdot t_{op}[t])$$
- Environmental: $f_{LCIA}^{obj} = LCIA = LCIA_{static}^{tot} + LCIA_{variable}^{tot}$

$$= \sum_{TEC} (LCIA_{constr}[tec] + LCIA_{op}[tec]) + \sum_{RES} LCIA_{op}[res]$$

$$= \sum_{TEC} (lcia_{constr}[tec] \cdot F[tec] + \sum_T (lcia_{op}[tec] \cdot F_t[tec] \cdot t_{op}[t]))$$

$$+ \sum_{RES} \sum_T (lcia_{op}[res] \cdot F_t[res] \cdot t_{op}[t])$$

$$\min f^{obj}(F, F_t) \quad \text{s.t.}:$$

Mass and Energy Balances
Resource Constraints
Technology Constraints
Demand Constraints
Policy Constraints

■ **Optimization Outputs:**

- Technology sizing F and operation F_t



min
Cost



min
Remaining Ecosystem
Quality Damage



min
Remaining Human Health
Damage



min
Carbon Footprint



min
Fossil & Nuclear Energy Use



min
Water Scarcity Footprint

$$\min f^{obj}(F, F_t) \quad \text{s.t.}:$$

Mass and Energy Balances

Resource Constraints

Technology Constraints

Demand Constraints

Policy Constraints

- Energy & Mass balance in every period t :

End-Uses: $EU(l, t)$

$$\begin{aligned} &= \sum_{tec} (F_t(tec, t) \cdot \eta(tec, l)) \\ &+ \sum_{res} (F_t^+(res, t) - F_t^-(res, t)) \\ &+ \sum_{sto} (F_t^+(sto, l, t) - F_t^-(sto, l, t)) \\ &- F_t^{Loss}(l, t) \end{aligned}$$

$$\min f^{obj}(F, F_t) \quad \text{s.t.}:$$

Mass and Energy Balances

Resource Constraints

Technology Constraints

Demand Constraints

Policy Constraints

■ Resource Constraints

$$\sum_t \left(F_t(res, t) \cdot t_{op}(t) \right) \leq avail(res)$$

■ Technology Constraints

- Sizing:

$$f_{min}(tec) \leq F(tec) \leq f_{max}(tec)$$

- Use:

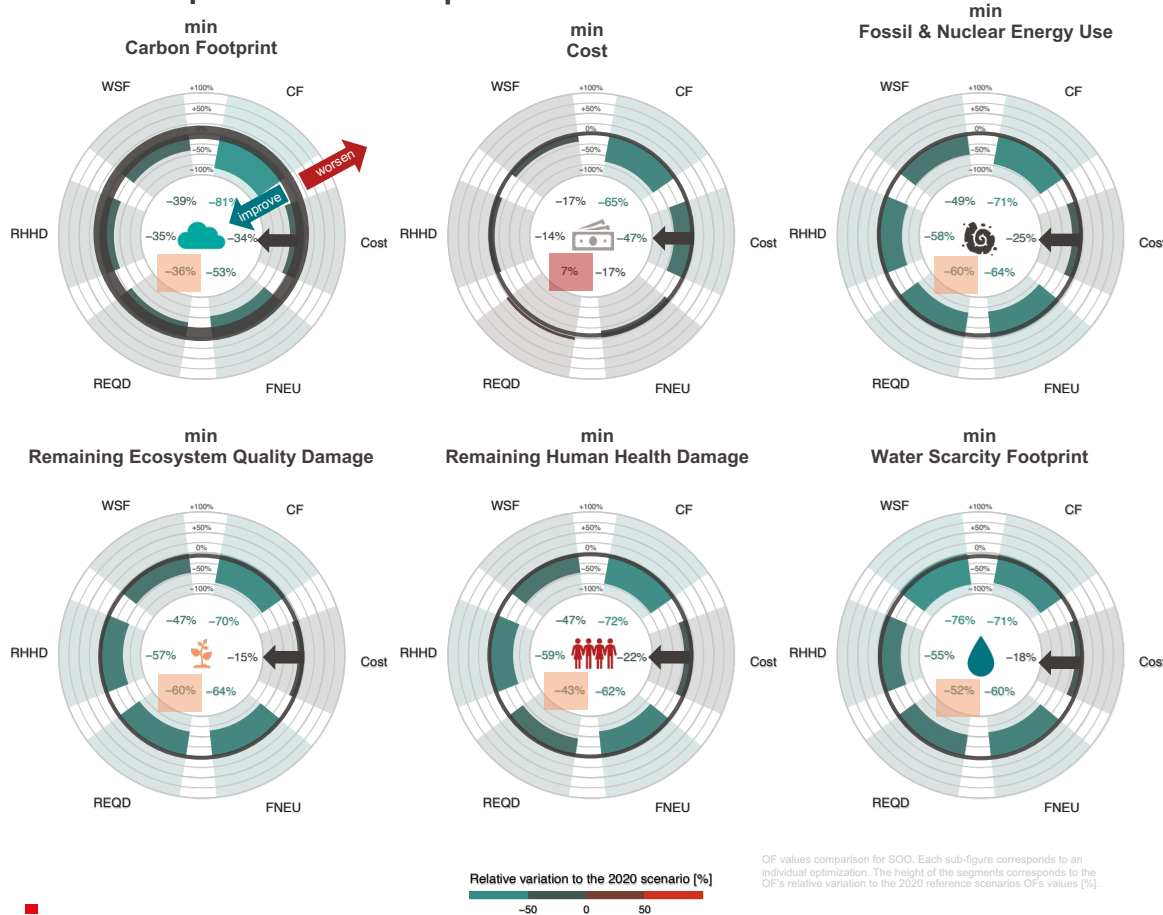
$$F_t(tec, t) \leq F(tec) \cdot c_{p,t}(t)$$

- Annual Capacity:

$$\sum_t \left(F_t(tec, t) \cdot t_{op}(t) \right) \leq F(tec) \cdot c_p \sum_t t_{op}(t)$$

