

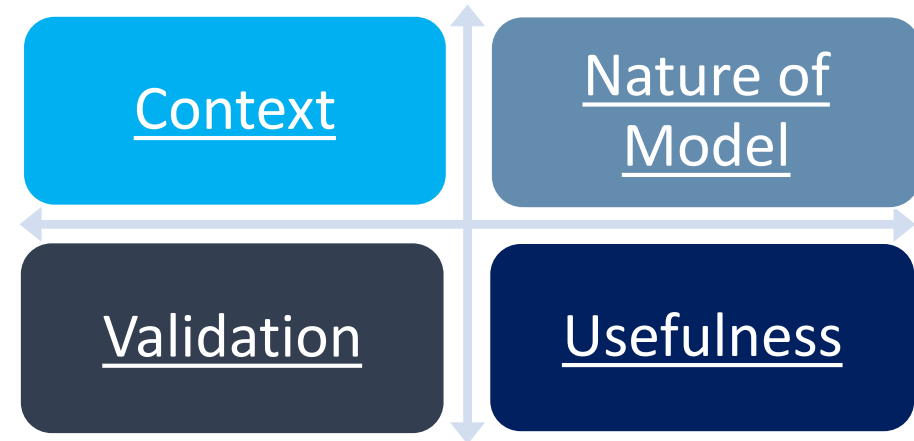
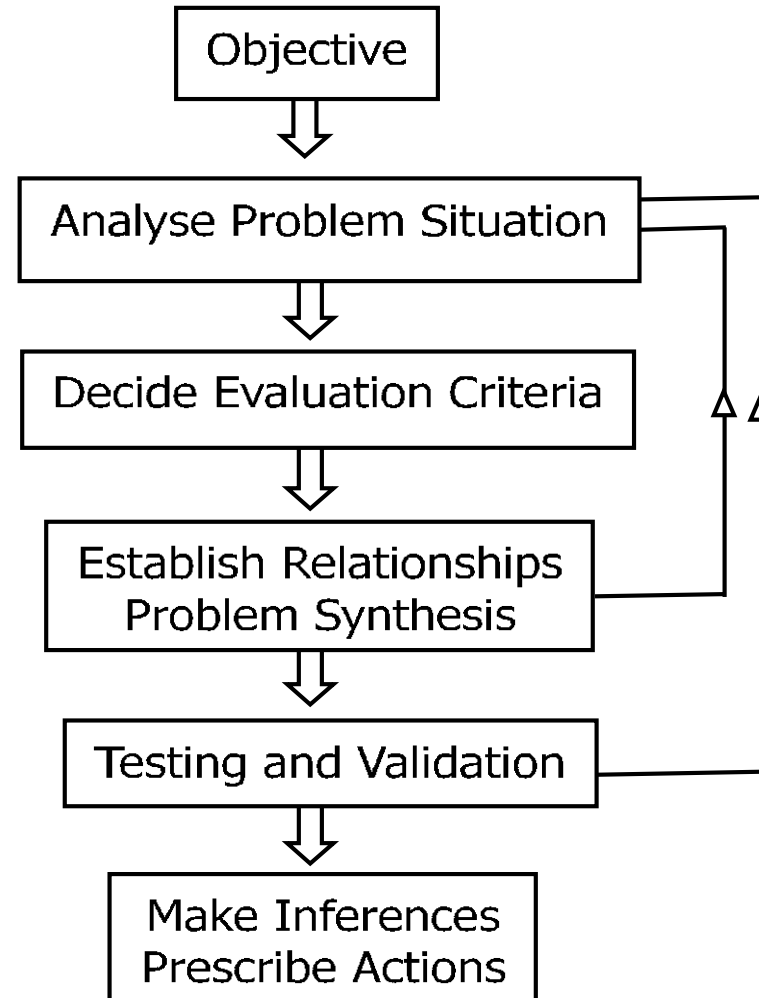
Modelling of Energy Systems

Rangan Banerjee

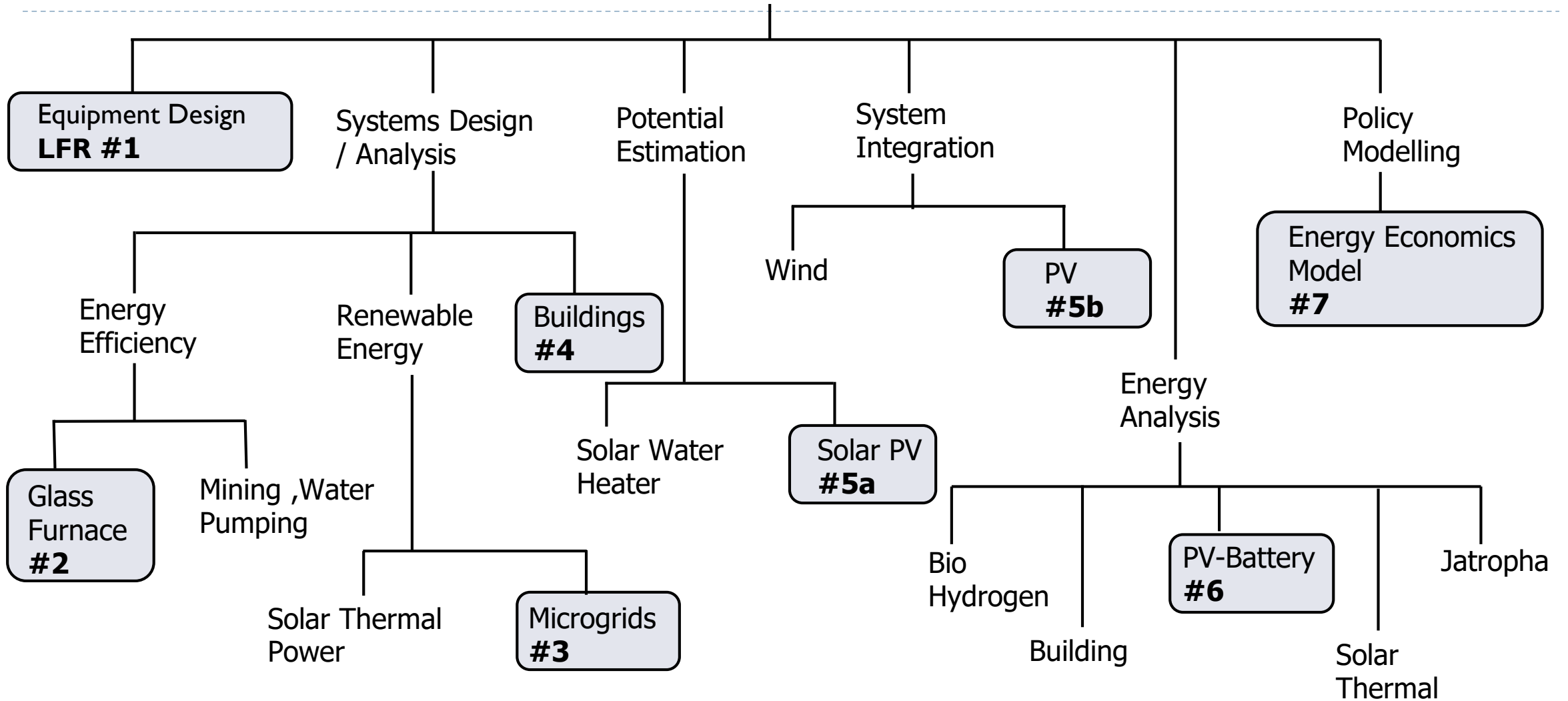
Department of Energy Science & Engineering,
IIT Bombay

Web: <http://www.ese.iitb.ac.in/~rb/>

- What is an Energy System?
- Why model Energy Systems?

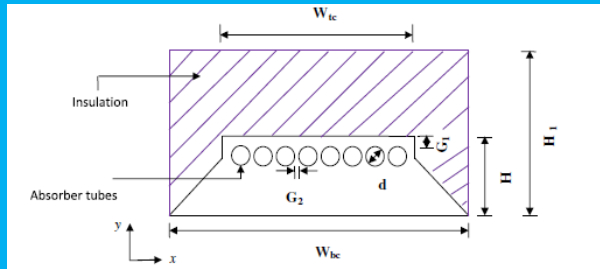


Energy Systems Modelling



Context

- Heat Losses Trapezoidal Cavity Linear Fresnel Reflector (LFR)

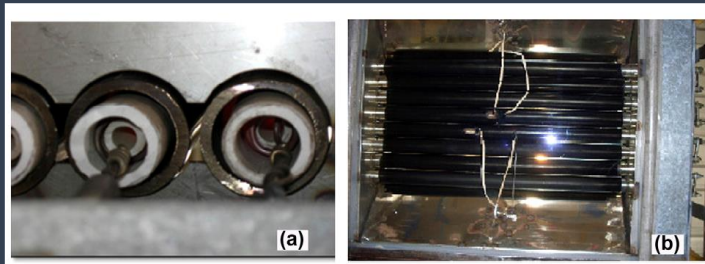


Nature of Model

- Two Dimensional- Mass, momentum, energy balances
- Fluent

Validation

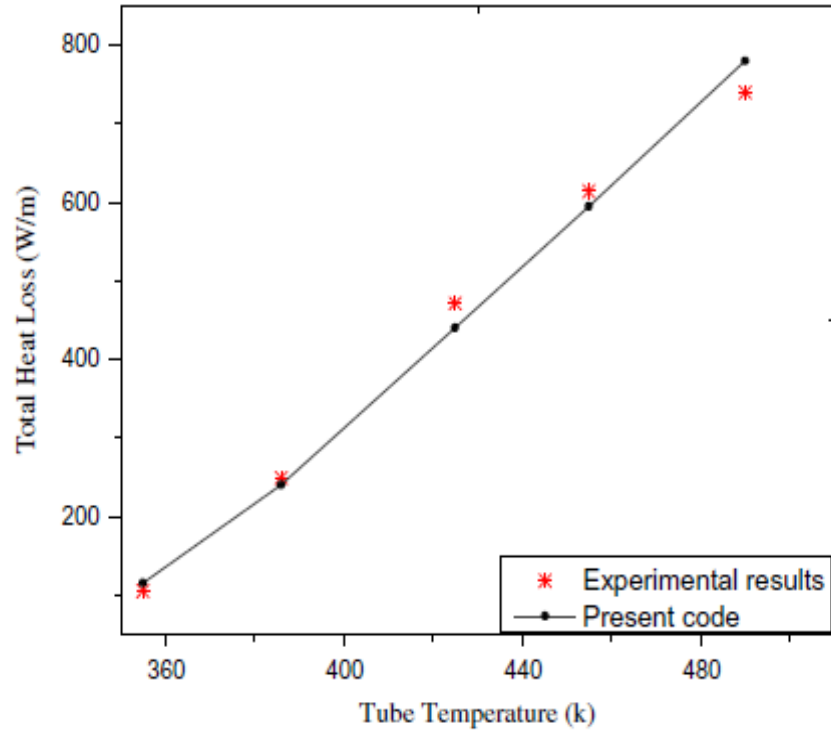
- Experimental setup, electric Heating, KG design



Usefulness

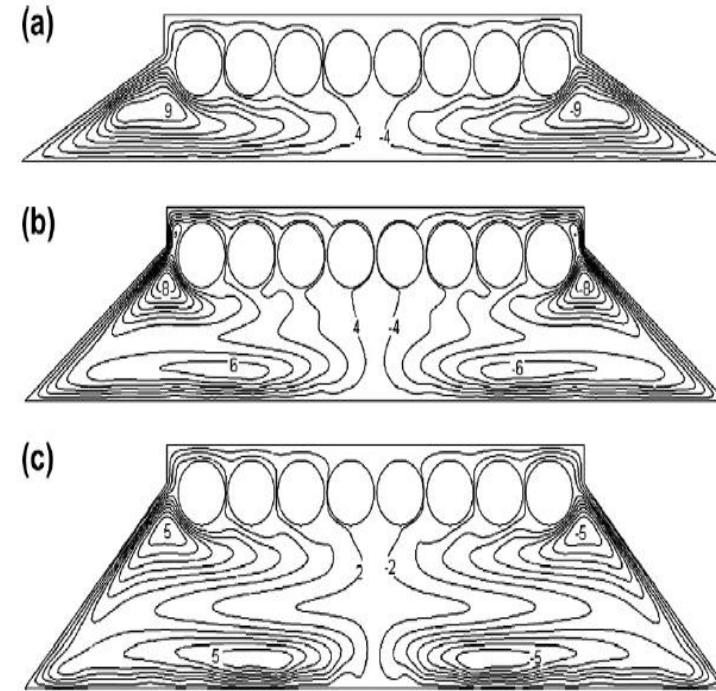
- Improved understanding of heat losses from LFR cavities
- Effect of variation of design parameters

$$\overline{Nu}_t = 0.084 Ra_d^{0.19} (T^*)^{0.53} \left(\frac{1}{H^*} \right)^{0.095} (1 + \varepsilon)^{2.292} \left(\frac{h_{ext} W_{bc}}{k_{air}} \right)^{0.0304} \quad (91)$$



