

Design Considerations for AI/ML, IoT & SaaS based Industrial Decarbonization

The Fifth Elephant (Aug 2023)

Narayanan Subramaniam
Founder & CTO VaNaShRi Technology Consulting
(<https://vanashri.com>)
LinkedIn: <https://www.linkedin.com/in/cnsubramaniam>

Image Credits: Various sources on the Internet

Copyright ©2023 VaNaShRi Technology Consulting



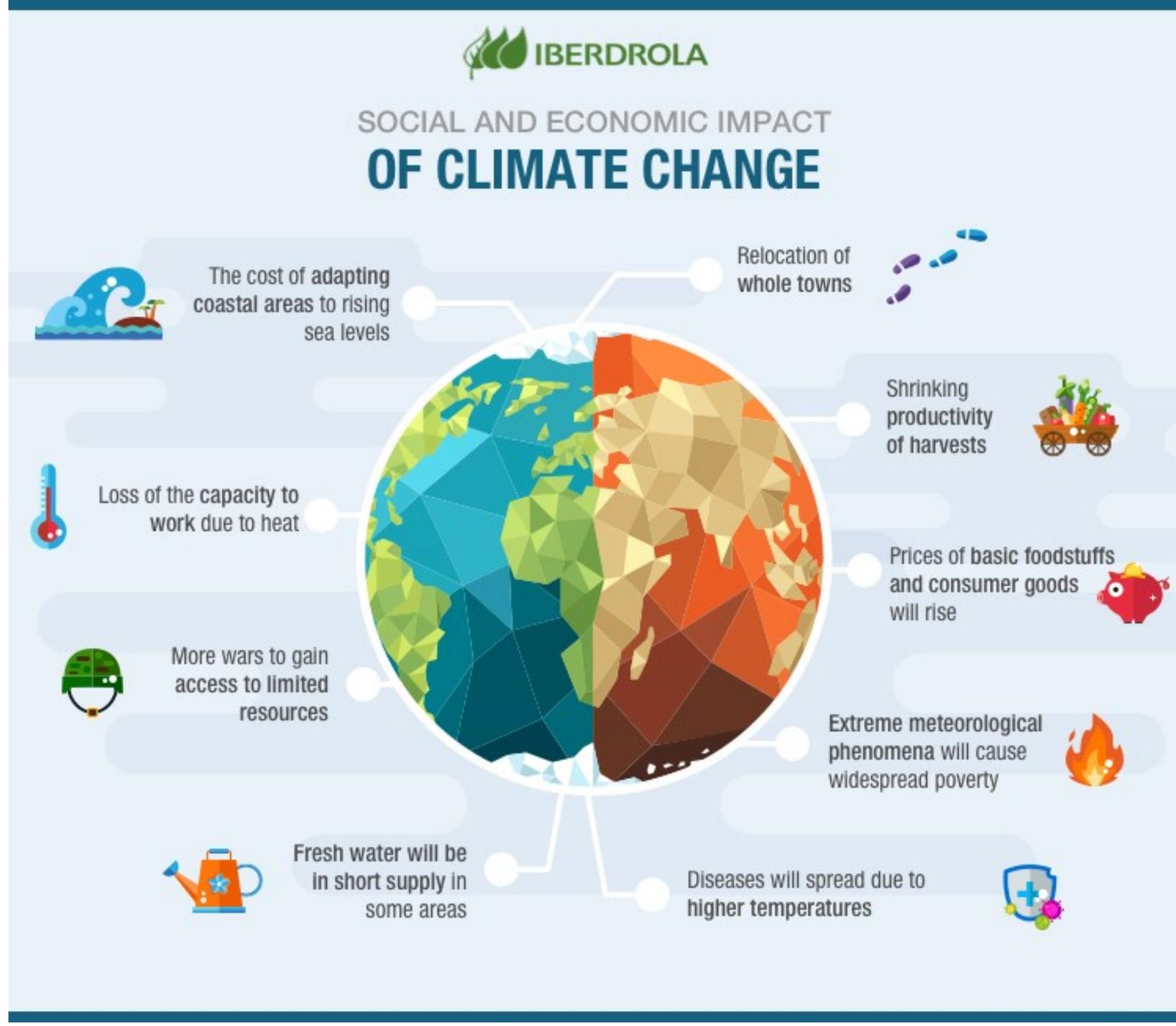
Climate Change & Global Warming



Consequences of Inaction →

Image Credits: Various sources on the Internet

Copyright ©2023 VaNaShRi Technology Consulting



GHGs causing
Global
Warming:
CO₂, CH₄, ...