Go green: minimizing Impacts

Impact assessment as compared to 2020



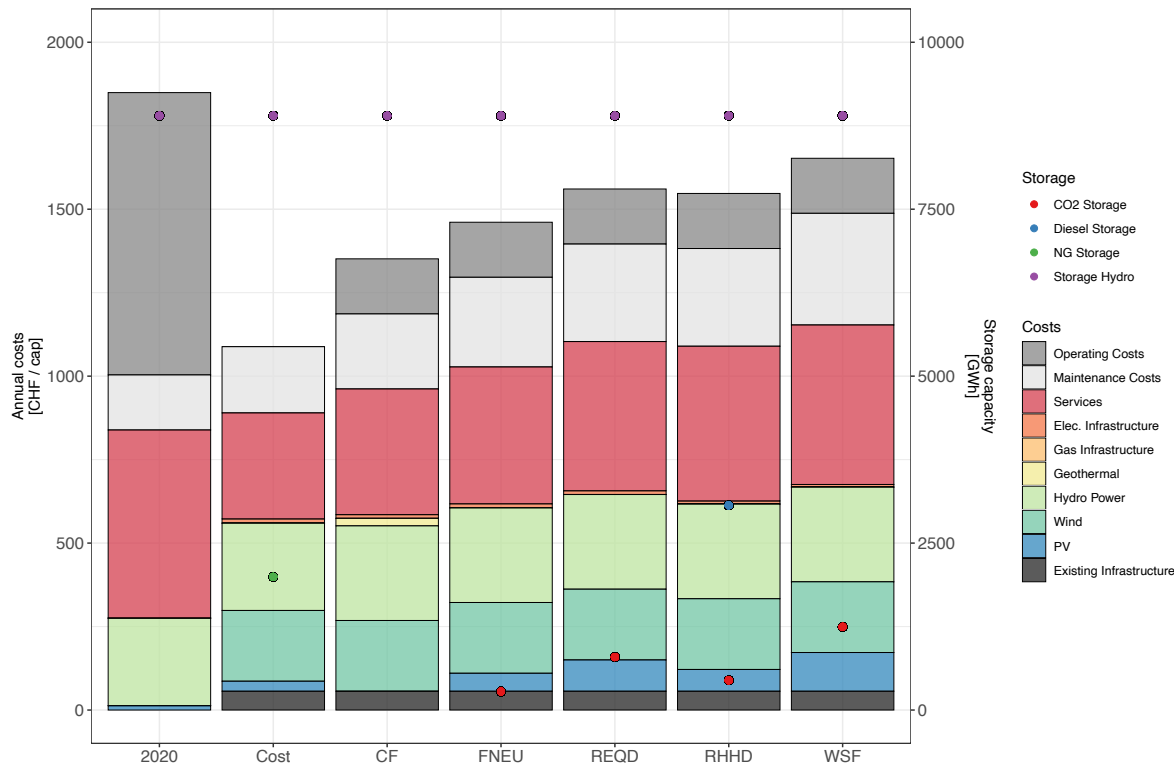
- Optimizing environmental indicators individually
- Tracking effect of other indicators in comparison to **2020**
- Minimizing any environmental indicator leads to **cost-reduction**
- Single-Objective Optimization leads to burden-shifting

→ Impact Trade-offs?



Single-Objective Optimization

Energy System Configurations



- Current System (2020 Reference Case):
 - High operational costs due to significant reliance on imported fossil fuels.
- Hypothetical Energy-Independent Scenarios:
 - Fully self-sustained systems utilizing only local energy resources.
 - Diverse optimization objectives lead to variations in system configurations.
- Renewable Energy Deployment:
 - Maximum utilization of wind and hydropower resources.
 - Photovoltaic (PV) capacity ranges from 7 GW to 20 GW, except in the CF minimization scenario, which favors geothermal power (3.5 GW).
- Energy Storage Capacity:
 - Significant installation to balance renewable intermittency.
 - Capacity details highlighted on the secondary axis of the figure.
- Biomass Utilization:
 - Maximized in most scenarios except for cost minimization.
 - Highlights local resource reliance.

Overall cost composition of energy systems for single-objective optimizations. The secondary axis highlights installed storage capacity. The 2020 scenario represents the current Swiss energy system, and the other six represent hypothetical scenarios for an energy-independent Switzerland in 2020 with single objective optimization.





A G E N D A

Energy System Modeling

Carbon Cycle

Technologies' Impacts

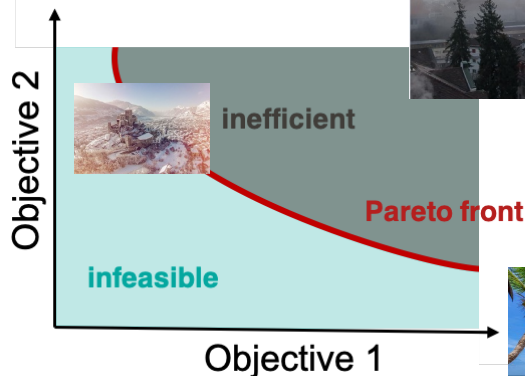
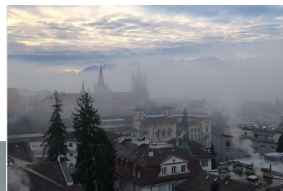
Optimization

Multi-Objective Optimization



Multi-Objective Optimization

Pareto-Front – 2 Dimensions



Choosing your holiday destination:

- Objective 1: Sunny Beach Paradise
- Objective 2: Alpine Mountain Retreat
- Infeasible: "Zermatt sur mer" (?)
- Inefficient: Foggy Lausanne
- Compromise: Sion: Sunny, alpine, beautiful, ...



The Pareto front represents the set of all non-dominated solutions in a multi-objective optimization problem, where no objective can be improved without worsening at least one other objective.