Validation of heat loss

Sahoo et al, Renewable Energy, 2016



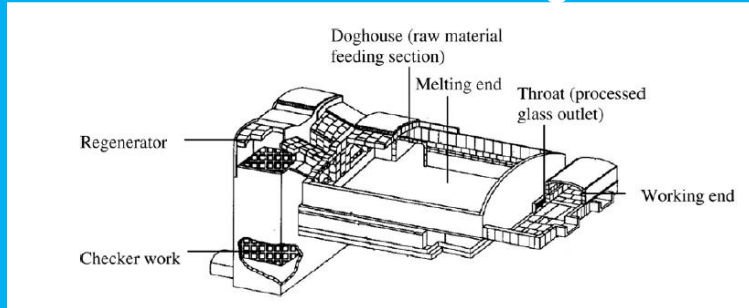
Effect of different height

$$H^* = H/d$$

$H^* =$ a)2.25, b)3.0,c)3.75

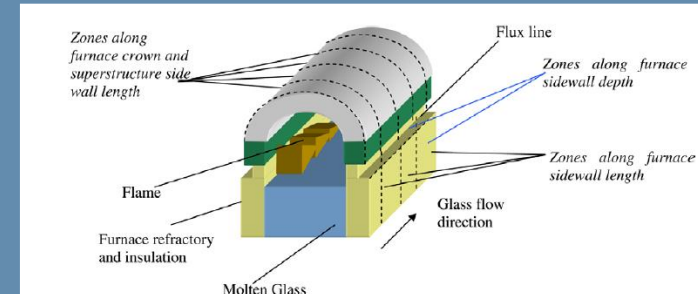
Context

- Industrial Glass furnace
- Model based benchmarking



Nature of Model

- Mass and Energy Balances –zones
- Heat loss correlations

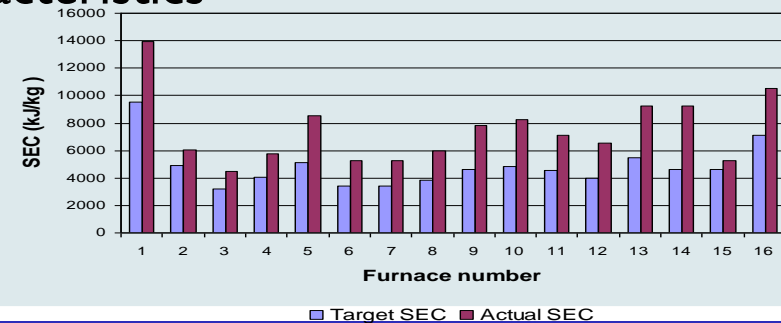


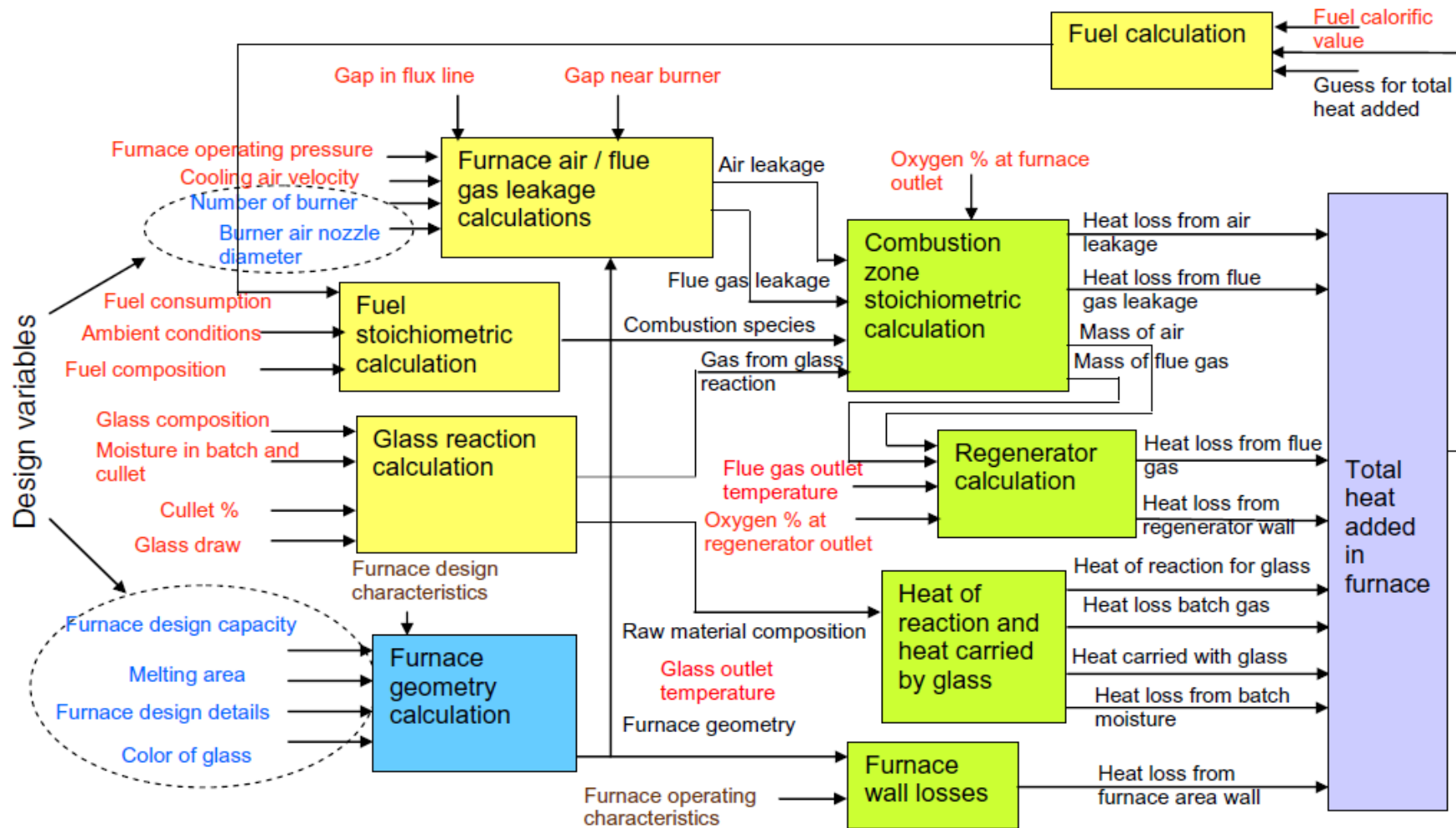
Validation

- Field Measurements, nine industrial glass furnaces
- Industry workshop

Usefulness

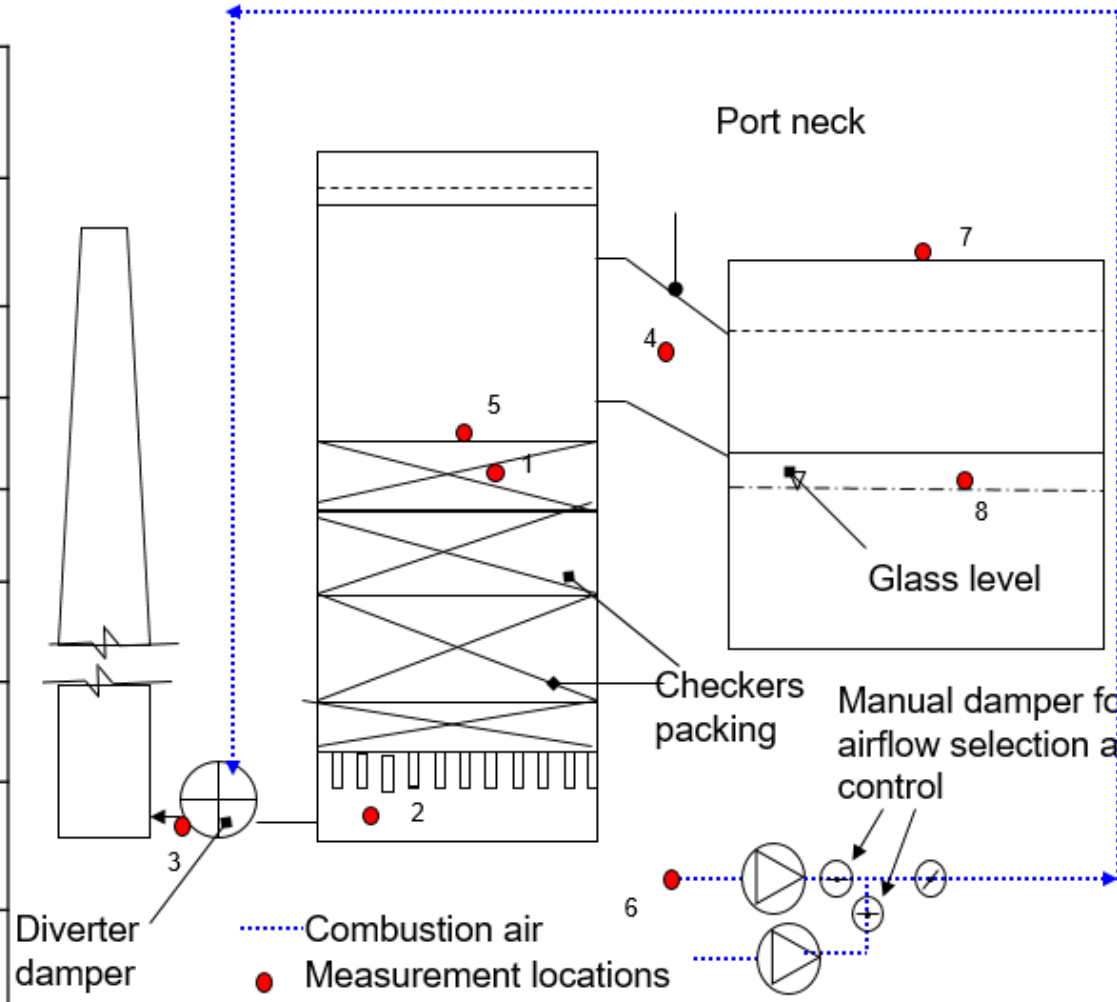
Target SEC – considering furnace characteristics



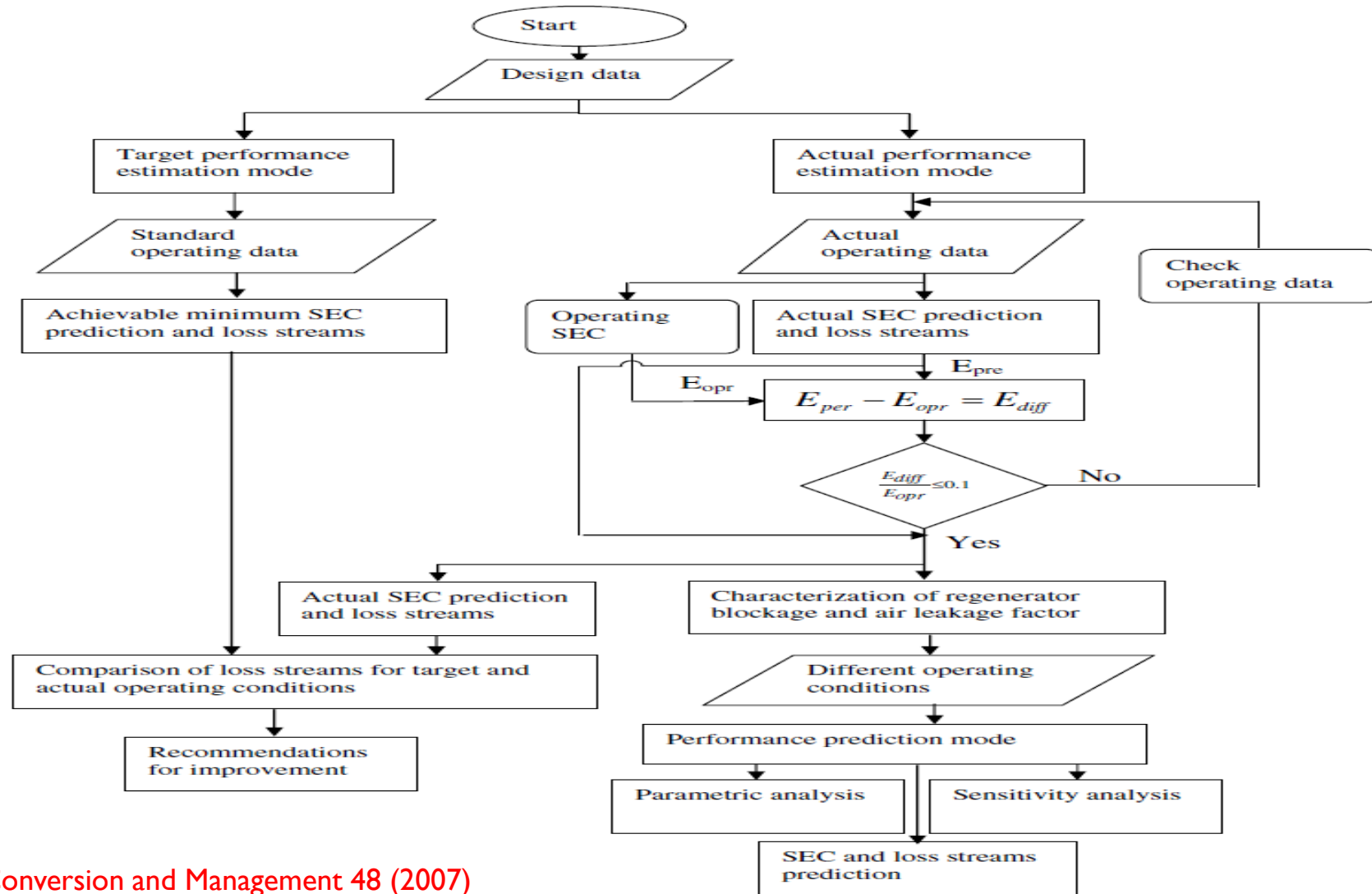


Vishal S. et al. Energy Conversion and Management 48 (2007)