4X-8X
Emission Cuts
Needed to
stave off an
Apocalypse →

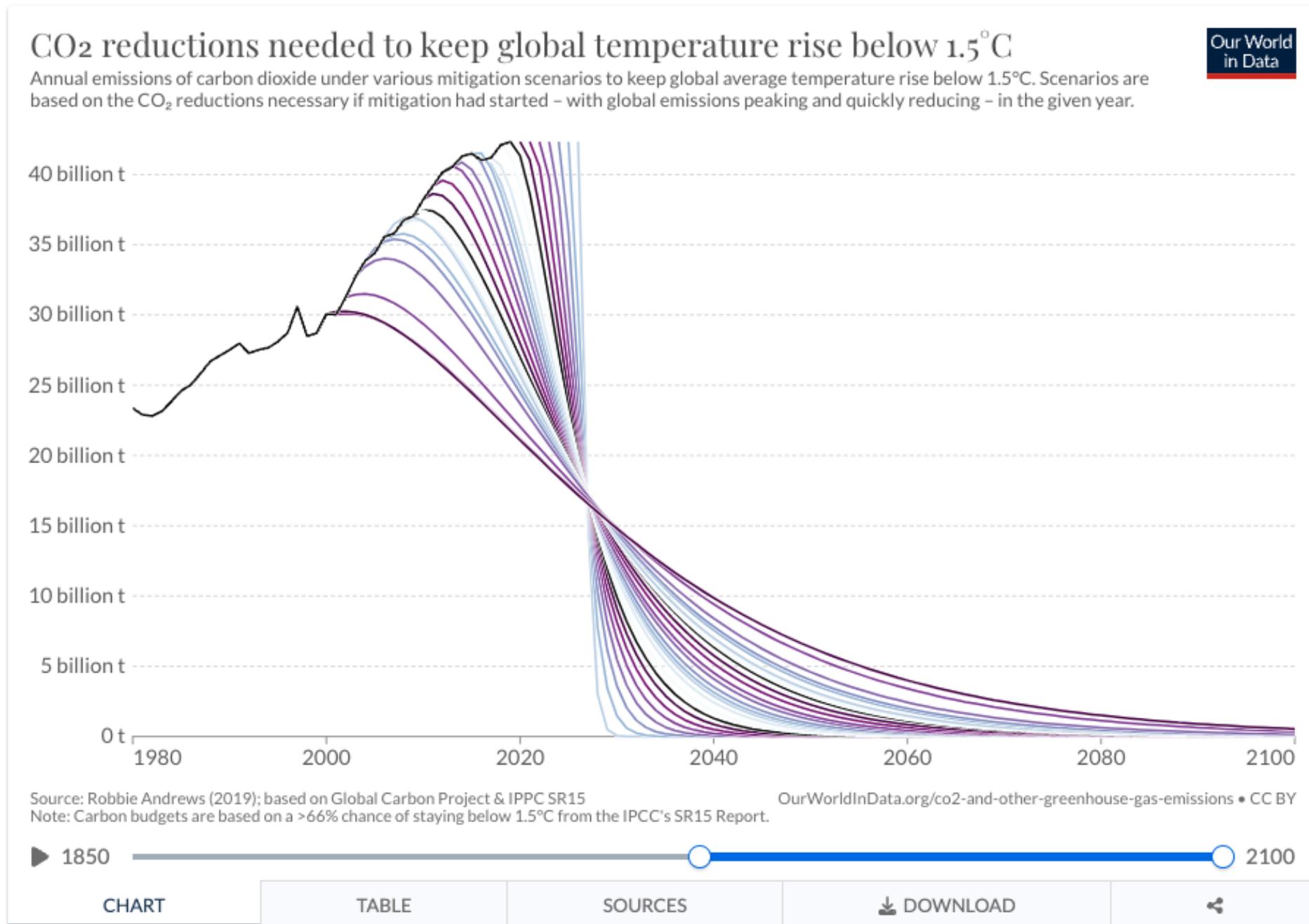


Image Credits: Various sources on the Internet

Copyright ©2023 VaNaShRi Technology Consulting

Global greenhouse gas emissions by sector

This is shown for the year 2016 – global greenhouse gas emissions were 49.4 billion tonnes CO₂eq.