Methods to derive the Pareto-Front

- **Epsilon-Method**
Optimize one objective while treating the other objectives as constraints bounded by epsilon values.

$$\min f_1^{obj}(F, F_t)$$

$$\text{s.t. } f_2^{obj} < \epsilon$$

- **Weighted Sum Method**
Combine multiple objectives into a single objective using weighted coefficients.

$$\min f_1^{obj}(F, F_t) \cdot \omega_1 + f_2^{obj}(F, F_t) \cdot \omega_2$$

$$\text{s.t. } \omega_1 + \omega_2 = 1, \omega_{1,2} \geq 0$$

! Issues with different orders of magnitude of f_i^{obj}

- **Weighted Epsilon-Method**
Define $f_2^{min,max}$ and parametrize using weighted epsilon values

$$\min f_1^{obj}(F, F_t)$$

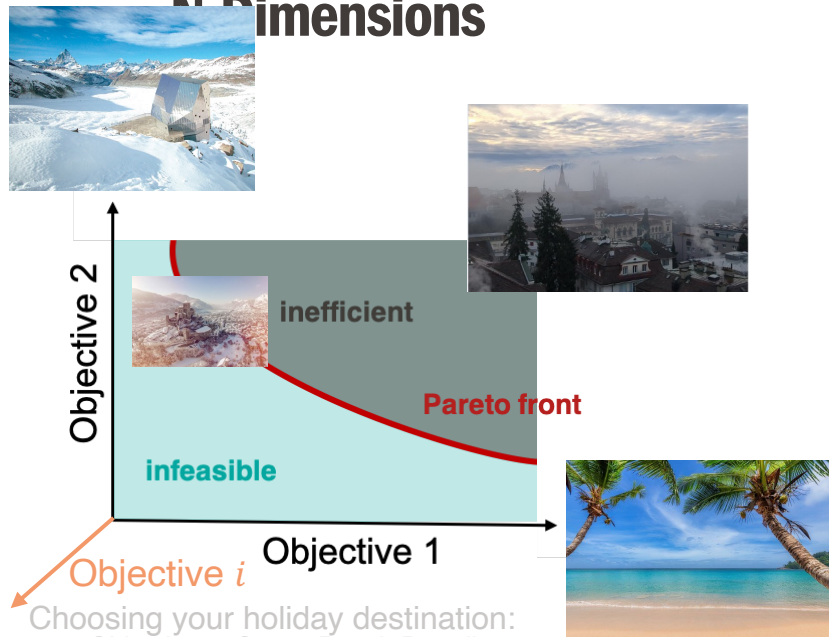
$$\text{s.t. } f_2^{obj} \leq \omega_2 \cdot f_2^{max} + (1 - \omega_2) \cdot f_2^{min}, \omega_2 \geq 0$$

When planning your holidays, finding the perfect balance between sunshine and snow activities can be a tricky optimization! Sion sits right on the Pareto front, offering the best of both worlds

Multi-Objective Optimization

N Dimensions

The Pareto front represents the set of all non-dominated solutions in a multi-objective optimization problem, where no objective can be improved without worsening at least one other objective.



Choosing your holiday destination:

- Objective 1: Sunny Beach Paradise
- Objective 2: Alpine Mountain Retreat
- Objective i: I want to minimize my CO₂ footprint
- Infeasible: "Zermatt sur mer" (?)
- Inefficient: Foggy Lausanne
- Compromise: Sion: Sunny, alpine, beautiful, ...

When planning your holidays, finding the perfect balance between sunshine and snow activities can be a tricky optimization! **Sion** sits right on the Pareto front, offering the best of both worlds

Methods to derive the Pareto-Front

- **Epsilon-Method**
Optimize one objective while treating the other objectives as constraints bounded by epsilon values.

$$\min f_i^{obj}(F, F_t)$$

$$\text{s.t. } f_j^{obj} < \epsilon_j$$

- **Weighted Sum Method**
Combine multiple objectives into a single objective using weighted coefficients.

$$\min \sum_i f_i^{obj} \cdot \omega_i$$

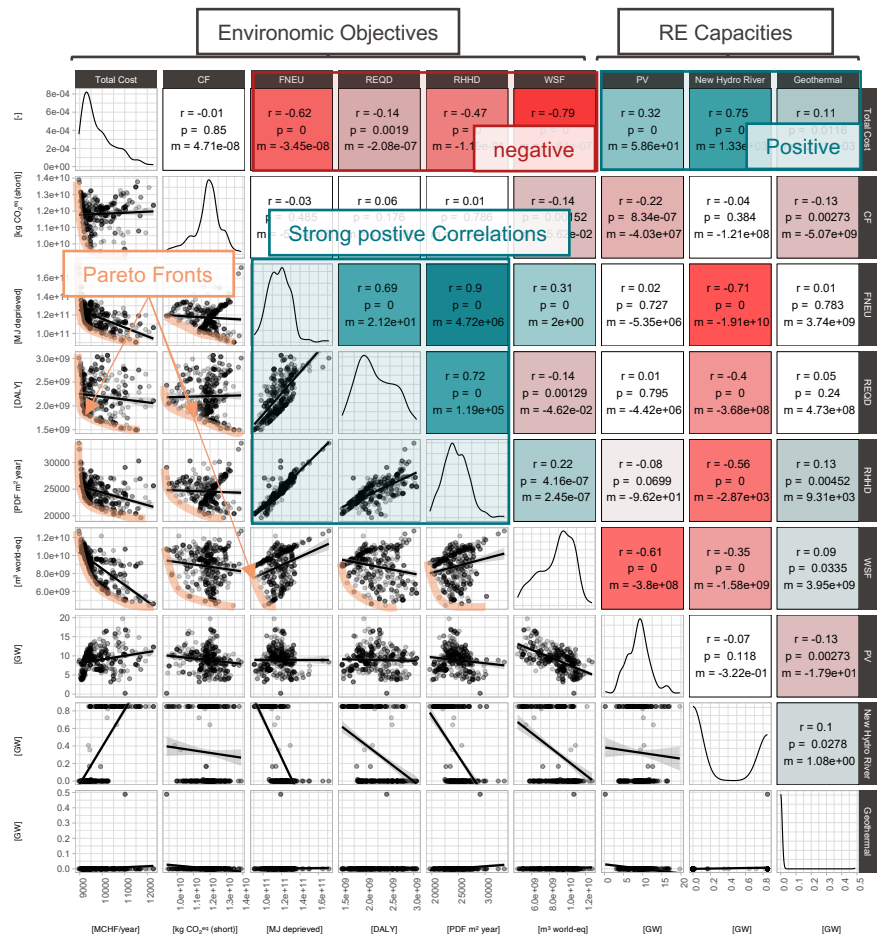
$$\text{s.t. } \sum_i \omega_i = 1, \omega_i \geq 0$$

- ! Issues with different orders of magnitude of f_i^{obj}

- **Weighted Epsilon-Method**
Define $f_i^{\min, \max}$ and parametrize using weighted epsilon values

$$\min f_j^{obj}(F, F_t)$$

$$\text{s.t. } f_i^{obj} \leq \omega_i \cdot f_i^{\max} + (1 - \omega_i) \cdot f_i^{\min}, \omega_i \geq 0$$