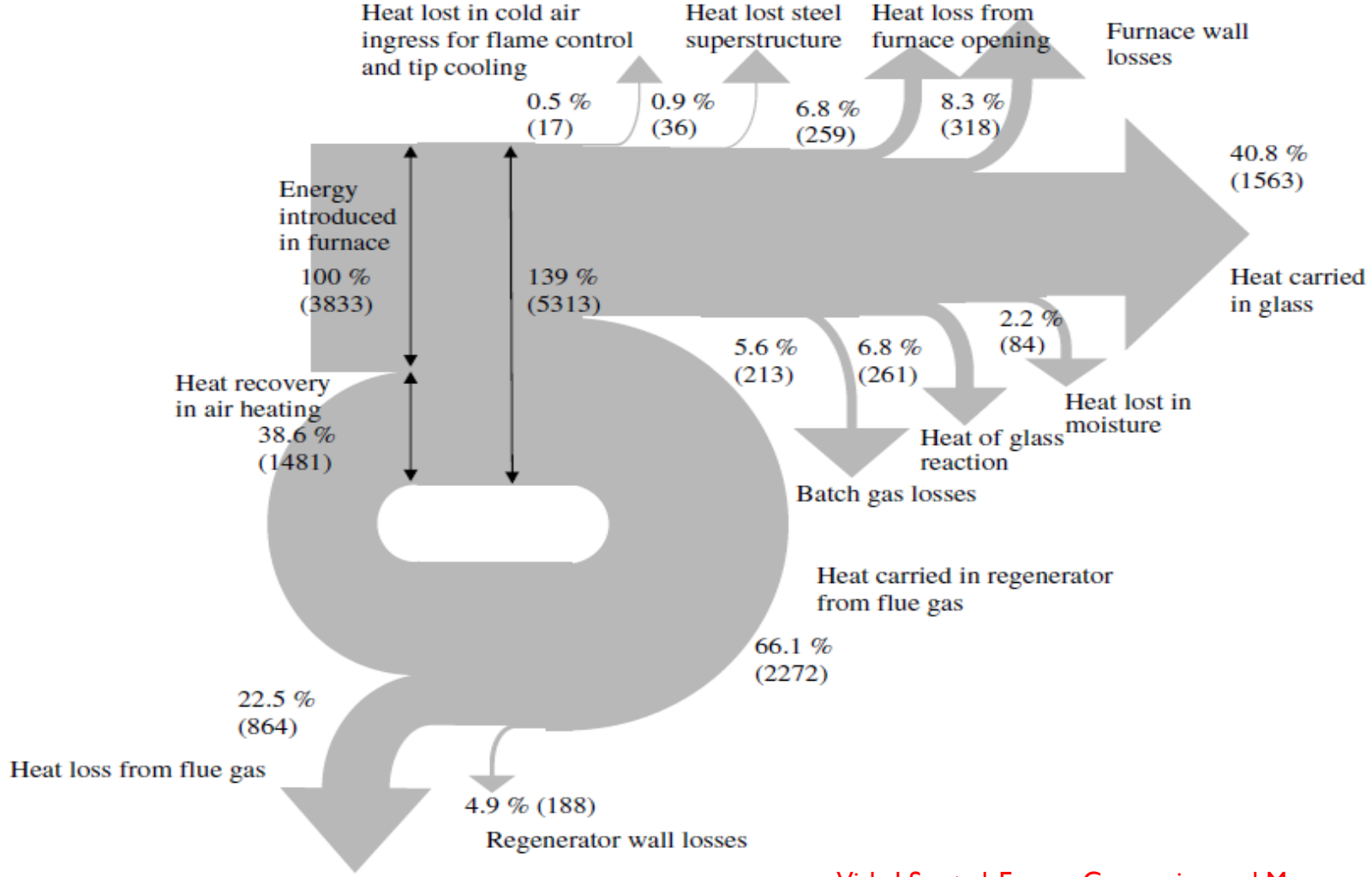
| Measurement location | Type of measurement |
|----------------------|---|
| 1 | Oxygen % , Pyrometer checkers surface temperature |
| 2 | Oxygen % , Flue gas temperature |
| 3 | Oxygen % , Flue gas temperature |
| 4 | Oxygen % , Skin temperature |
| 5 | Pyrometer checkers surface temperature |
| 6 | Velocity of air at the suction of blower |
| 7 | Outside wall temperature for crown and side wall |
| 8 | Pyrometer glass surface temperature |



V.Sardeshpande et al.Applied Energy 88 (2011)



Vishal S. et al. Energy Conversion and Management 48 (2007)



Vishal S. et al. Energy Conversion and Management 48 (2007)

Context

- Microgrid –isolated, grid support
- PV- battery- ultracapacitor- hydrogen storage
- Optimal sizing hybrid storage
- Supply, Demand variability

Nature of Model

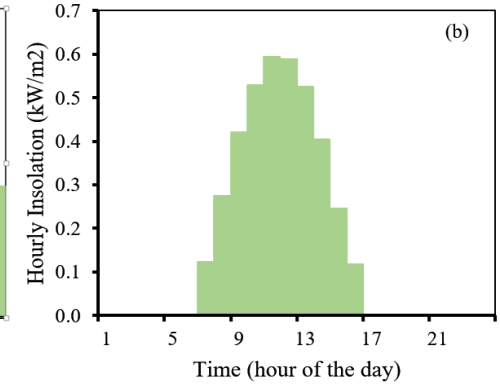
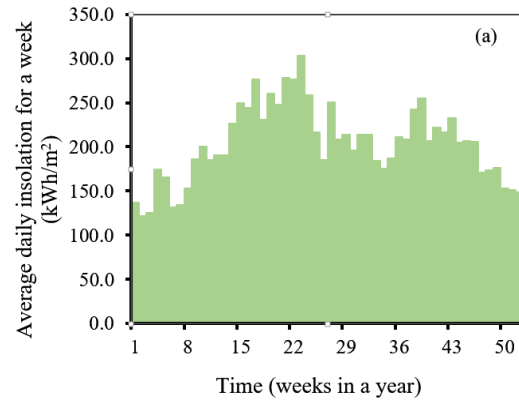
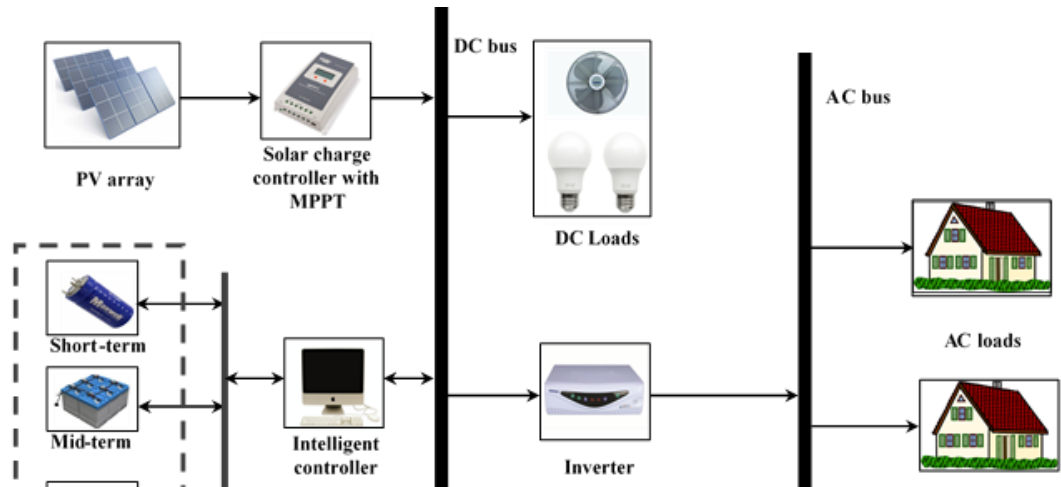
- Lumped parameter model for components
- Cumulative Energy supply and demand curves-Pinch , Design space approach
- Different time horizons
- Polynomial equations , Optimisation

Validation

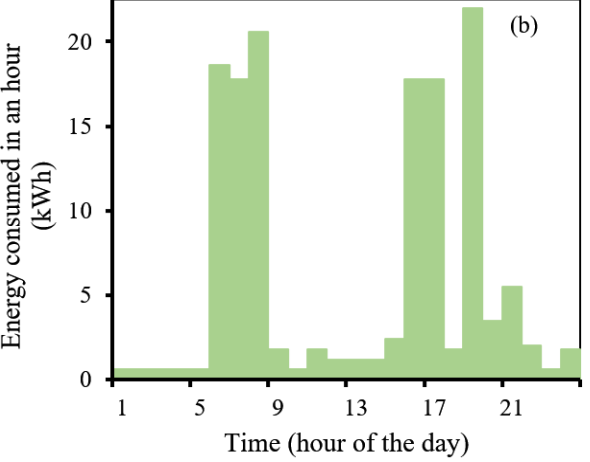
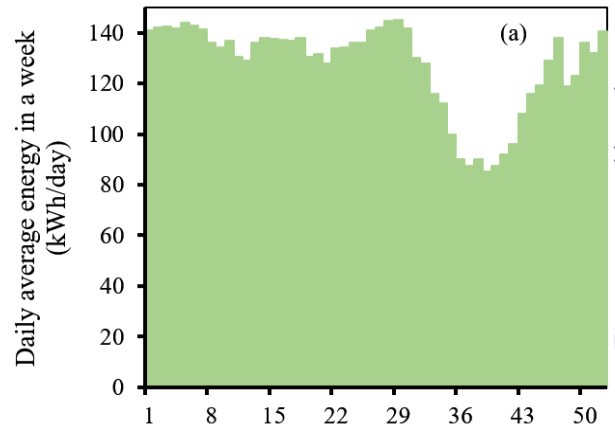
- Data logging for load variation – lift, welding loads
- Characteristics of components – manufacturers catalog
- Comparison with detailed simulation

Usefulness

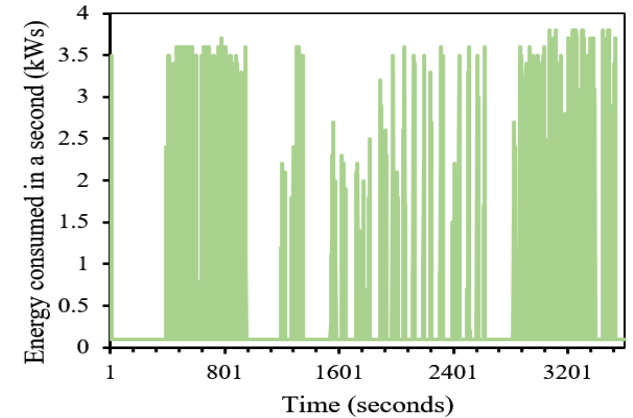
- Identification of feasible regions – design space
- Based on cost – can obtain minimum cost hybrid solutions



Insolation at New Delhi: (a) weekly average annual data and (b) hourly average daily data for a winter day



For a **remote village** (a) Seasonal variation of load for the year, (b) hourly variation of load



For a remote **welding shop** instantaneous variation in load for an hour

Jacob et al., Applied Energy, 2017

- Rate of Energy stored in storage device,

- $$\frac{dQ_s}{dt} = (P(t) - D(t))\eta_{stored} \quad P(t) - \text{Source Power, } D(t) - \text{Load Power}$$

Or
$$Q_s(t + \Delta t) = Q_s(t) + \int_t^{t+\Delta t} (P(t) - D(t))\eta_{stored}$$

- For a very small time interval ' Δt '

$$Q_s(t + \Delta t) = Q_s(t) + (P(t) - D(t))\eta_{stored}$$

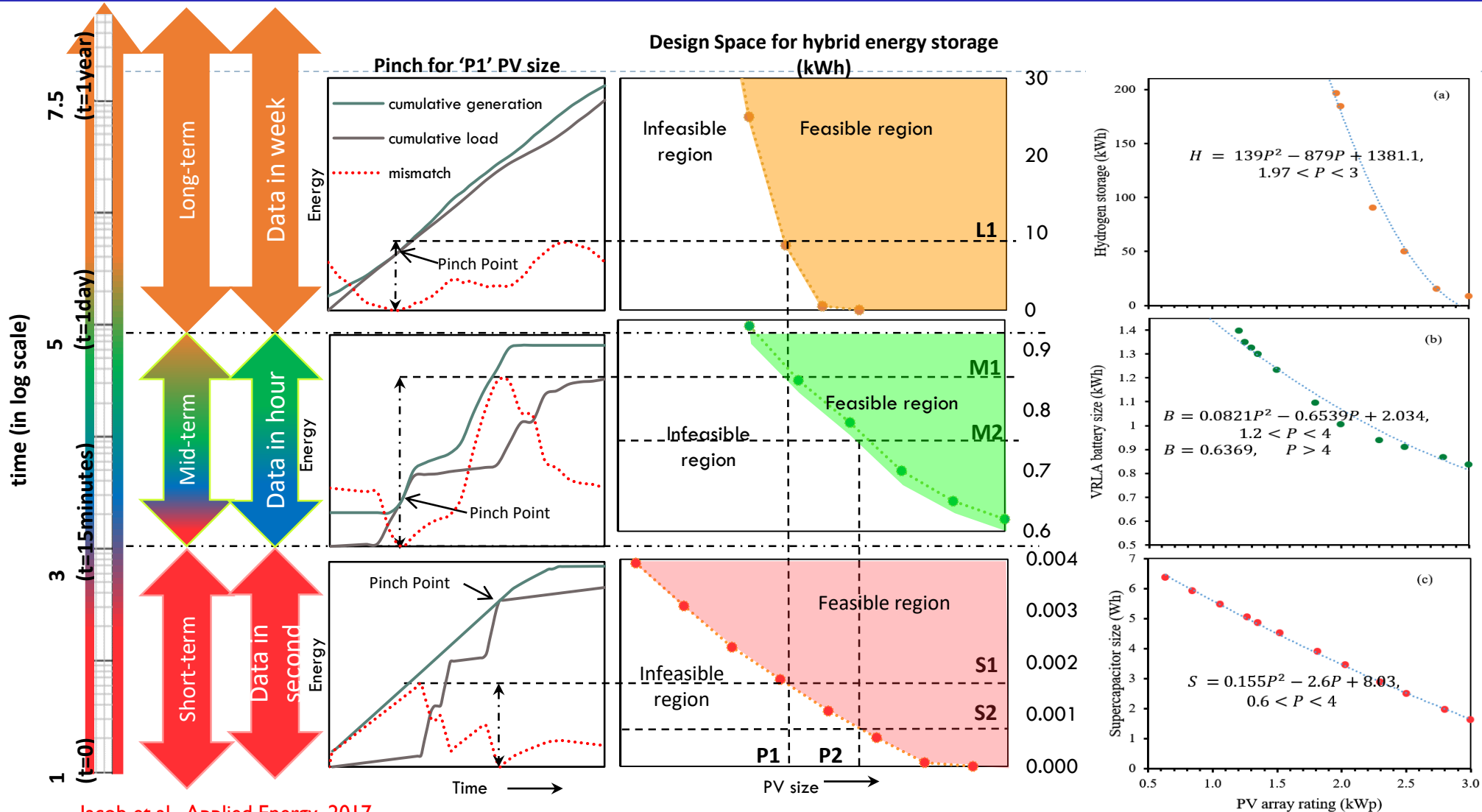
- Constraints

1) For the entire time horizon, T $Q_s(t=0) = Q_s(t=T)$

2) Storage charge level $Q_s(t) \geq 0$ for all time values

3) Generation should not exceed maximum power

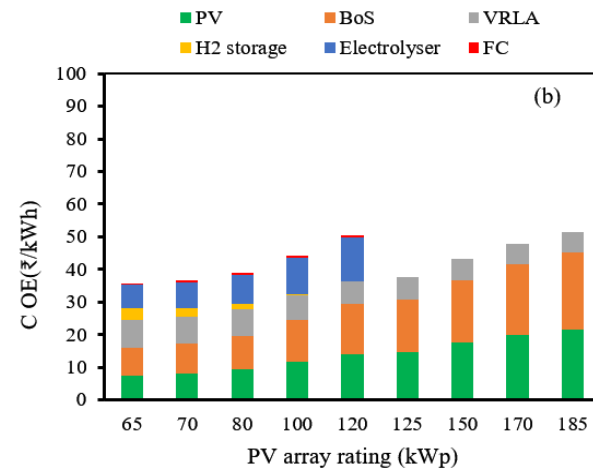
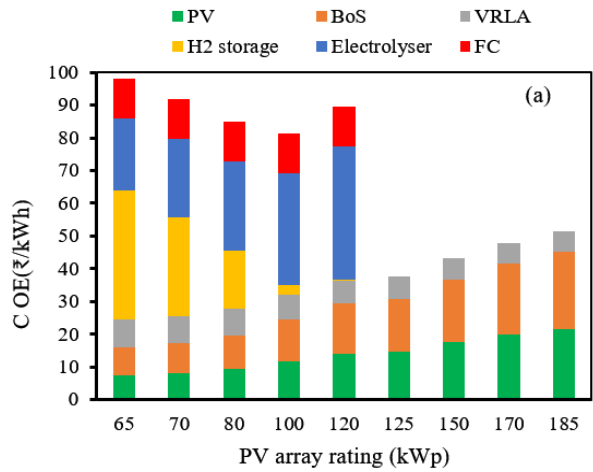
$$\text{Minimum Storage Capacity} = \frac{\max(Q_s)}{DoD}$$