GHG Sources:
Steel, Cement, Oil,
Chemicals, Mining
add up to 40+% →

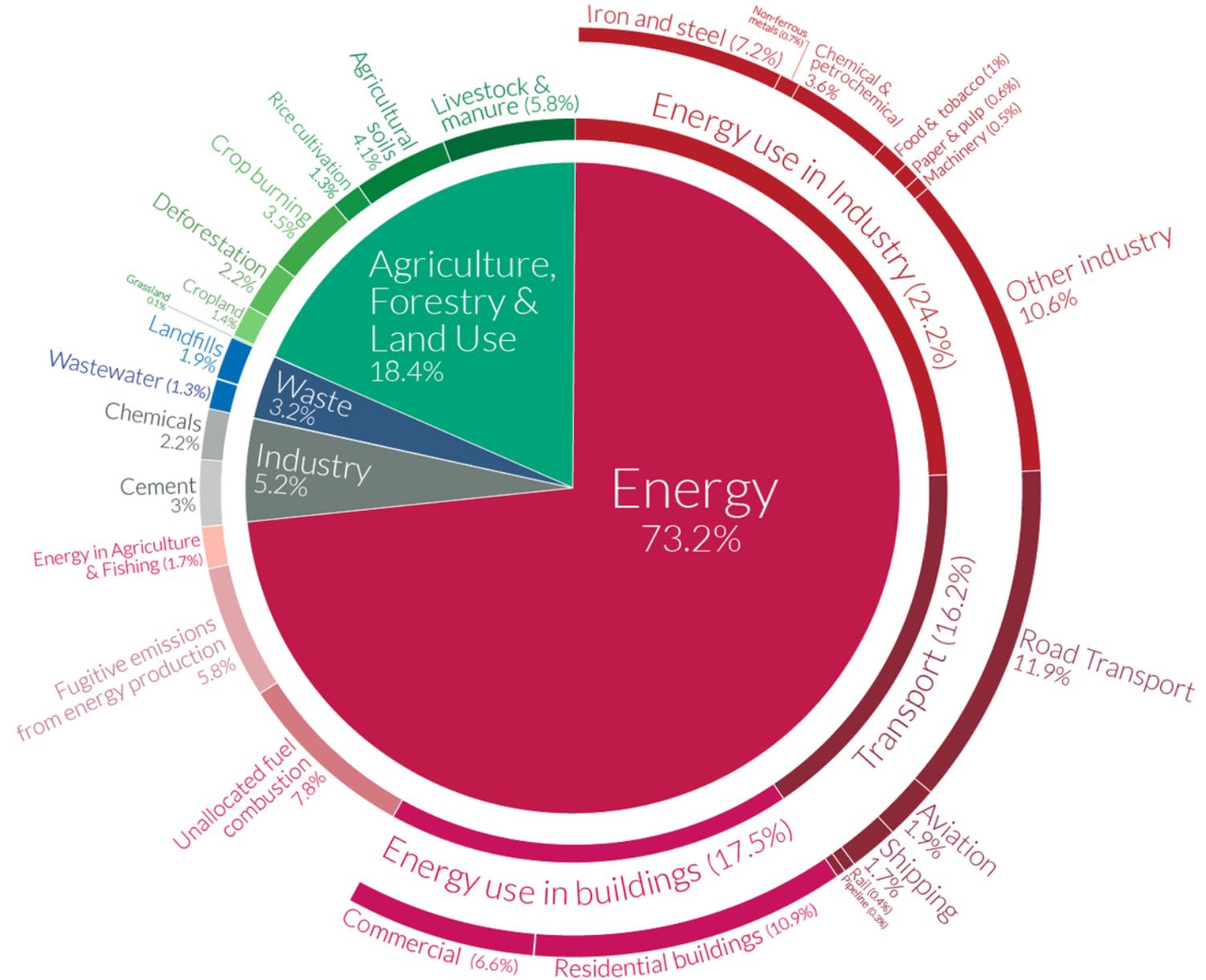
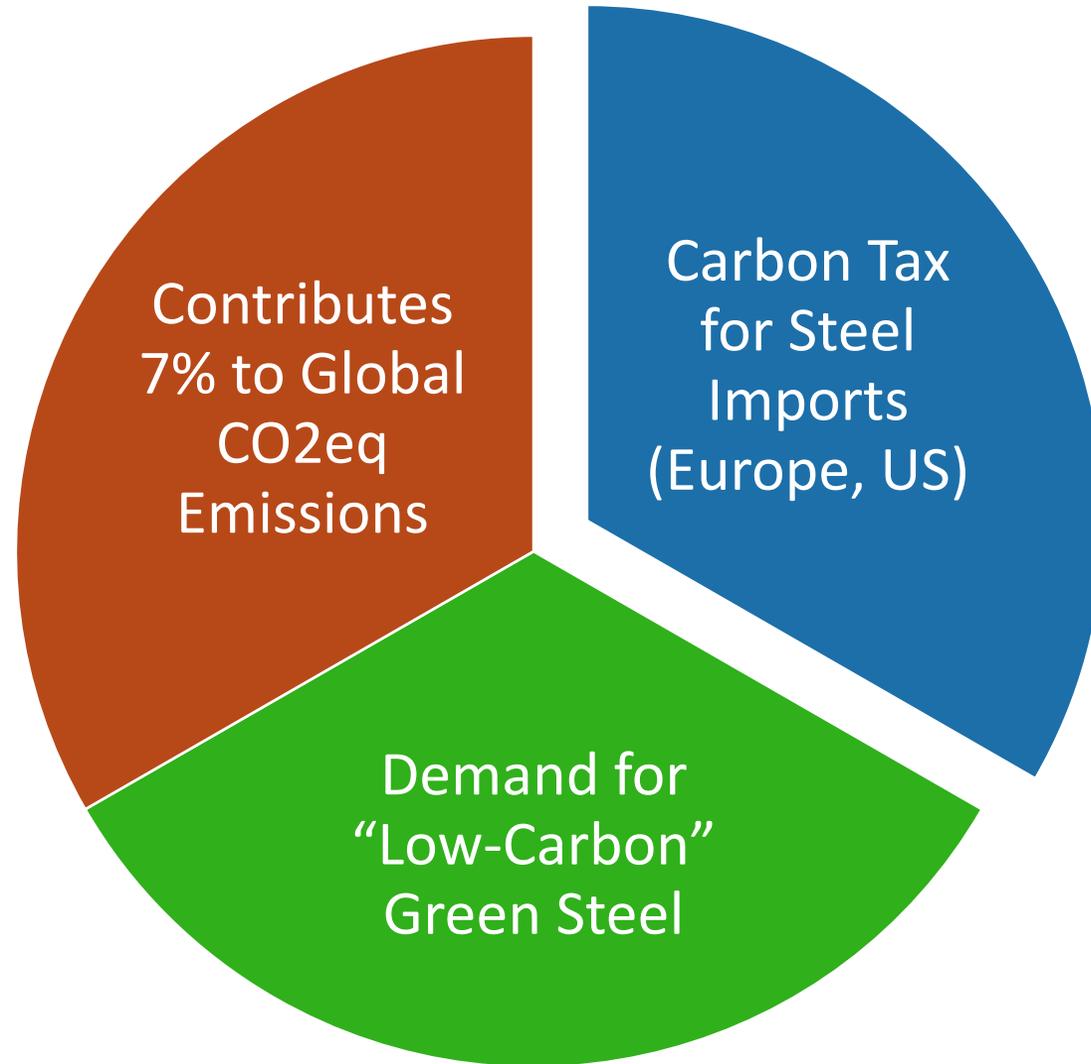