- Pareto-Optimal Solution Space

- Environmental Correlation

RHHD & FNEU & REQD

CF no significant correlations

- Economic:

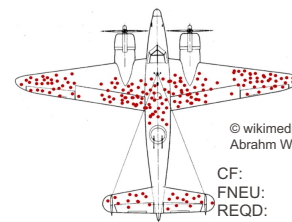
Negative correlation wrt sustainability

Positive correlation wrt technologies

! Correlations in the optimal solution space

- Wind always deployed at maximum

- Sub-optimal configurations & technologies not represented



© wikimedia commons
Abrahm Wald Problem

CF:	Carbon Footprint
FNEU:	Fossil & Nuclear Energy Use
REQD:	Remaining Ecosystem Quality Damage
RHHD:	Remaining Human Health Damage
WSF:	Water Scarcity Footprint

$$\begin{aligned} & \min_{\mathbf{F}, \mathbf{f}_i} \mathbf{C}_{\text{tot}} \\ & \text{s.t. } \mathbf{f}_{\text{obj}}(i) \leq \omega(i) \cdot \mathbf{f}_{\text{obj}}^{\max}(l) + (1 - \omega(i)) \cdot \mathbf{f}_{\text{obj}}^{\min}(l) \\ & \quad \forall \quad i \in \mathcal{OF} = \mathcal{C} \cup \mathcal{LCTA} - \mathcal{I} \\ & \mathbf{F}(i), \mathbf{F}_t(i): \quad f((\mathbf{F}(i), \mathbf{F}_t(i)), \omega(i)) \\ & \quad \text{s.t. } \omega(i) = P(\tilde{\omega}, \mathbf{U}(0, 1)) \end{aligned}$$

Conclusion

How can LCA indicators be integrated into energy system models to optimize both environmental and economic outcomes for the energy transition?

- **Integration of LCA Indicators**
 - Essential for balancing economic efficiency with environmental sustainability
 - Environmental optimization = -15% to -33% costs compared 2020
- **Multi-Parametric Optimization for generating configurations**
 - Reveals positive correlation between economic and environmental objectives
 - Sustainable MOO solution space = renewable energies
- **Burden shifting**
 - Economic optimization efficiency reduces Carbon Footprint by 63% compared to 2020
 - but shifts burdens to other environmental areas Water Scarcity, Fossil & Nuclear Energy use, due to the construction of technologies
- **Focus on Environomic Indicators**
 - Tracking potential burden shifting
 - Quantifies benefits and burden associated to different dimensions of sustainability
- **Limitations**
 - Static LCA based on historic activities
 - Not considering prospective LCA





A G E N D A

Energy System Modeling

Carbon Cycle

Technologies' Impacts

Optimization

Multi-Objective Optimization

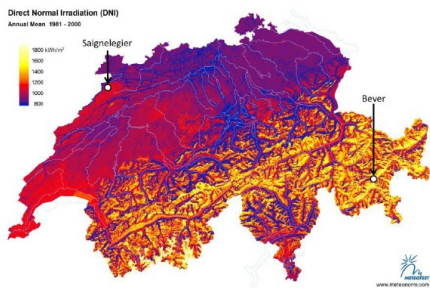
Addressing contemporary challenges



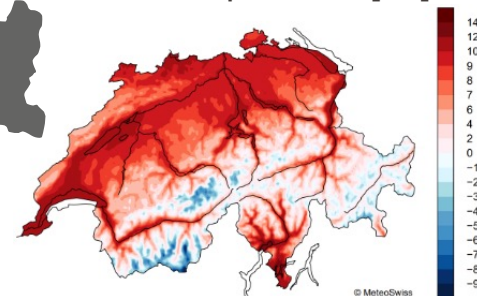
District archetypes

How to characterize the Swiss decentralized energy system

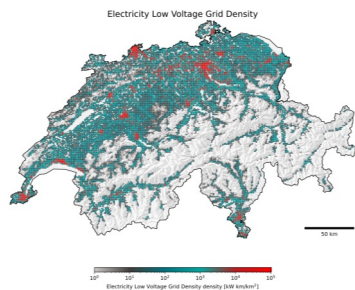
Irradiance [W/m^2]



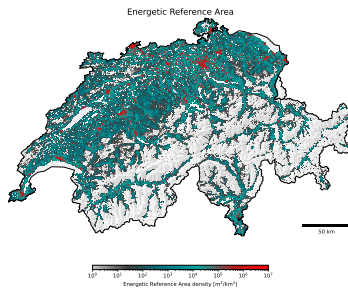
Temperature [$^{\circ}\text{C}$]



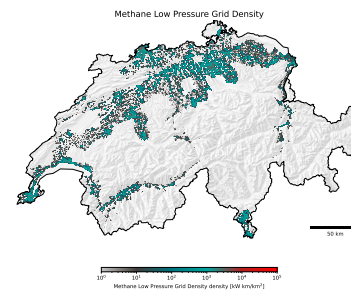
Electric grid [Wm/m^2]



Demand Density [W/m^2]



Gas grid [Wm/m^2]



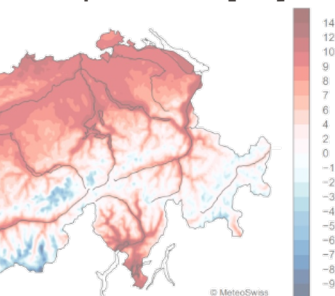
District archetypes

How to characterize the Swiss decentralized energy system

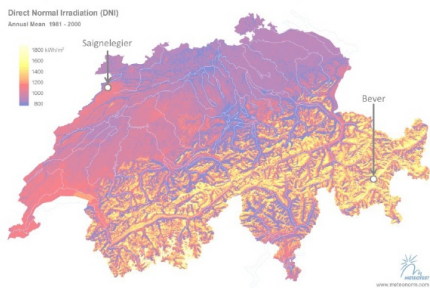
Geographic Clusters Distribution

- Alpine
- Alpine w/o Gas
- Countryside
- Countryside w/o Gas
- Rural
- Sub-Urban
- Urban