Jacob et al., Applied Energy, 2017

Quadratic fitting for the boundary points of the design space of stand-alone welding shop

- (a) Hydrogen storage,
- (b) VRLA battery
- (c) Supercapacitor



| Case Study | Optimum Configuration | | | | | | COE (₹/kWh) |
|--------------------------|-----------------------|----------------|-------------------|---------------------------------|--------------------|---------------------|-------------|
| | PV (kW _p) | Fuel Cell (kW) | Electrolyser (kW) | Hydrogen tank (m ³) | VRLA battery (kWh) | Supercapacitor (Wh) | |
| Case 1 – Rural Village | 65 | 12 | 65 | 12.3 | 165 | - | 35 |
| Case 2 – Telecom tower | 40 | 6 | 40 | 5.2 | 58 | - | 33 |
| Case 3 – Welding Shop | 2 | 0.32 | 2 | 0.27 | 0.78 | 3.5 | 24 |
| Case 4 – Backup for lift | 1 | - | - | - | 2.7 | 69 | 30 |

COE (₹/kWh) for the different configurations of remote rural village case study (a) using present cost (b) using US DOE target cost for hydrogen storage

Jacob et al., Applied Energy, 2017

Context

- Direct coupling of Optimisation algorithm with simulation computationally intensive for Building design
- Three storey building in Delhi

Nature of Model

- Surrogate Models to approximate simulation
- Experimental Design (Fractional Factorial Design, Response Surface Methodology)
- Annual Cooling, Lighting Energy
- Optimisation- Genetic Algorithms

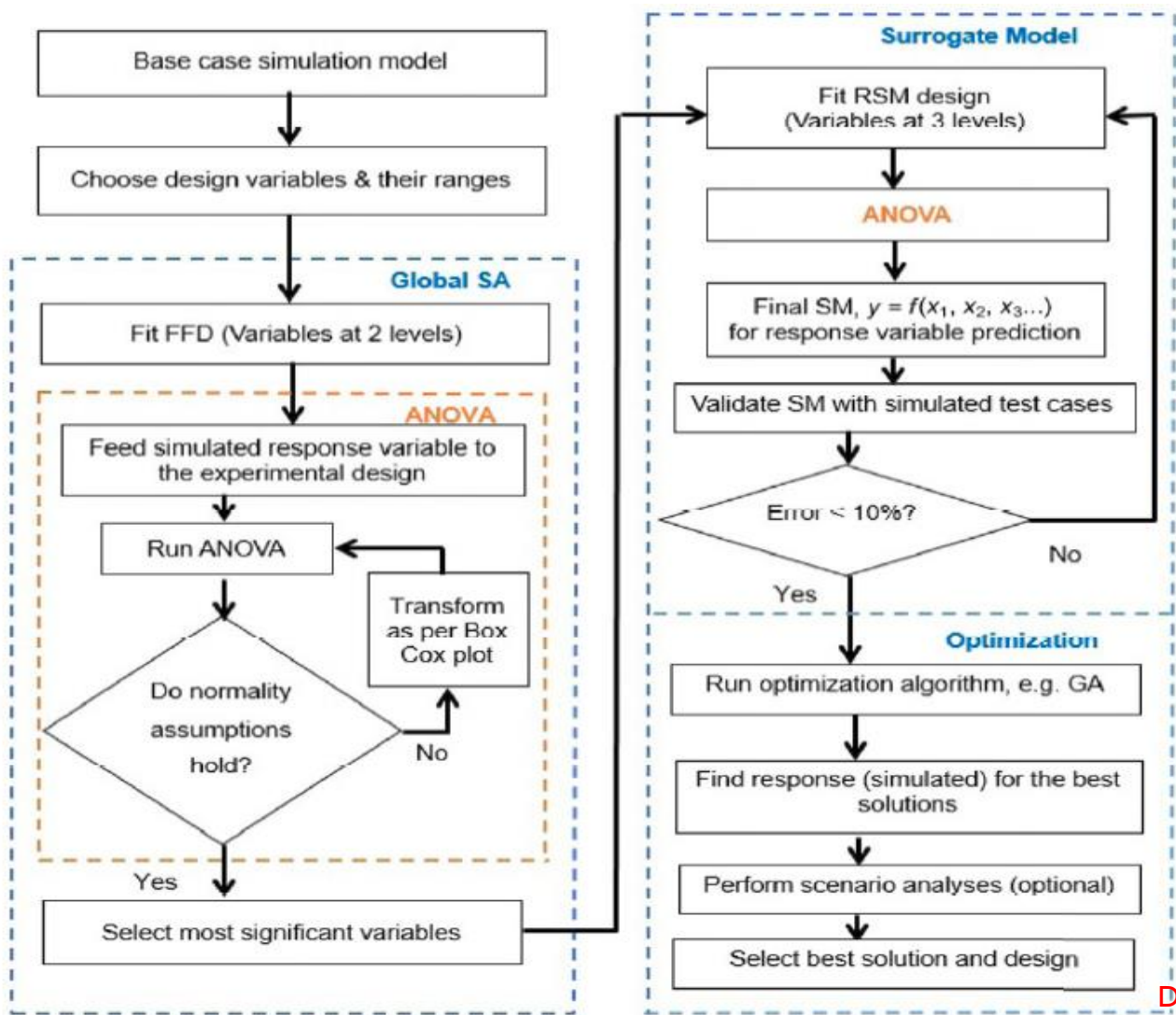
Validation

- Energy Plus used for validation
- Error less than 10%
- Used for 3-storeyed house in Delhi
- Team Shunya

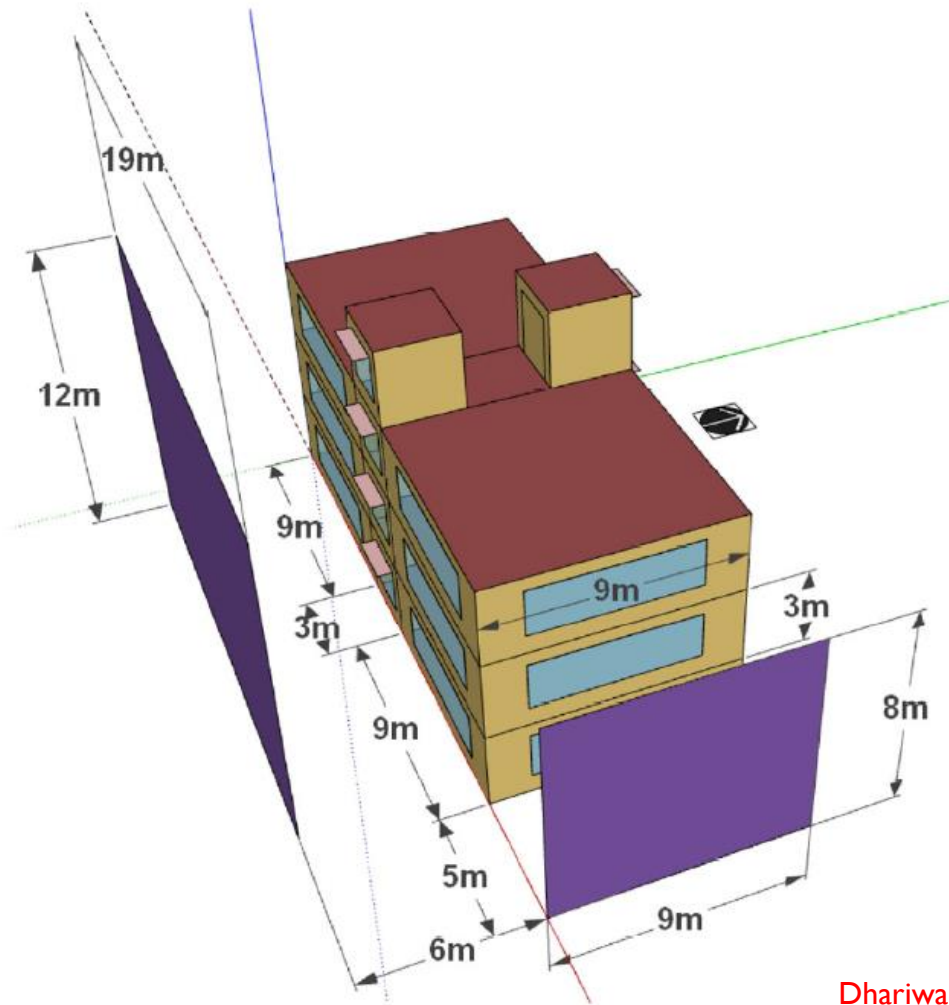
Usefulness

- Methodology- reduces computational time, enables optimisation





Dhariwal and Banerjee, Building Simulation, Springer Nature (2017)



| Design variable | Range |
|--|--|
| WWR | 15% to 75%, 15% to 50%, 15% to 50%, 15% to 40%* |
| Window SHGC | 0.2 to 0.8 for each facade* |
| Window VLT | 0.2 to 0.8 for each facade* |
| Overhang depth | 0 to 1 m for each facade* |
| Fin depth | 0 to 1 m for each facade* |
| Window U -value | 1 to 5.8 $W/(m^2 \cdot K)$ |
| Roof reflectivity (ρ) | 0.25 to 0.85 |
| Wall insulation thickness (t) | 0–0.1 m |
| Roof insulation thickness (t) | 0–0.2 m |
| Wall additional thermal mass thickness (t) | 0–0.1 m |
| Roof additional thermal mass thickness (t) | 0–0.2 m |

*North (N), east (E), west (W), south (S) facades respectively.

Dhariwal and Banerjee, Building Simulation, Springer Nature (2017)

$$\text{Zone 1 cooling capacity (in W)} = 634 \times CE_{\text{zone1}} + 749 \quad (9)$$

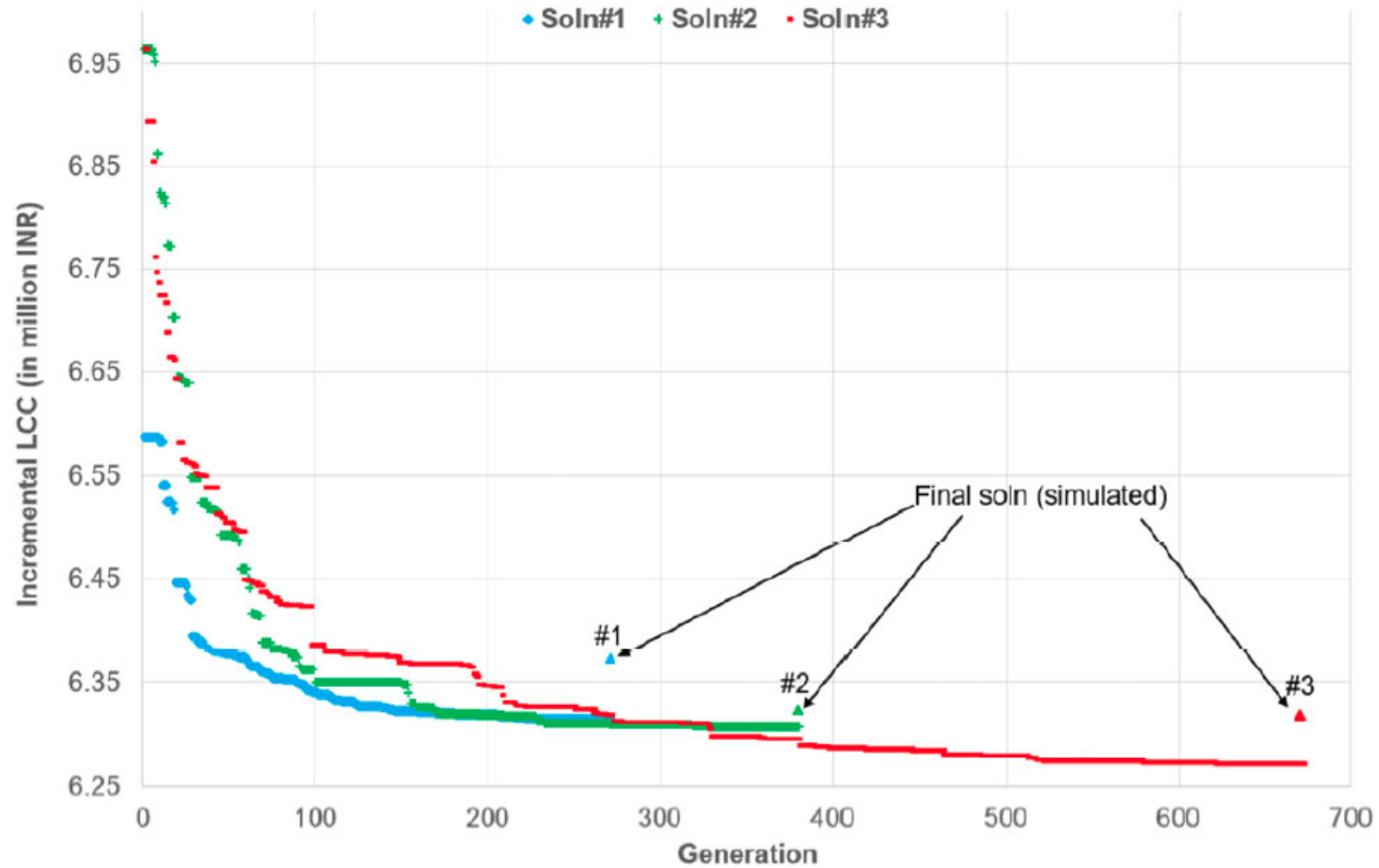
$$\text{Zone 2 cooling capacity (in W)} = 603 \times CE_{\text{zone2}} - 3264 \quad (10)$$

$$\text{Zone 3 cooling capacity (in W)} = 495 \times CE_{\text{zone3}} + 84 \quad (11)$$

$$\begin{aligned} \text{SHGC} = & -0.16 + 0.65 \times \text{VLT} + 0.13 \times (\text{Window } U\text{-value}) \\ & - 0.01 \times (\text{Window } U\text{-value})^2 \end{aligned} \quad (12)$$

$$\frac{\text{WWR-N} + \text{WWR-E} + \text{WWR-W} + \text{WWR-S}}{4} \geq 40\% \quad (13)$$