Image Credits: Various sources on the Internet

Opportunity: Reduce Emissions in Steel Production



Sintering process contributes 17% of GHGs in Steel Production

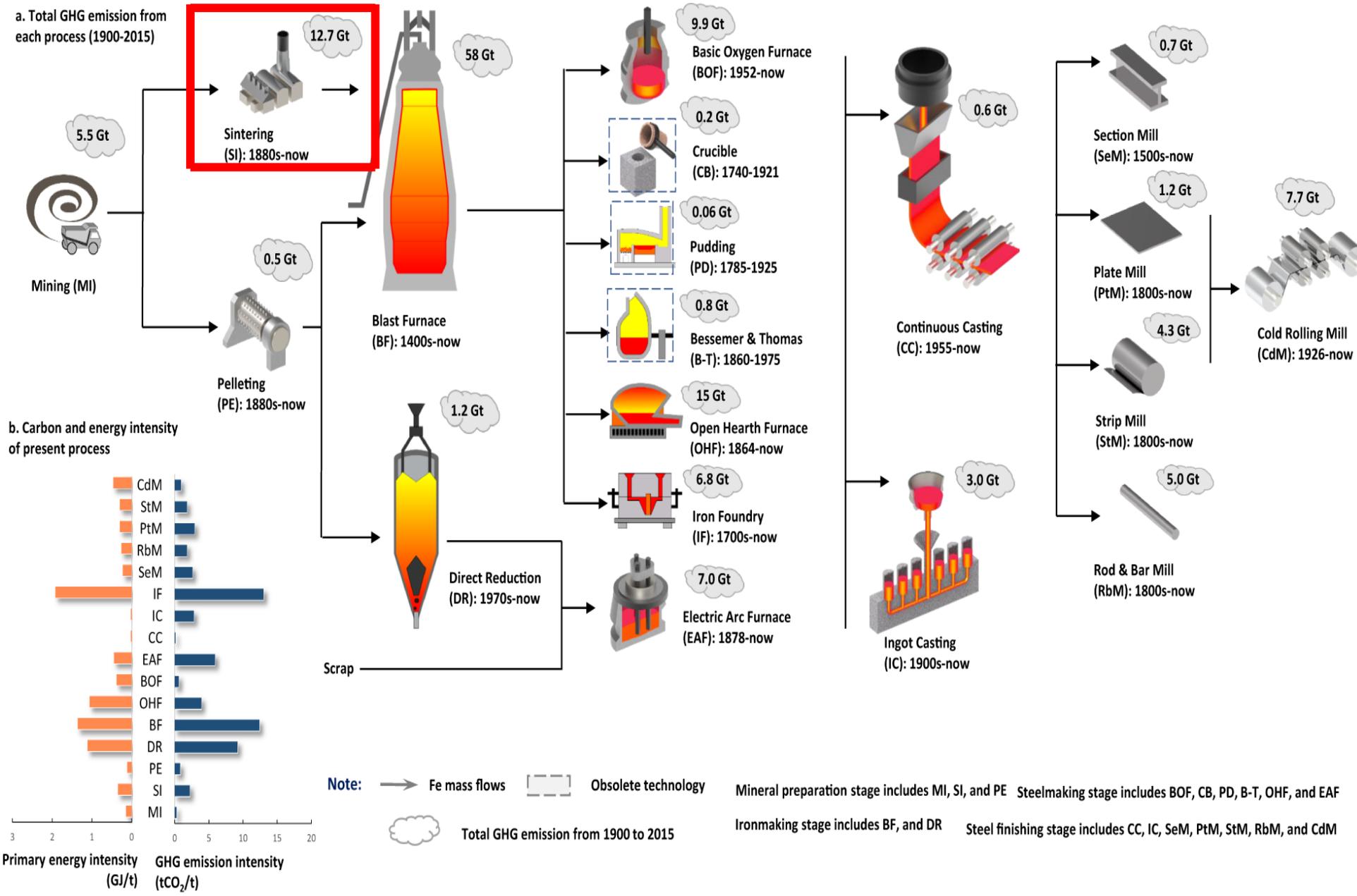


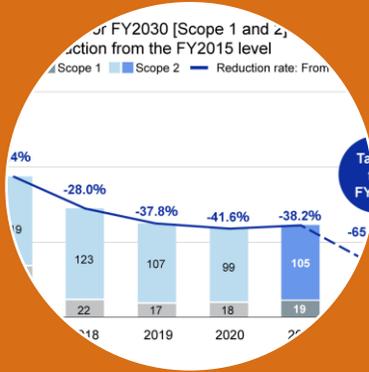
Image Credits: Various sources on the Internet