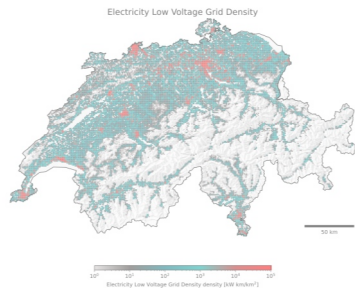
Temperature [°C]



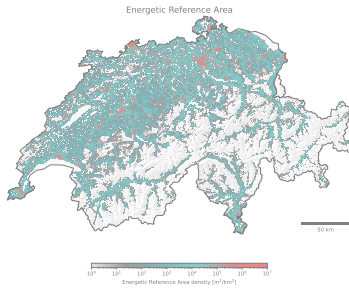
Irradiance [W/m²]



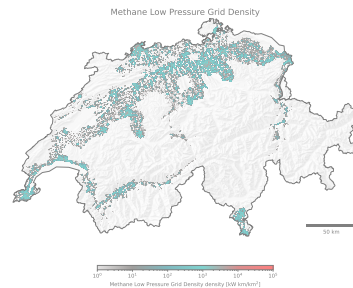
Electric grid [Wm/m²]



Energy demand [Wm/m²]



Gas grid [Wm/m²]



Prosumers

Integrating Self-Consumption & Investments at District Scale

District KPI



Single family house



Multi family house



Chalet



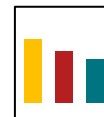
Office Tower



Gas
Electricity



School



Mansion



Energy demands:

- Electricity
- Space heating
- Hot water



Communities of Prosumers

Integrating Self-Consumption & Investments at District Scale

District KPI



Energy demands:

- Electricity
- Space heating
- Hot water



Single family house



Multi family house



Chalet



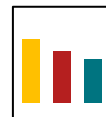
Office Tower



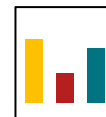
Gas
Electricity



School

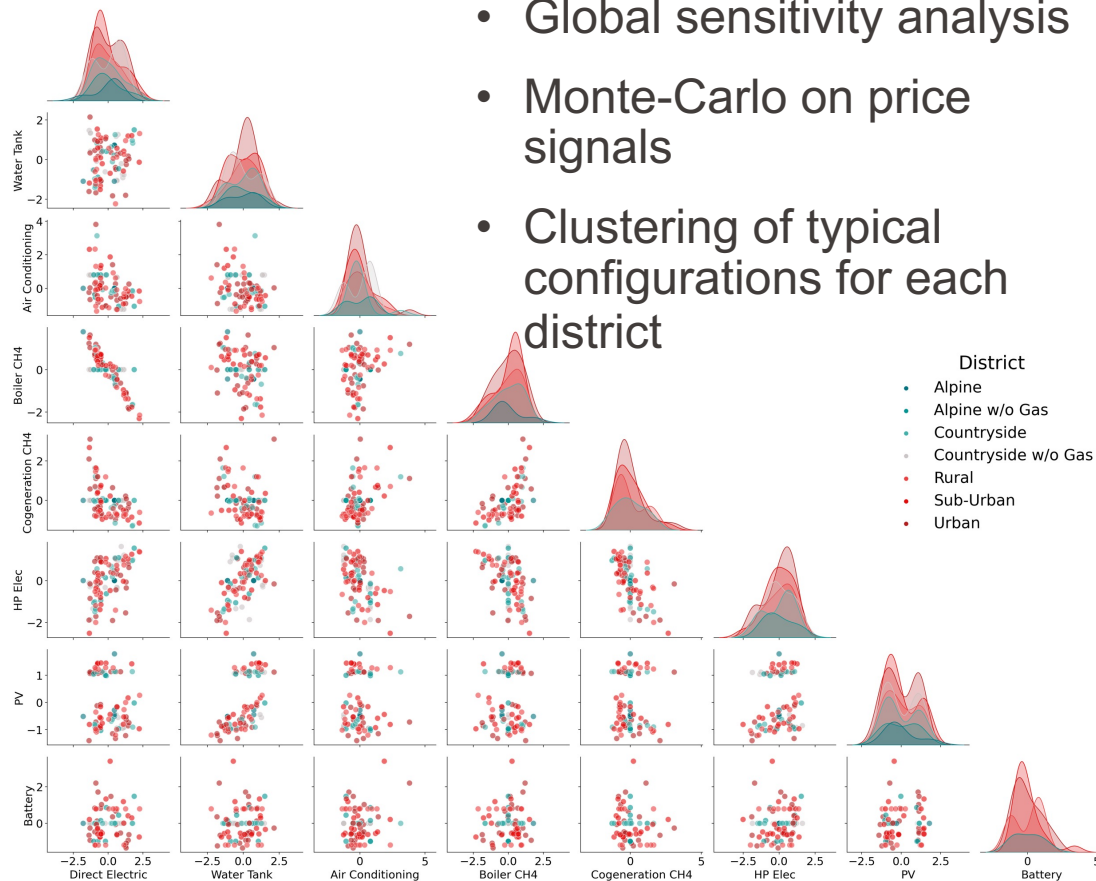
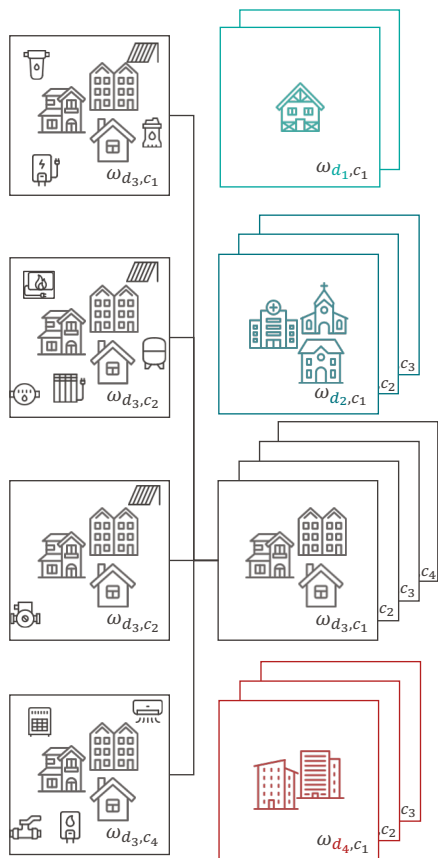