Dhariwal and Banerjee, Building Simulation, Springer Nature (2017)



Dhariwal and Banerjee, Building Simulation, Springer Nature (2017)

Context

- Roof top PV potential in city
- Solar Mission – focus on land based large plants

Nature of Model

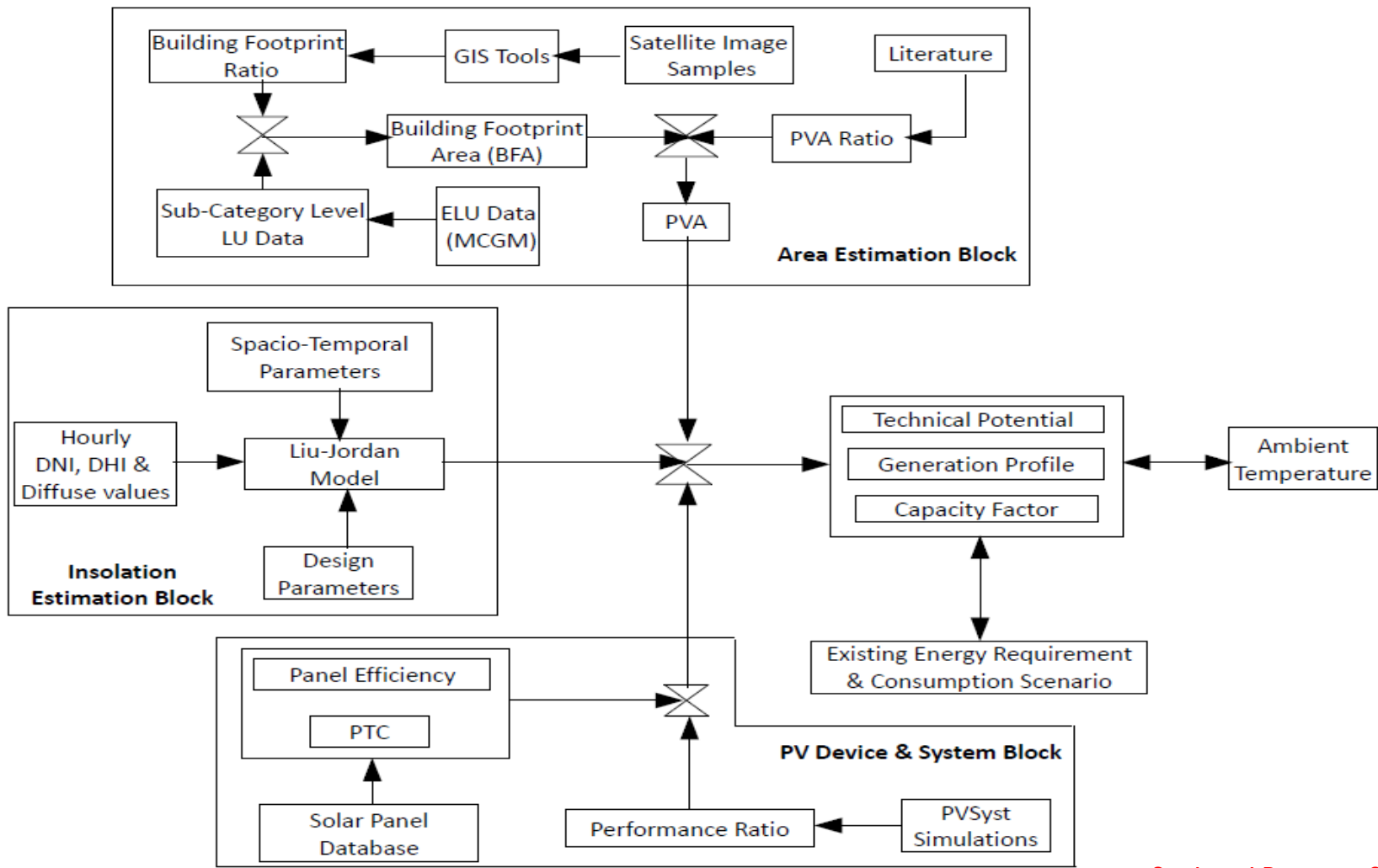
- Area Estimation,
- Insolation estimation,
- PV Device and System(PV Syst)
- Results and Analysis

Validation

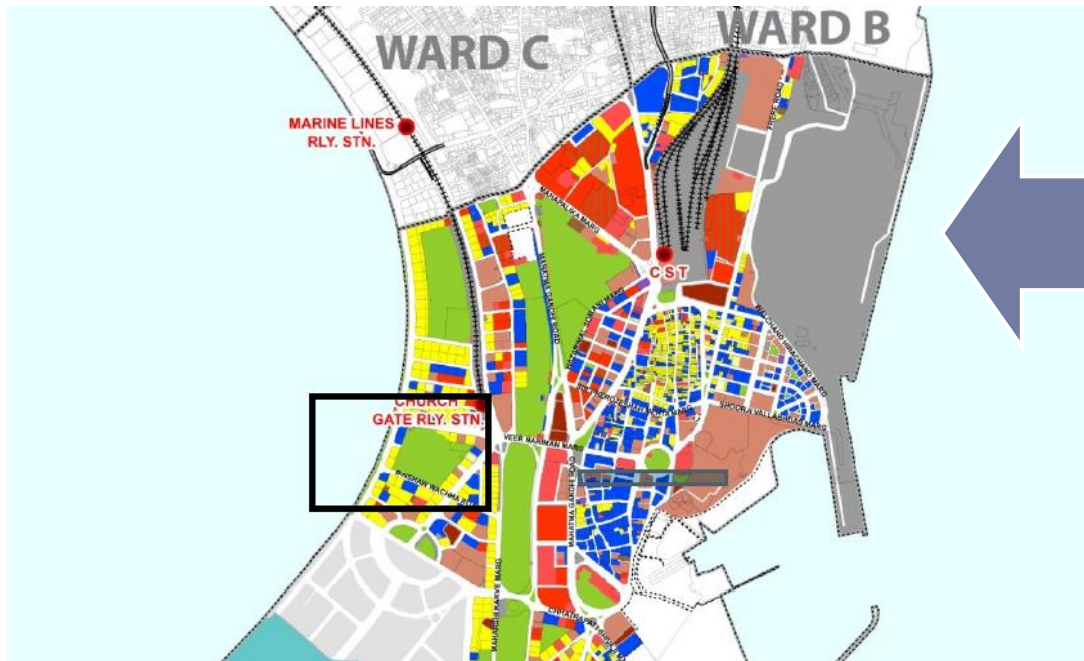
- Samples for land use types
- Comparison of PVA with literature

Usefulness

- Transparent Methodology
- Load profiles and PV generation-typical days – different months
- Applied for other cities – Bengaluru



Singh and Banerjee, Solar Energy, 115, 2015



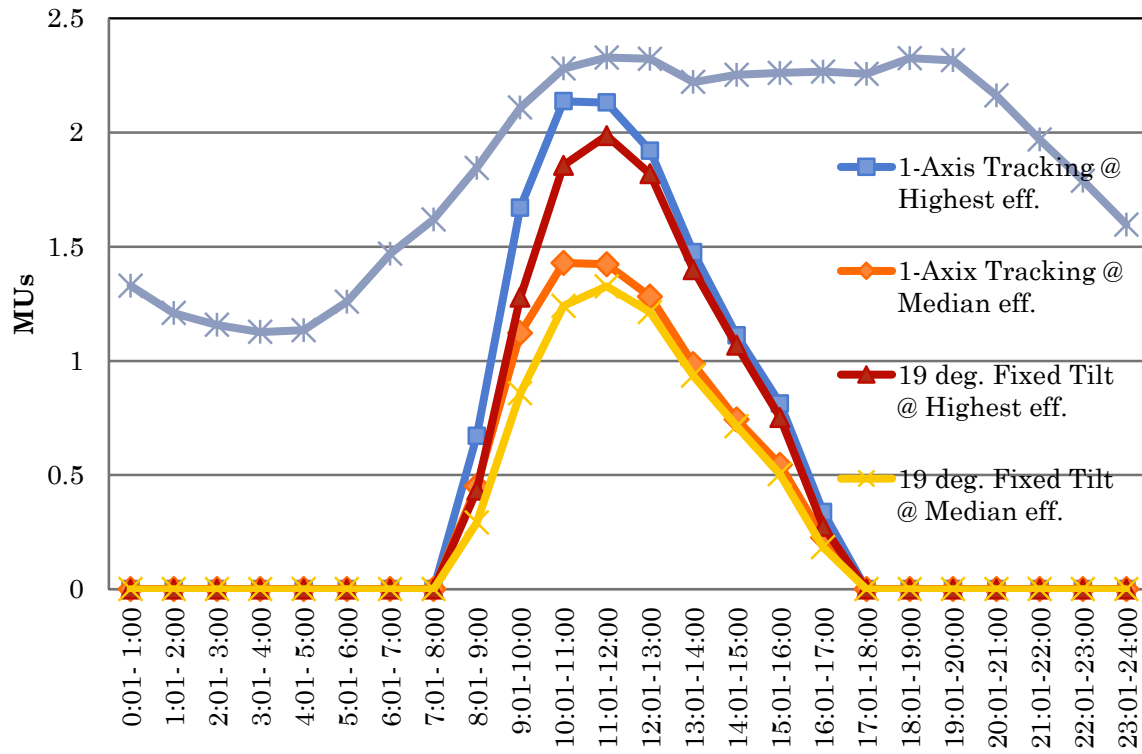
ELU map of Ward A of MCGM

Corresponding
Satellite Imagery for
the area from Google
Earth

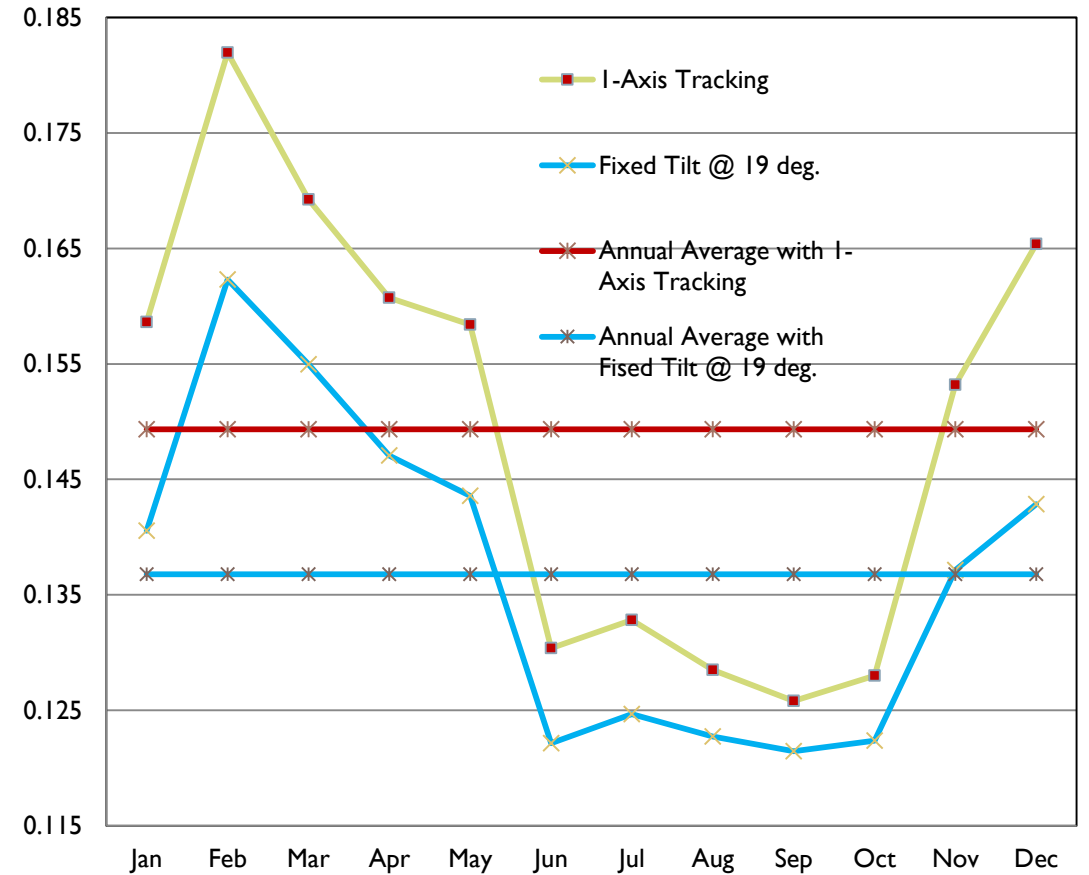


Singh and Banerjee, Solar Energy, 115, 2015

Jan, 2014 Typical Load Profile vs PV Generation



Capacity Factor for Mumbai



Context

- Conventionally Hydro Thermal scheduling problem
- Increased PV and renewable share
- Impact on grid ?