Objectives in Steel (Sinter) Production



Throughput

Maintain or Increase:
impacts Revenue



Emissions

Maintain or Reduce:
impacts Carbon Offset
Costs



Equipment Availability & Efficiency

Zero Down-Time: Spares
Cost, Profitability



Scale across new Sinter Processes, Technologies

Image Credits: Various sources on the Internet

Top-of-Mind Questions for Models/Data to Answer



Estimation

Next Batch: Steel
Throughput

Next Batch: Return Materials



Forecasting with Recommendations

Future Batches: Steel
Throughput & Emissions

Future Batches: Input
Change Recommendations
for Target Output



What-If Analysis

Cost-Benefit Tradeoffs:
Simulations of Input vs
Forecast

Evolution of Models in Steel (Sinter) Production

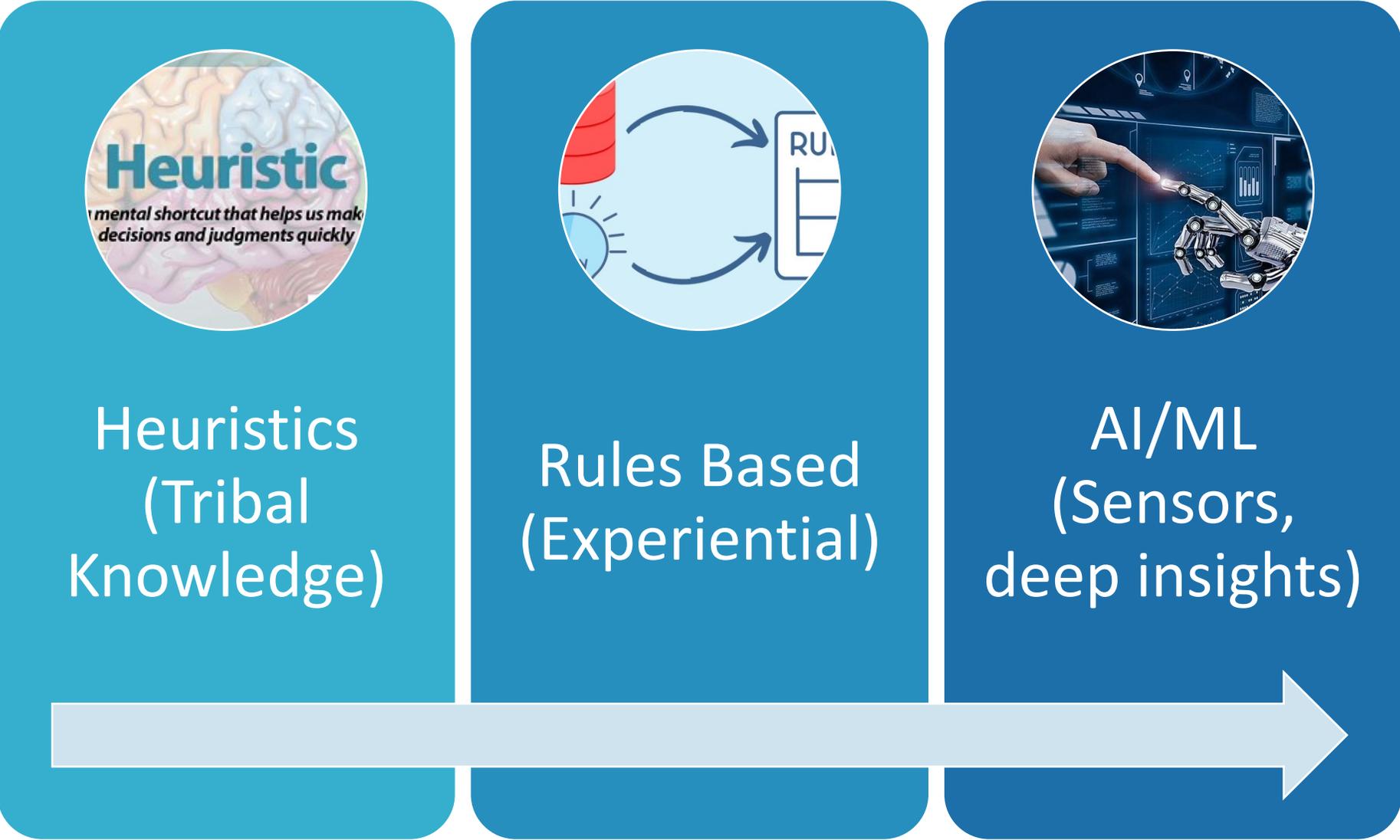


Image Credits: Various sources on the Internet

Sinter AI/ML Models : Sensors (p.s. learn how they work!)