Mansion



District energy system archetypes

Generating configurations by parametric optimization



Linking national and district model

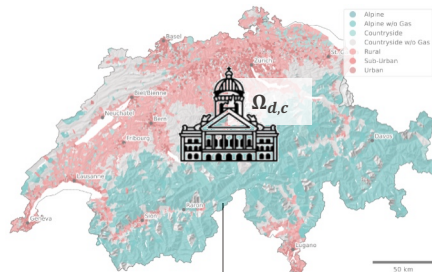
Selecting regional optimal configurations from the national point of view

Objective Function

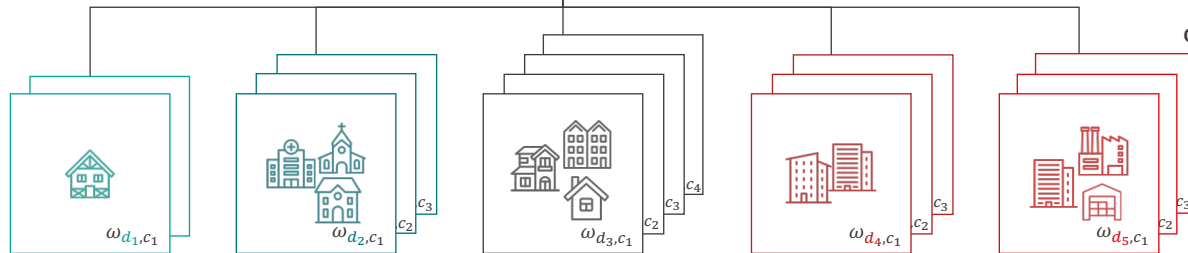
$$\min_{F, F_t, \Phi} C_{\text{tot}} \quad \text{s.t.}$$

$$C_{\text{tot}} = C_{\text{tot}}^{\Omega} + C_{\text{tot}}^{\omega}$$

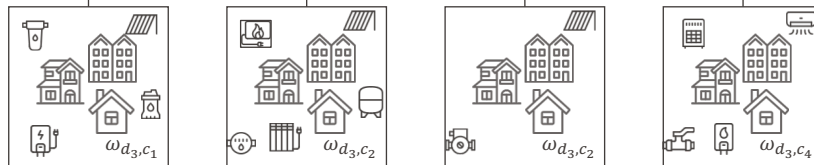
national optimization



Typification



regional configurations



Total Cost Centralized System

$$C_{\text{tot}}^{\Omega} = C_{\text{op}}^{\Omega} + C_{\text{inv}}^{\Omega} + C_{\text{maint}}^{\Omega}$$

$$C_{\text{op}}^{\Omega} = \sum_{\text{res}} \sum_t C_{\text{op}}(\text{res}, t) \cdot F_t(\text{res}, t) \cdot t_{\text{op}}(t)$$

$$C_{\text{inv}}^{\Omega} = \sum_{\text{tec}} C_{\text{inv}}(\text{tec}) \cdot (F(\text{tec}) - F_3^{\Omega}(\text{tec}^*))$$

$$C_{\text{maint}}^{\Omega} = \sum_{\text{tec}} C_{\text{maint}}(\text{tec}) \cdot F(\text{tec})$$

Configuration Selection

$$\sum_c \Phi(c, d) = 1, \quad 0 \leq \Phi(c, d) \leq 1$$

Energy Balance

$$\begin{aligned} \text{EU}(l, t) = & \sum_{\text{tec}} F_t(\text{tec}, t) \cdot \eta(\text{tec}, l) - F_t^{\text{Loss}}(l, t) \\ & + \sum_{\text{sto}} F_t^+(sto, l, t) - F_t^-(sto, l, t) \\ & + \sum_d F_t^{\omega+}(d, l, t) - F_t^{\omega-}(d, l, t) \end{aligned}$$

Total Cost Districts

$$C_{\text{tot}}^{\omega} = \sum_d (C_{\text{inv}}^{\omega}(d) + C_{\text{maint}}^{\omega}(d))$$

$$C_{\text{inv}}^{\omega}(d) = \sum_{\text{tec}} \sum_c (C_{\text{inv}}(\text{tec}) \cdot f^{\omega}(\text{tec}, c, d) \cdot \Phi(c, d))$$

$$C_{\text{maint}}^{\omega}(d) = \sum_{\text{tec}} \sum_c (C_{\text{maint}}(\text{tec}) \cdot f^{\omega}(\text{tec}, c, d) \cdot \Phi(c, d))$$

$$\forall c \in S, d \in \mathcal{D}, t \in \mathcal{P},$$

$$\text{tec} \in \mathcal{T}, \text{tec}^* \in \mathcal{G},$$

$$\text{res} \in \mathcal{R}, \text{sto} \in \mathcal{T} - S, l \in \mathcal{L}$$

Linking national and district model

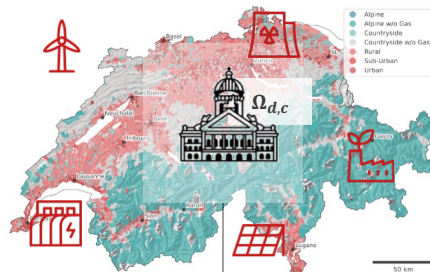
Selecting regional optimal configurations from the national point of view

Objective Function

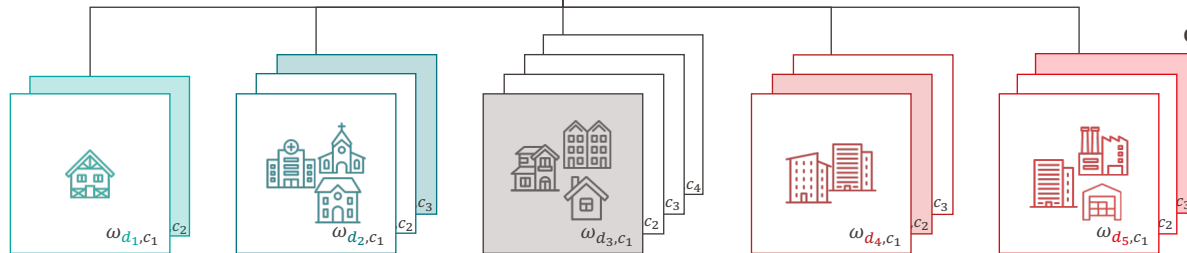
$$\min_{F, F_t, \Phi} C_{\text{tot}} \quad \text{s.t.}$$