Nature of Model

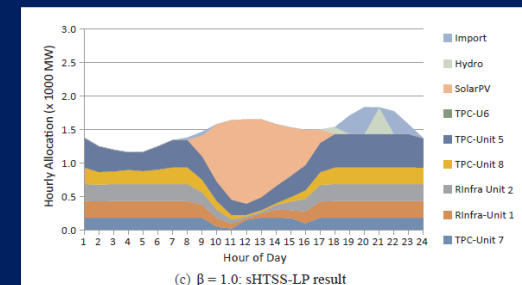
- PV must run
- Two stage formulation using dynamic programming and linear programming

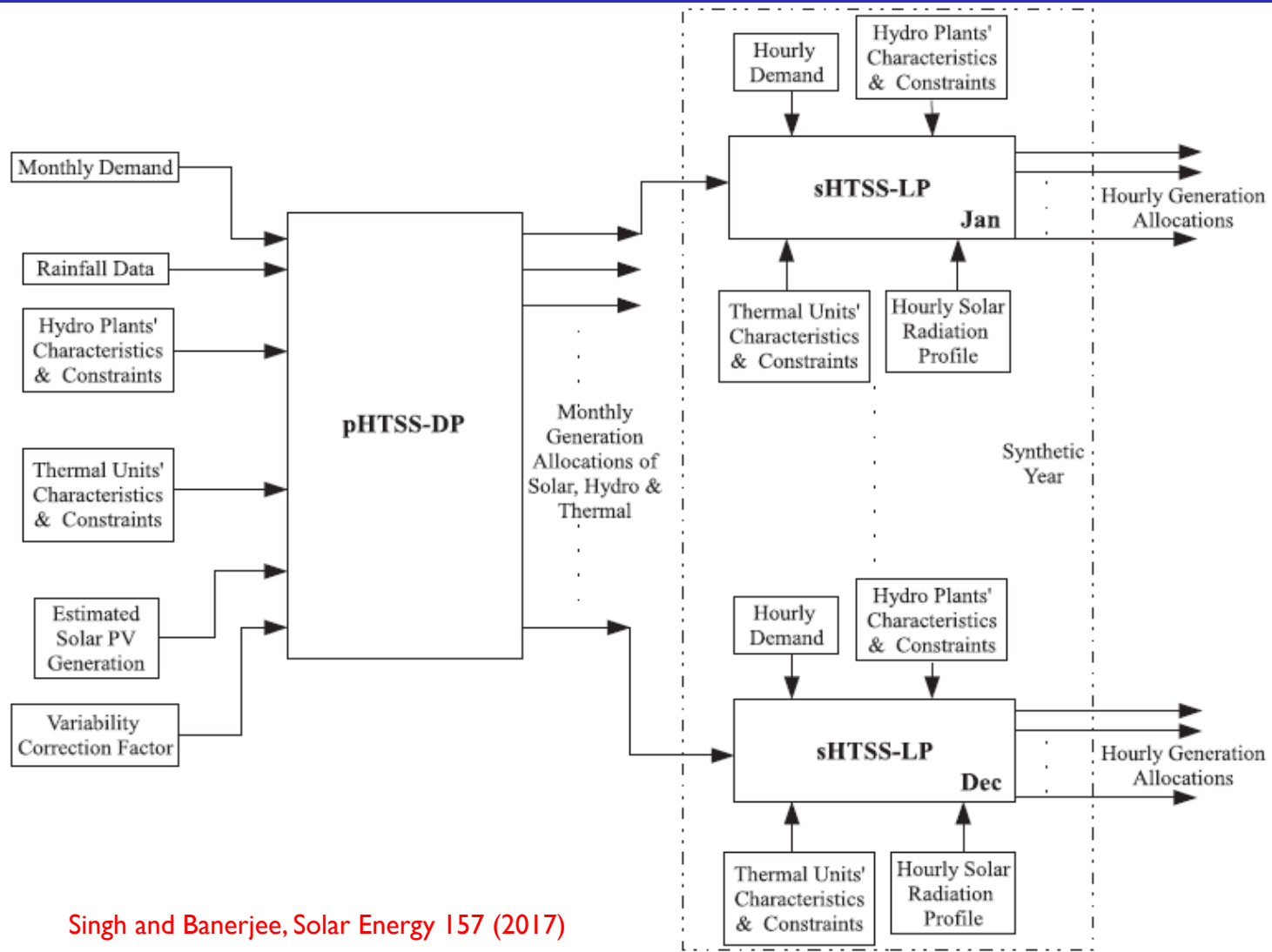
Validation

- Proposed HTSS algorithm compared with stochastic algorithm in literature
- Savings of 8.2%
- Case study of Mumbai – actual data

Usefulness

Impact on load profile for different PV penetration





Singh and Banerjee, Solar Energy 157 (2017)

Minimize $h_h^{reg}, h_h^{hol}, \theta_{h,i}^{reg}, \theta_{h,i}^{hol}$

$$\left[\begin{array}{l} d_{reg} \times \left(\sum_{i=1}^n (c_i \times \sum_{h=1}^{24} \theta_{h,i}^{reg}) + c_{imp} \times \sum_{h=1}^{24} (D_h^{reg} - (h_h^{reg} + \beta \cdot E_h^{solar} \cdot CF_{var} + \sum_{i=1}^n \theta_{h,i}^{reg})) \right) + \\ d_{hol} \times \left(\sum_{i=1}^n (c_i \times \sum_{h=1}^{24} \theta_{h,i}^{hol}) + c_{imp} \times \sum_{h=1}^{24} (D_h^{hol} - (h_h^{hol} + \beta \cdot E_h^{solar} \cdot CF_{var} + \sum_{i=1}^n \theta_{h,i}^{hol})) \right) \end{array} \right]$$

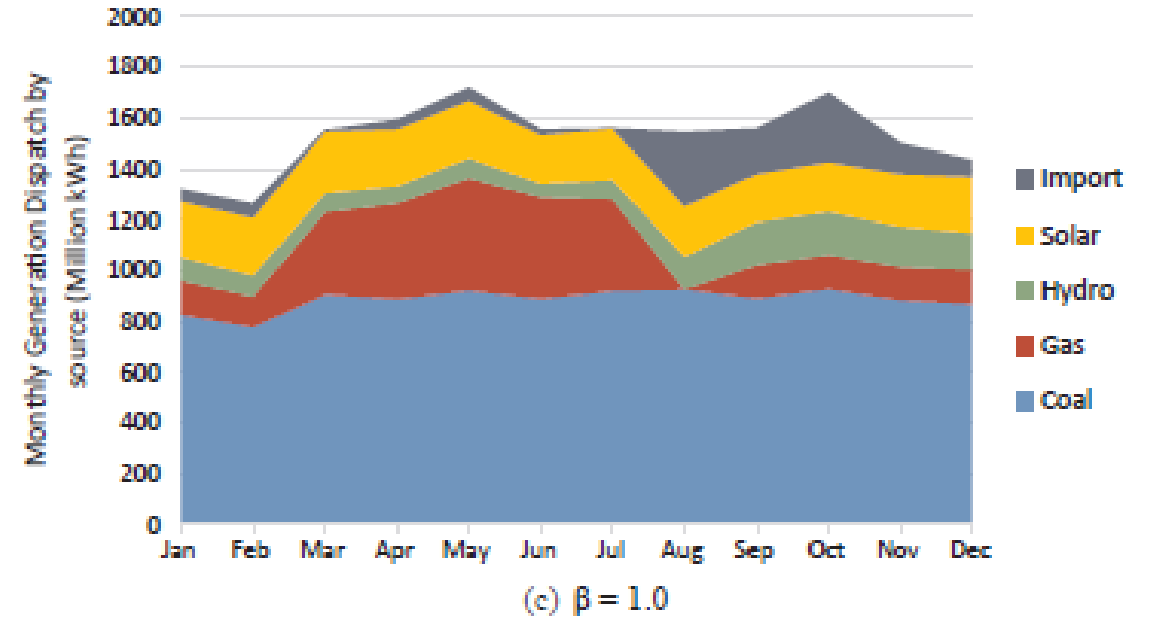
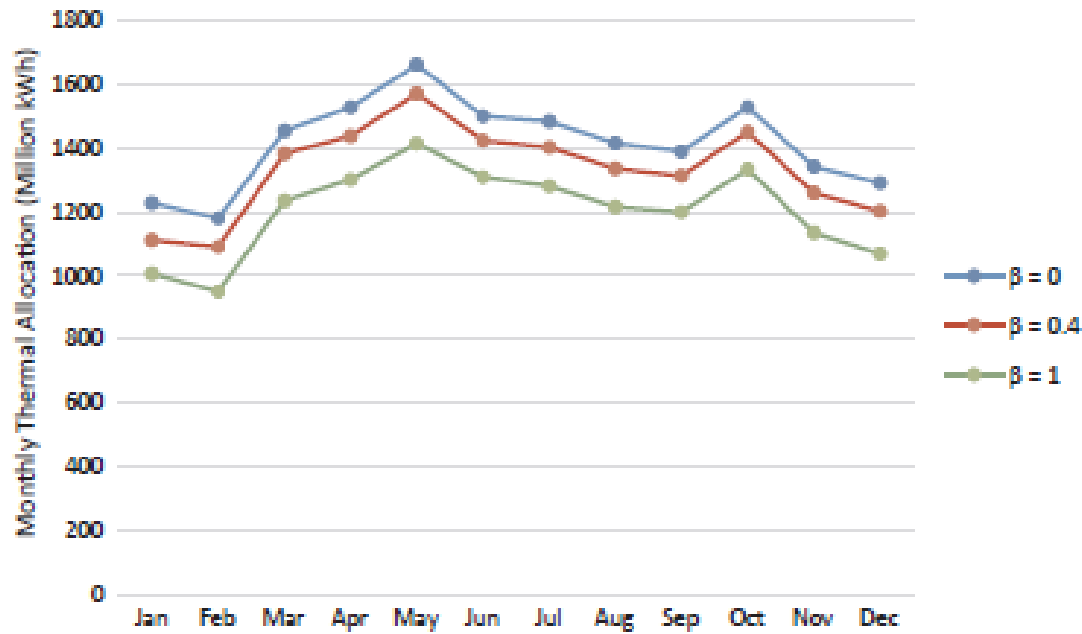
Singh and Banerjee, Solar Energy 157 (2017)

Priority List of Fossil-Fuel Based Generating Units for Mumbai.

| S.No. | Unit name | Annual Average Energy Cost per Unit (INR/kWh) | Rated capacity (MW) | Fuel | Ramp rate (MW/min) |
|-------|---------------|---|---------------------|------|--------------------|
| 1 | TPC-Unit 7 | 2.81 | 180 | Gas | 8–12 |
| 2 | RInfra Unit 1 | 2.88 | 250 | Coal | 1–3 |
| 3 | RInfra Unit 2 | 2.88 | 250 | Coal | 1–3 |
| 4 | TPC-Unit 8 | 4.06 | 250 | Coal | 1–3 |
| 5 | TPC-Unit 5 | 4.15 | 500 | Coal | 1–3 |
| 6 | TPC-Unit 6 | 10.76 | 500 | Gas | 8–12 |
| 7 | Grid Import | Variable | Varies | – | – |
| 8 | TPC-Unit 4 | 14.18 | 150 | Oil | |

| Name of Hydro Unit | Khopoli | Bhivpuri | Bhira |
|---------------------------------------|----------------|----------------|----------------|
| Location | 18.8°N, 73.4°E | 18.9°N, 73.5°E | 18.5°N, 73.4°E |
| Name of Main Reservoir | Walwan Lake | Andhra Lake | Mulshi Lake |
| Design Capacity (MWe) | 72 | 75 | 300 |
| Annual Design Energy (Million kWh) | 293 | 339 | 1217 |
| Capacity Factor (%) | 46.45 | 51.6 | 46.31 |
| Full Reservoir Level (m) | 635.2 | 668 | 607.1 |
| Minimum Draw Down Level (m) | 619.4 | 646 | 590.1 |
| Gross Storage Capacity (MCM) | 72.5 | 363.7 | 747 |
| V_i^p (m) (as in Eq. (6)) | 627.3 | 657 | 598.6 |

Singh and Banerjee, Solar Energy 157 (2017)



| Case | Annual Generation-Cost Estimate from sHTSS-LP (Million INR) | % change from $\beta = 0$ case |
|---------------|---|--------------------------------|
| $\beta = 0$ | 97,530 | 0 |
| $\beta = 0.2$ | 91,449 | - 6.2 |
| $\beta = 0.4$ | 85,469 | - 12.4 |
| $\beta = 0.6$ | 79,672 | - 18.3 |
| $\beta = 0.8$ | 74,313 | - 23.8 |
| $\beta = 1.0$ | 69,610 | - 28.6 |

Singh and Banerjee, Solar Energy 157 (2017)

Context

- Microgrid scenarios for India-Urban, rural, industrial and isolated
- Model based analysis of energy for a PV battery system