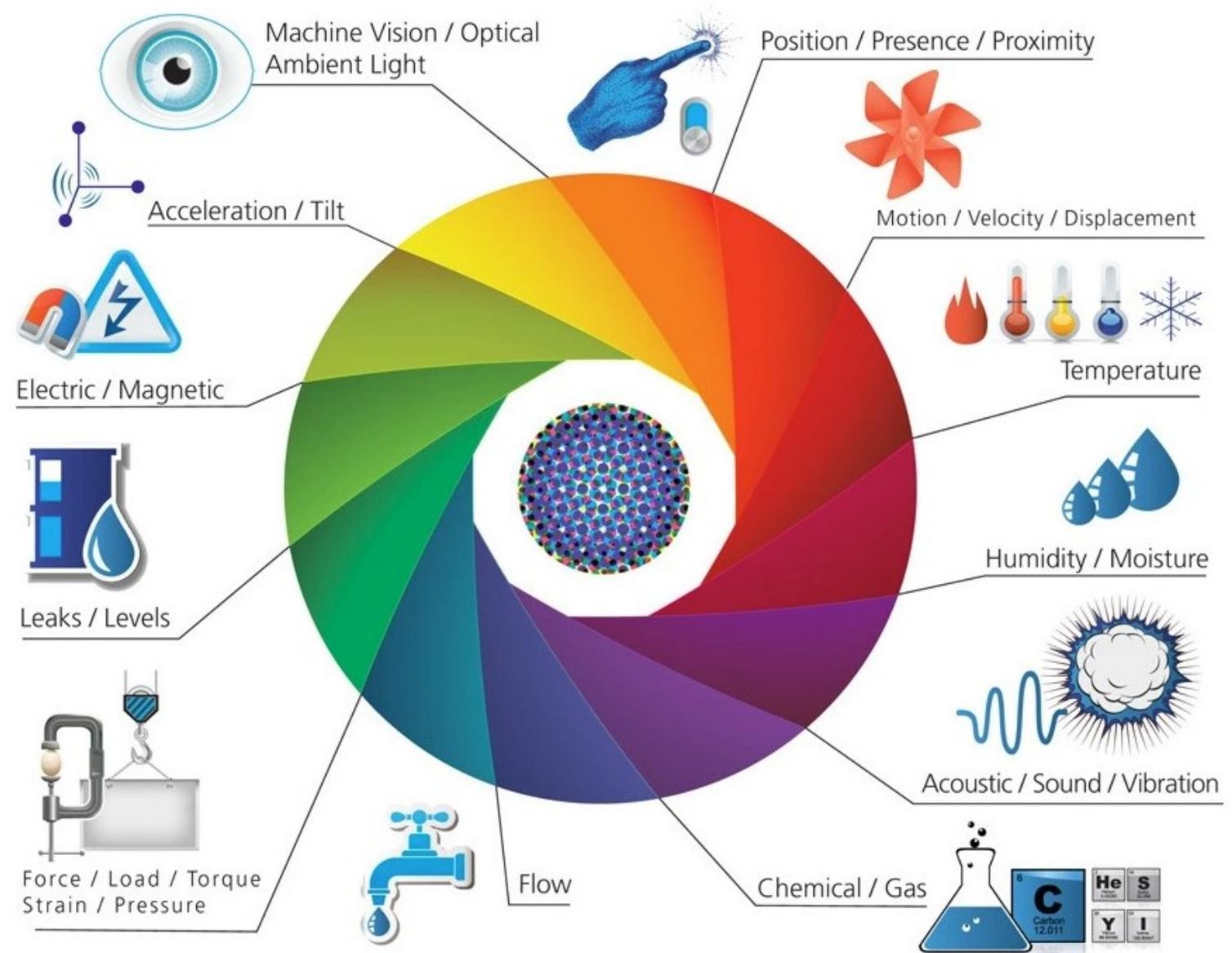


Image Credits: Various sources on the Internet

Copyright ©2023 VaNaShRi Technology Consulting

AI/ML Models: Data Challenges

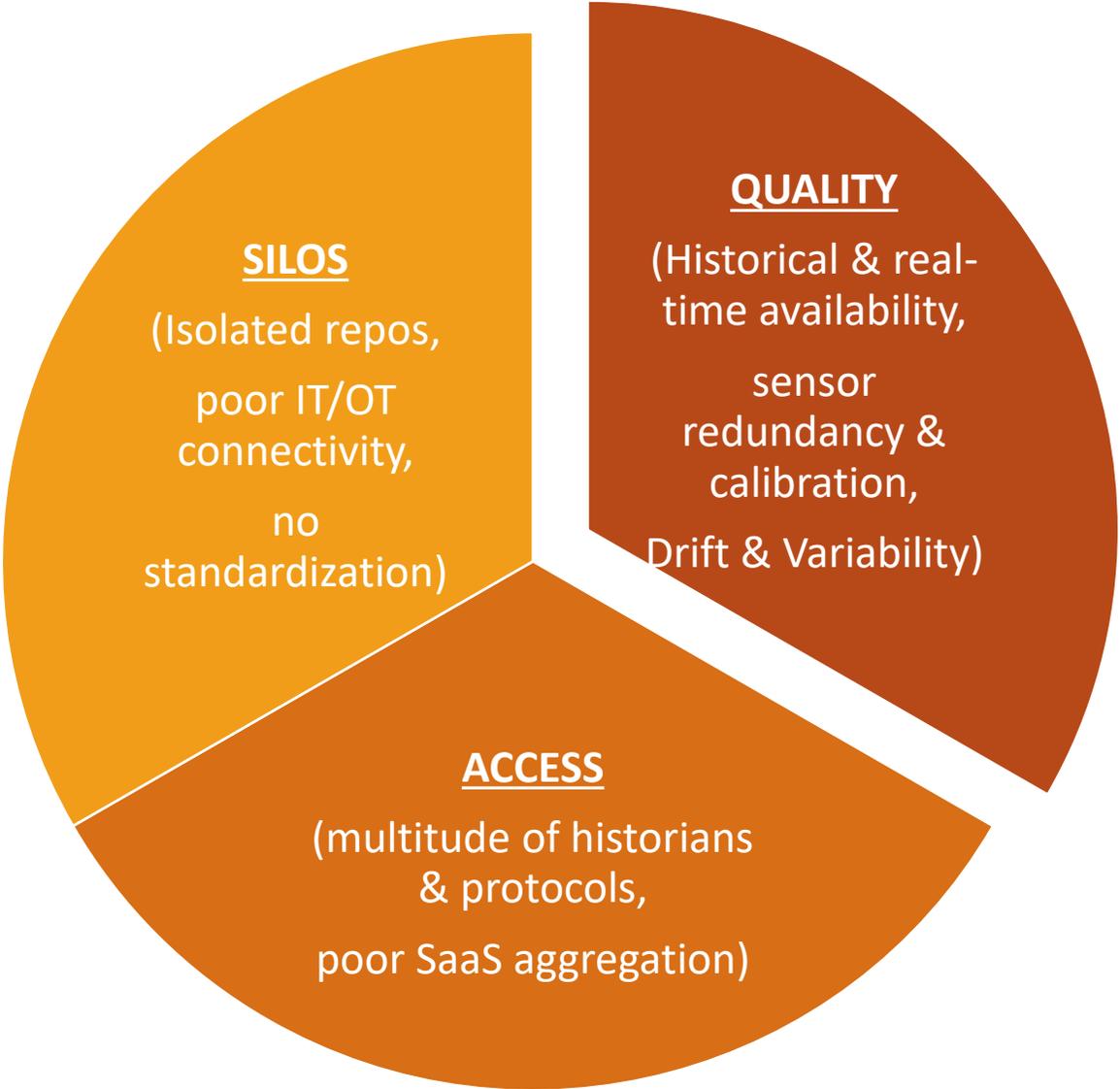


Image Credits: Various sources on the Internet

Mitigating Data Challenges



Connectivity
Standarize IT/OT connectivity, data frequency, formats



Quality
Redundancy design,
Calibration error detection,
Data error design,
Data drift detection,
Environmental monitoring