$$C_{\text{tot}} = C_{\text{tot}}^{\Omega} + C_{\text{tot}}^{\omega}$$

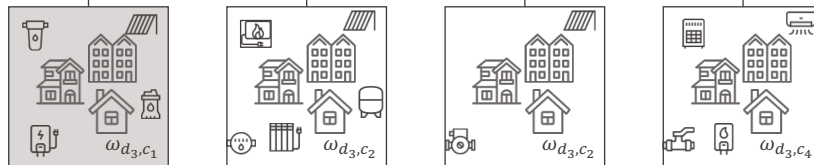
national optimization



Typification



regional optimization



Total Cost Centralized System

$$C_{\text{tot}}^{\Omega} = C_{\text{op}}^{\Omega} + C_{\text{inv}}^{\Omega} + C_{\text{maint}}^{\Omega}$$

$$C_{\text{op}}^{\Omega} = \sum_{res} \sum_t c_{\text{op}}(res, t) \cdot F_t(res, t) \cdot t_{\text{op}}(t)$$

$$C_{\text{inv}}^{\Omega} = \sum_{tec} c_{\text{inv}}(tec) \cdot (F(tec) - F_{\text{I}}^{\Omega}(tec^*))$$

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$$C_{\text{maint}}^{\omega}(d) = \sum_{tec} \sum_c (c_{\text{maint}}(tec) \cdot f^{\omega}(tec, c, d) \cdot \Phi(c, d))$$

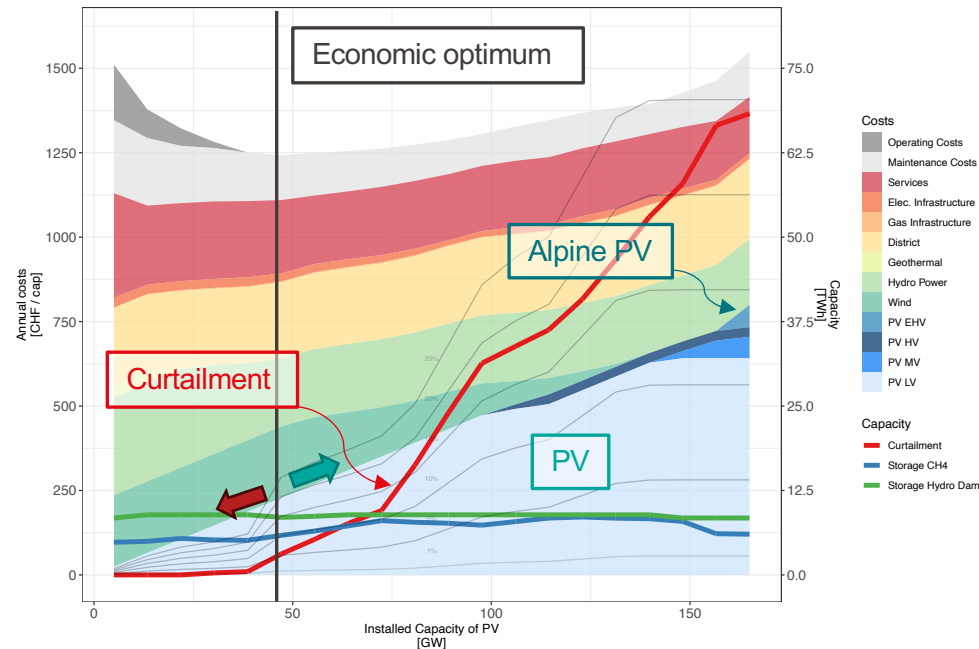
$$\forall \quad c \in S, d \in \mathcal{D}, t \in \mathcal{P},$$

$$tec \in \mathcal{T}, tec^* \in \mathcal{G},$$

$$res \in \mathcal{R}, sto \in \mathcal{T} - S, l \in \mathcal{L}$$

Wind-PV tradeoff & *self-consumption*

The transition towards a decentralized system

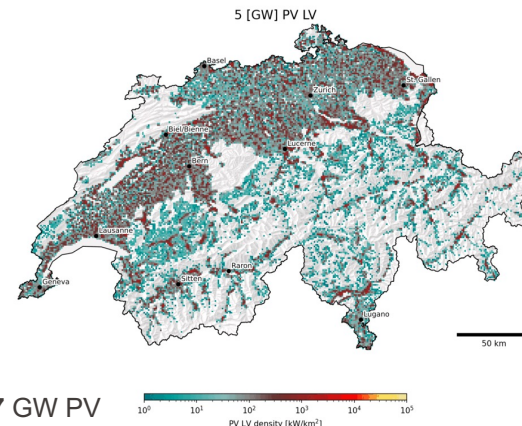


Evolution of energy system costs composition and storage capacities of the Swiss energy system

PV installation parametrization.

The transparent lines represent the annual PV-LV production fractions, allowing us to compare them with the curtailment depending on the installed PV capacity. The case study represents the economic optimization of a neutral (no net emission) and independent (no import) Swiss energy system in 2050 for a population of 10 Million.

Animation of the geographic PV installation density



- Minimum Cost:
20 GW Wind & 37 GW PV
 - PV: Limitation by the LV grid but more (37 GW)
 - Wind: installation to its maximum potential (20 GW)
- PV
 - Wind at maximum
 - Compensation by biomass resources
0-15% biomass potential
 - Methane storage via power-to-methane (4.3-6.1 TWh)
- PV
 - Wind reduction
 - Seasonal dephasing
 - Methane storage via power-to-methane (6.1-8.8 TWh)

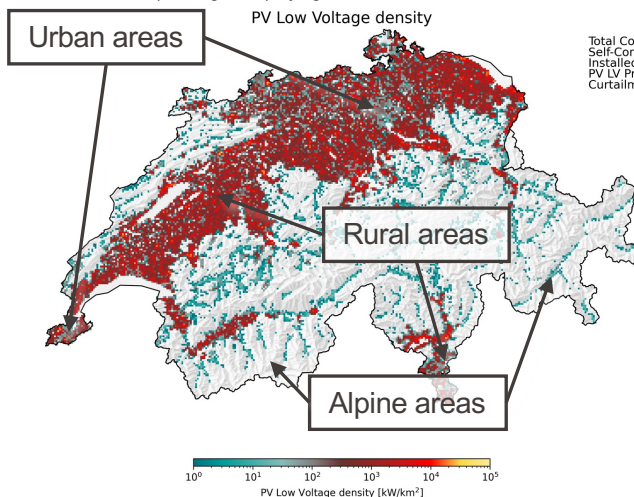
What about the grid?