Nature of Model

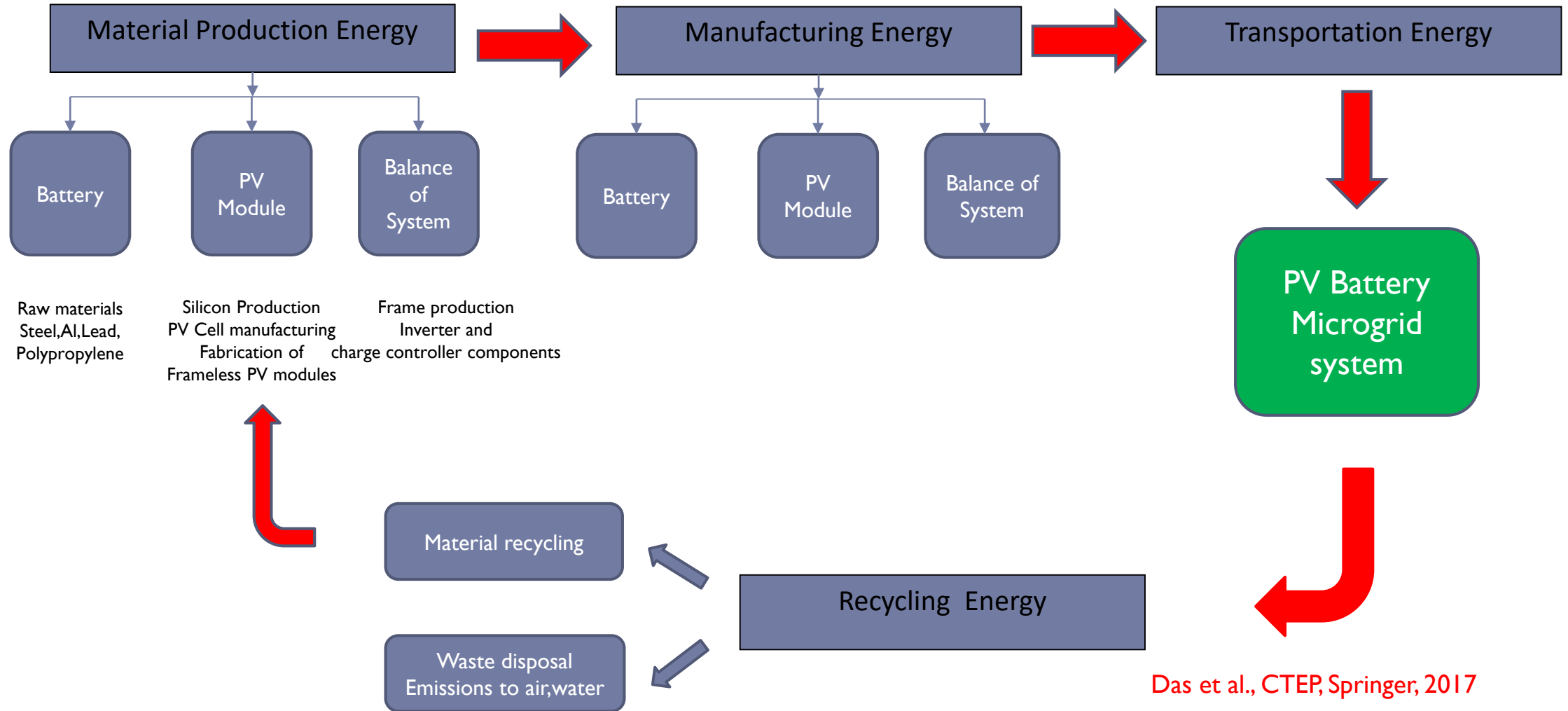
- Direct energy requirements for the PV battery system for the designed microgrid

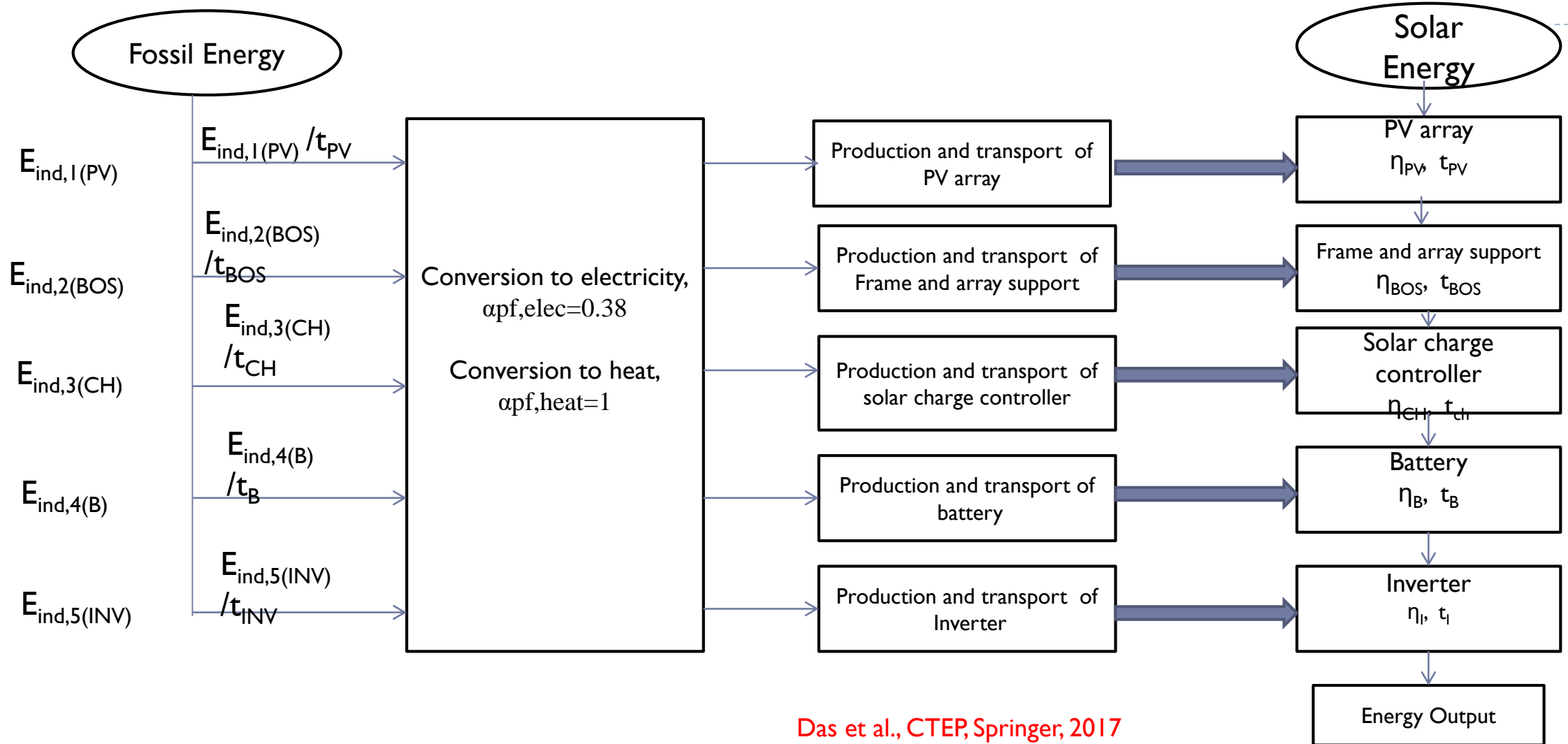
Validation

- Actual system components and their service life

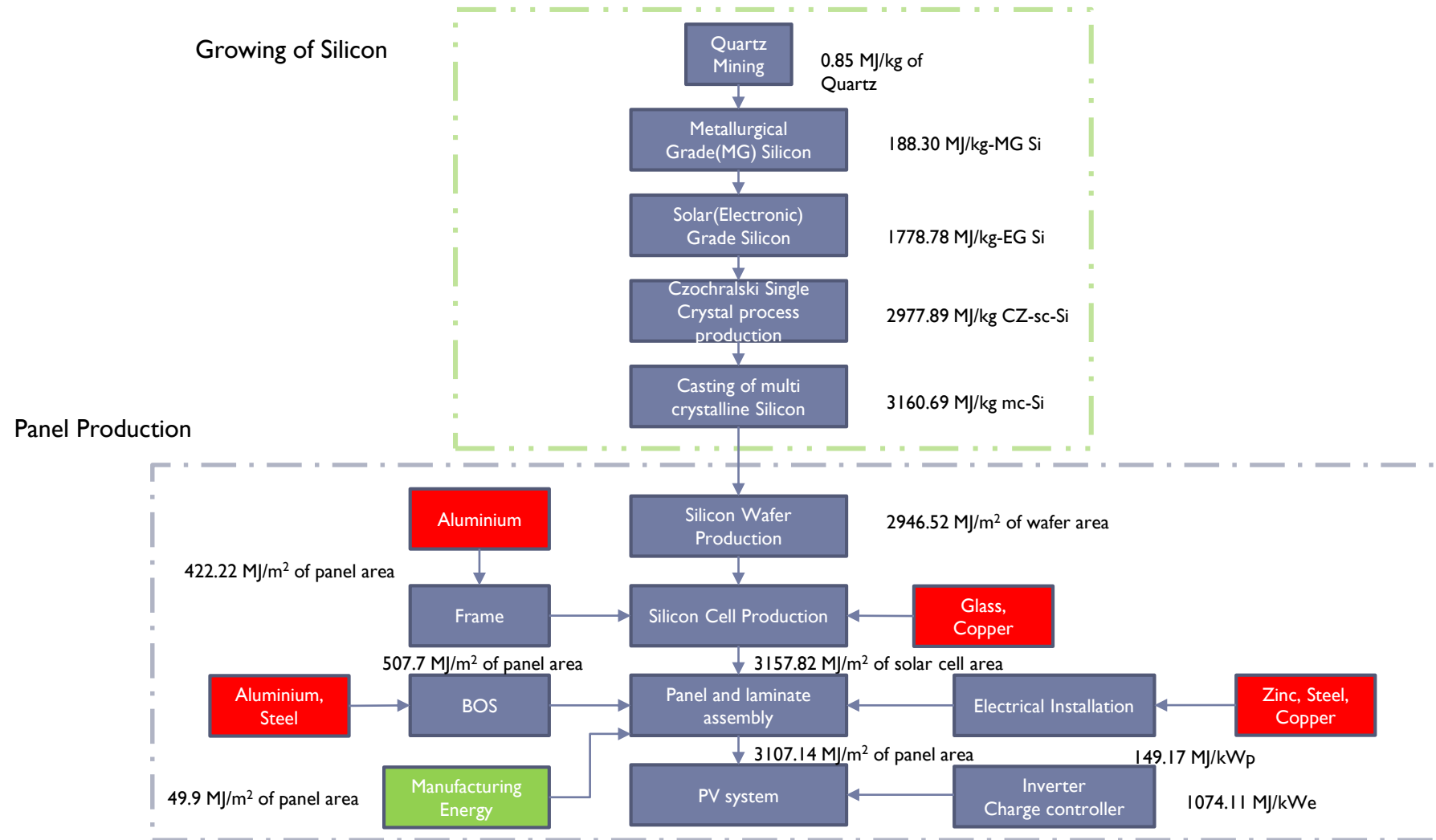
Usefulness

- Sustainability analysis
- Cradle to Gate
- Pay back time and net energy ratio
- Comparison of different options



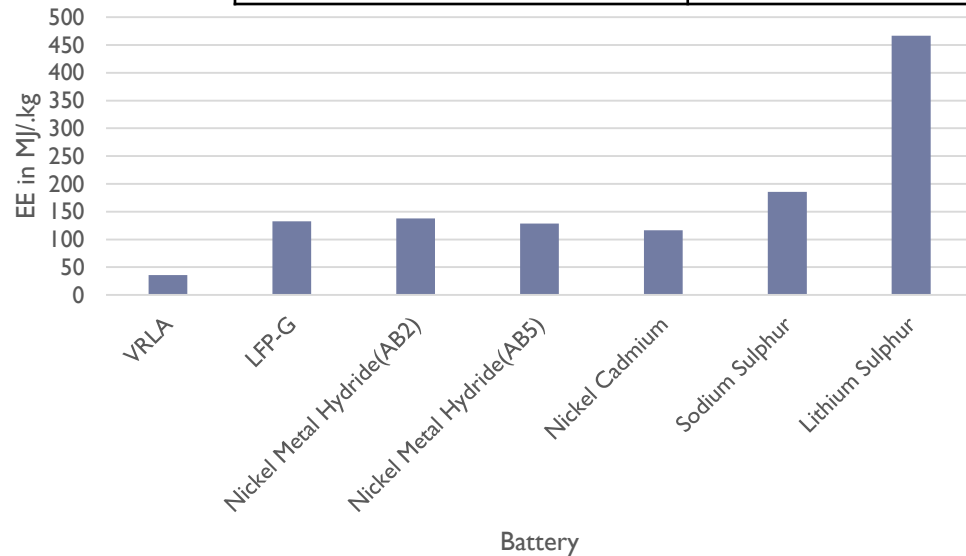


Das et al., CTEP, Springer, 2017

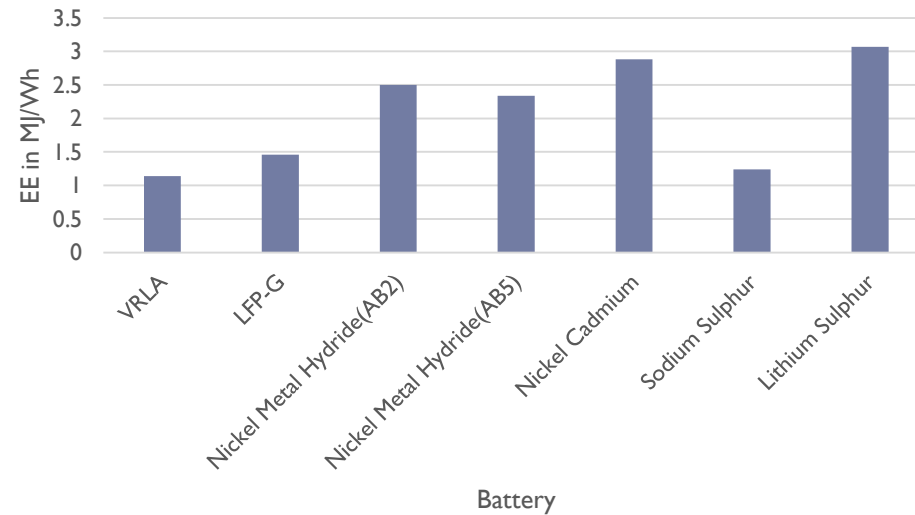


Das et al., CTEP, Springer, 2017

| Component | Material Production Energy | Manufacturing Energy of PV panel, frame and BOS | Transportation Energy |
|--------------------------------------|--|---|---|
| PV system (multicrystalline-Silicon) | 3107.14 MJ _{pf} /m ² of sensing area | 49.9 MJ _{pf} /m ² of panel area | 0.34 MJ _{pf} /kg of panel weight |
| Frame | 422.22 MJ _{pf} /m ² of frame area | | |
| Balance of System(BOS) | 507.7 MJ _{pf} /m ² of panel area | | |
| Electrical Installation | 149.17 MJ _{pf} /kW _p | | |
| Charge Controller | 1074.11 MJ _{pf} /kW _e | | |
| Inverter | 1074.11 MJ _{pf} /kW _e | | |