PV deployment and grid reinforcement in the transition towards a decentralized system

Urban areas:

- Limited PV deployment
50-250 kW/km²
- Reinforcement due to electrification of heating sector
5-21 MW km / km²

Minimal Cost corresponding to deploying 37 [GW] PV LV



Rural areas:

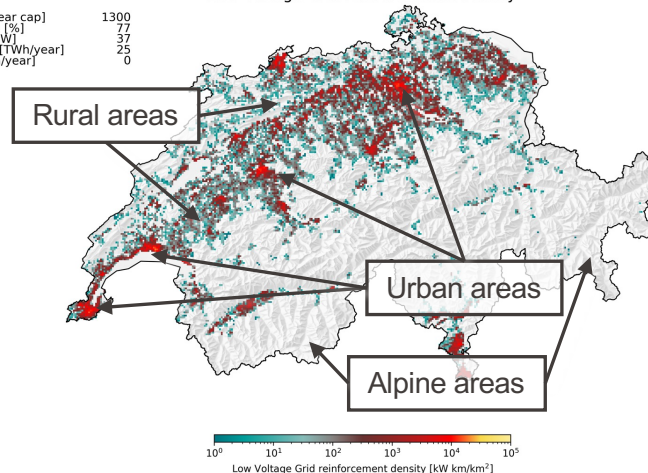
- High PV deployment
1-50 MW/km²
- Self-consumption & Export to urban
- Low reinforcement due to lower energy demands
0-0.5 MW km / km²

Alpine areas:

- PV deployment to maximise self-consumption
15-100 kW km / km²
- No export
- No reinforcements needed

Total Cost [CHF/year cap] 1300
Self-Consumption [%] 77
Installed PV LV [GW] 37
PV LV Production [TWh/year] 25
Curtailment [TWh/year] 0

Low Voltage Grid reinforcement density



Geographical deployment of LV PV and respective grid reinforcement in CH 2050.
Case study of the economic optimization of a neutral (no net emissions) and independent (no imports) Swiss energy system in 2050.
Cost minimization

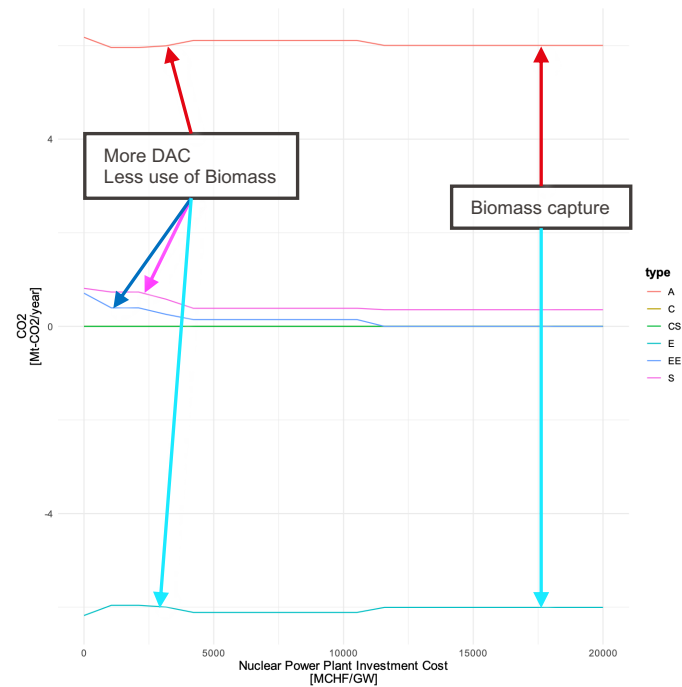
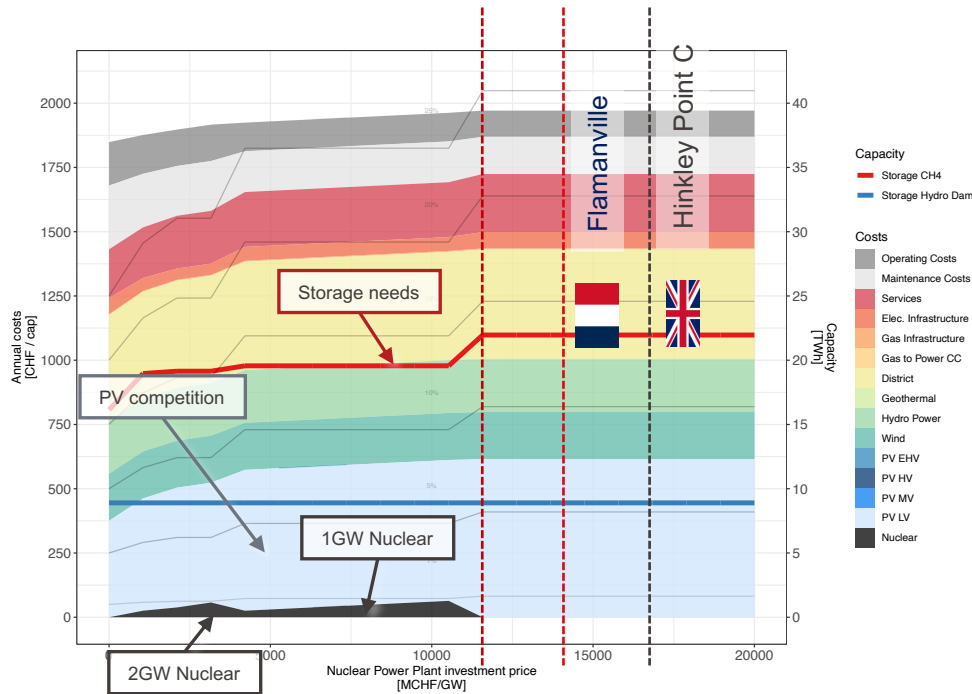
Independence of Switzerland

Critical price of methane import price



Independence of Switzerland

Critical price of nuclear power plants



Une énergie suisse et décarbonnée d'ici à 2050 ? L'HES-SO et l'EPFL amènent leur éclairage...

Swiss and decarbonized Swiss Energy by 2050? The HES-So and EPFL enlighten the question