Embodied Energy per unit mass for different batteries



Embodied Energy per unit storage capacity for different batteries

Das et al., CTEP, Springer, 2017

| PV+battery configuration | Energy Pay Back Time(years) | Net Energy Ratio | Energy Pay back (% of cycle life) | Emission Factor (kgCO ₂ /kWh of output) |
|--------------------------|-----------------------------|------------------|-----------------------------------|--|
| VRLA | 2.75 | 4.63 | 20.10 | 0.26 |
| LFP-G | 2.62 | 6.55 | 6.88 | 0.14 |
| NiMH(AB ₂) | 3.44 | 2.52 | 36.92 | 0.42 |
| NiMH(AB ₅) | 3.4 | 2.63 | 34.12 | 0.42 |
| NaS | 4.3 | 3.9 | 7.58 | 0.67 |
| NiCd | 2.69 | 4.2 | 64.12 | 0.091 |
| LiS | 3.38 | 2.8 | 67.56 | 0.31 |

$$EPBT = \frac{\text{LifeTime Energy Input to the system}}{\text{Annual Energy Output}}$$

$$NER = \frac{\text{Annualised energy output}}{\text{Annualised energy input}}$$

Das et al., CTEP, Springer, 2017

Context

- Modelling framework to link Energy, Economy, Environment – not adequate
- Impact of structure of economy, climate constraint

Nature of Model

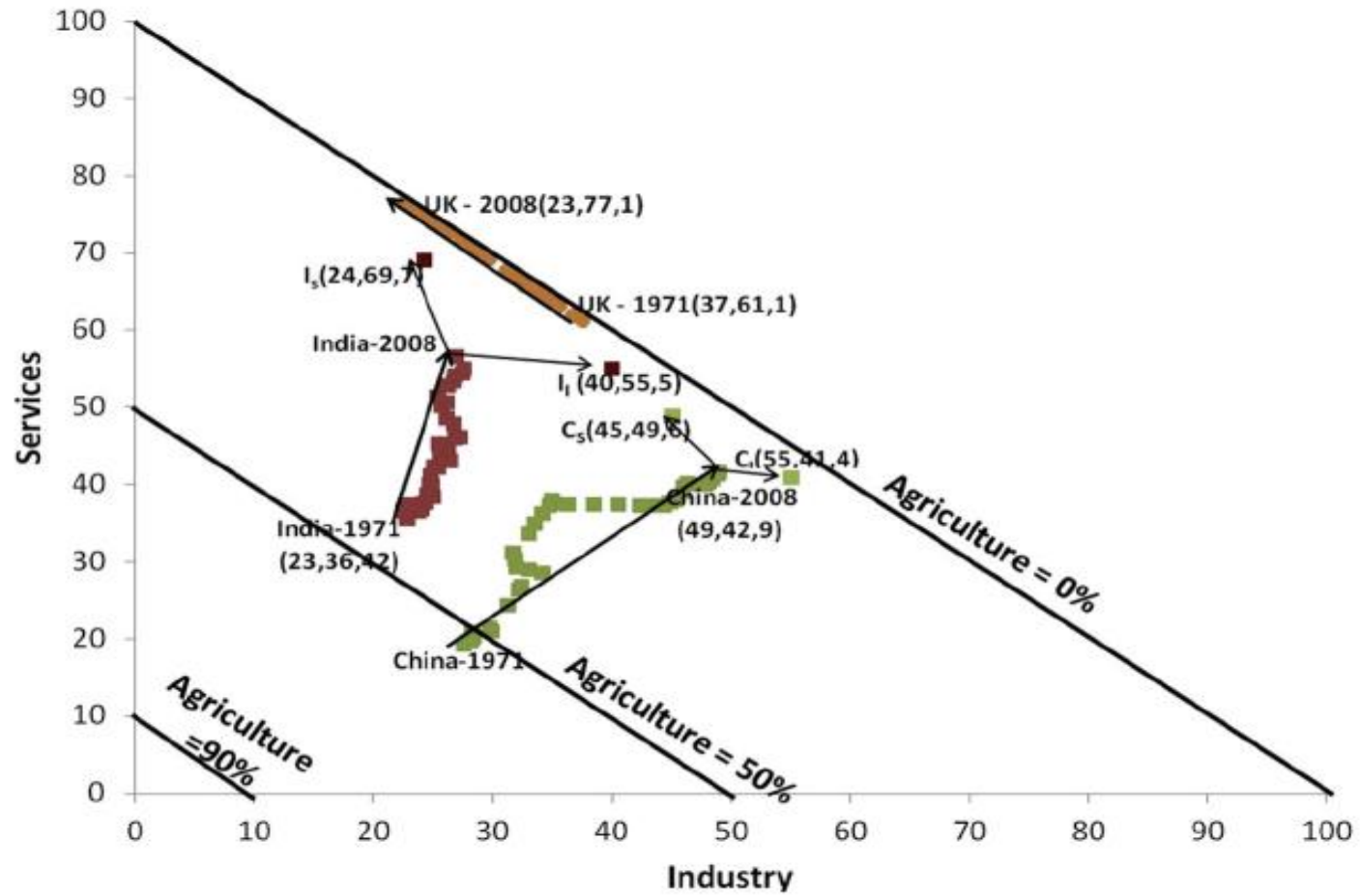
- Decomposition Analysis
- Input-Output Analysis
- Sectoral Optimisation model (Example for Power sector)
- Integrated Modelling Framework

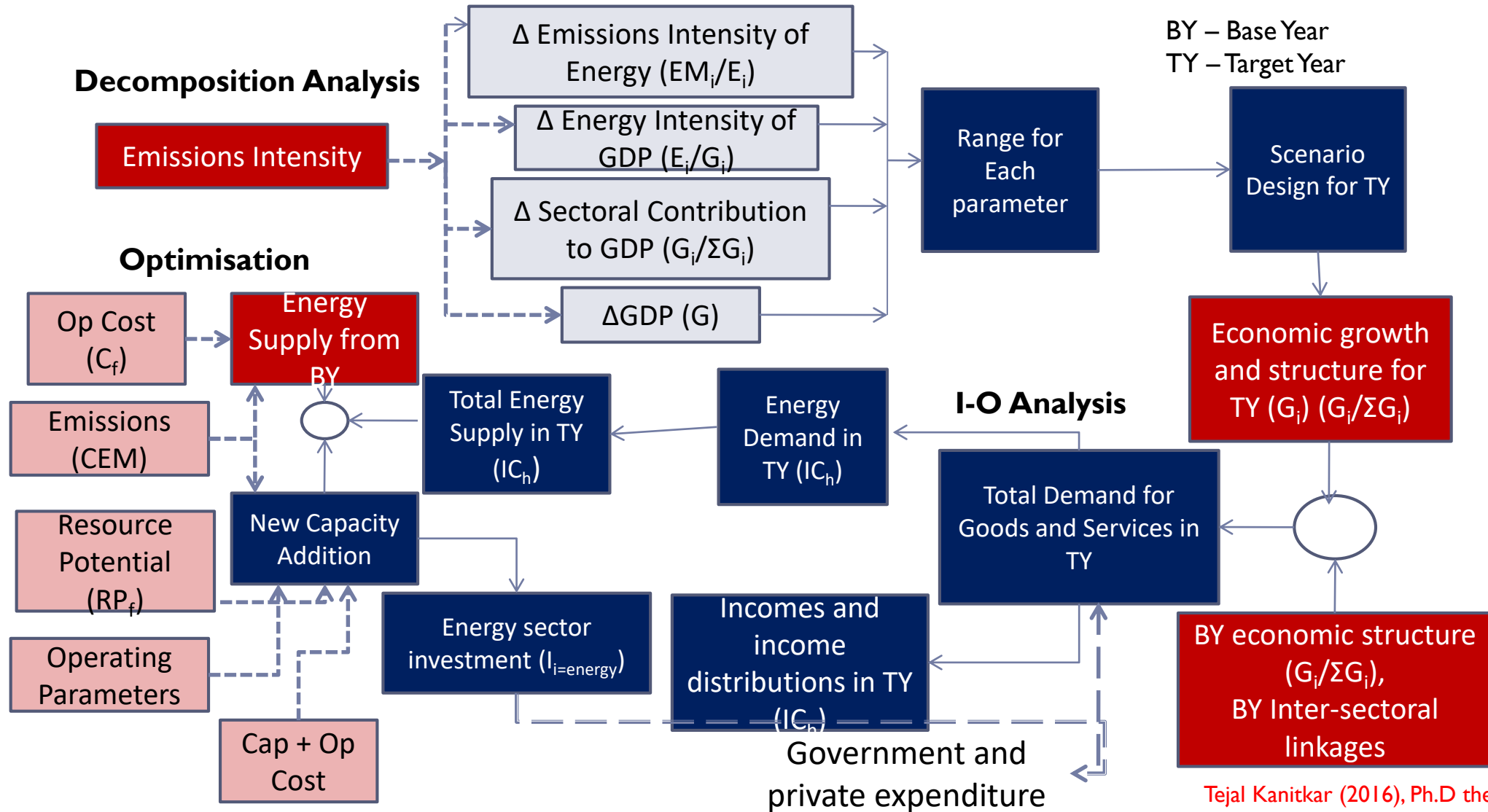
Validation

- Individual models compared with literature
- Comparison of sectoral shares globally
- Overall check of reasonableness

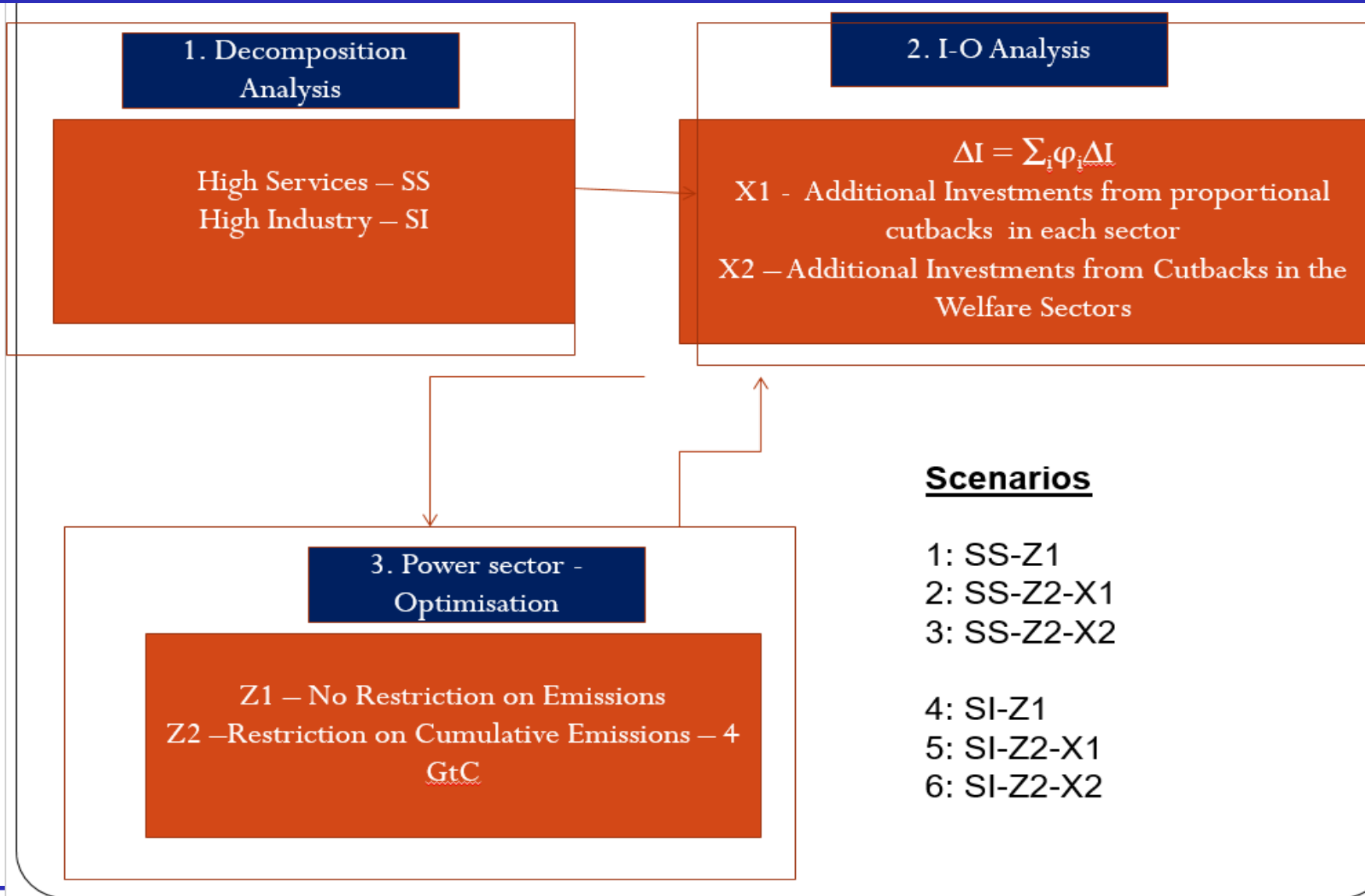
Usefulness

- Impact of different development pathways – high industry, high services
- Impact on Inequality – Gini coefficient
- Impact on overall growth rate





Tejal Kanitkar (2016), Ph.D thesis, IIT Bombay



Scenarios

- 1: SS-Z1
- 2: SS-Z2-X1
- 3: SS-Z2-X2

- 4: SI-Z1
- 5: SI-Z2-X1
- 6: SI-Z2-X2

| | Base year | 1-SS-Z₁ | 2-SS-Z₂-X₁ |
|---|------------------|---------------------------|---|
| GDP (Billion \$) | 415 | 2415 | 2354 |
| GDP Growth p.a. | - | 7.08% | 6.98% |
| Per Capita Income of all Households (\$/year/person) | 338 | 1354 | 1308 |
| Per Capita Income of all Households class RH1(\$/year/person) | 62 | 185 | 154 |
| Per Capita Income of all Households RH4 (\$/year/person) | 354 | 1662 | 1600 |
| Per Capita Income of all Households UH1(\$/year/person) | 77 | 108 | 92 |
| Per Capita Income of all Households UH3 (\$/year/person) | 323 | 1231 | 1185 |
| GINI Coefficient | 0.497 | 0.531 | 0.536 |

- Models – representation of reality
- Improved decision making - component, system design, future sustainable routes, assess impacts, what if?
- Value judgements- trade-offs between criteria-Optimising/
Satisficing
- Blend of models/ prototypes – develop future sustainable energy systems
- Energy Systems in transition – need new modelling approaches



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Thank you

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