Access
Security design,
SaaS connectivity,
Edge design

Scale across different Sinter plants and Process Variations

Image Credits: Various sources on the Internet

Software Deployment Architecture

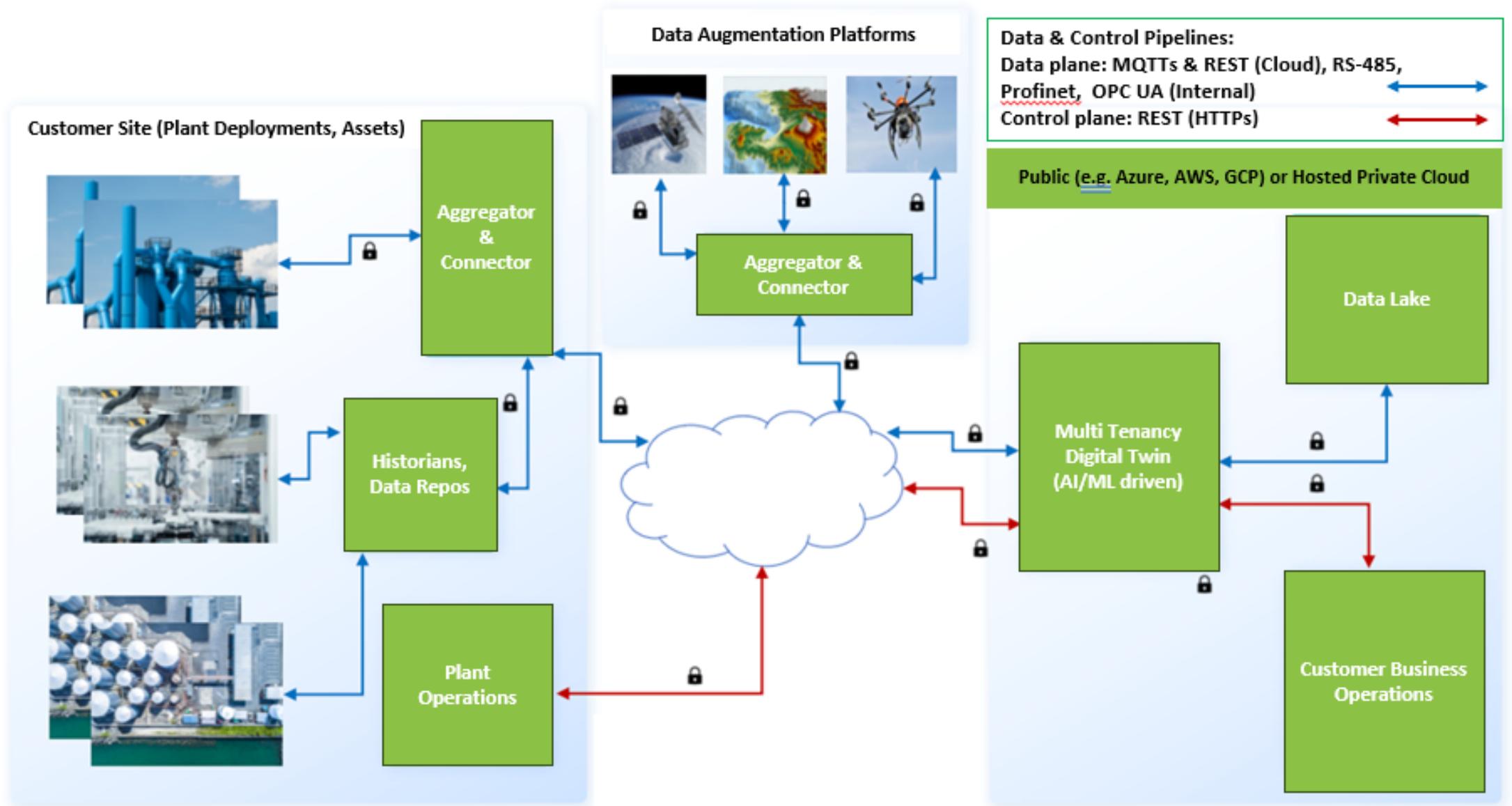


Image Credits: various sources on the internet

Copyright ©2023 VaNaShRi Technology Consulting

Data Pipelines (p.s. this too is important!)

Evolution of Data in Motion

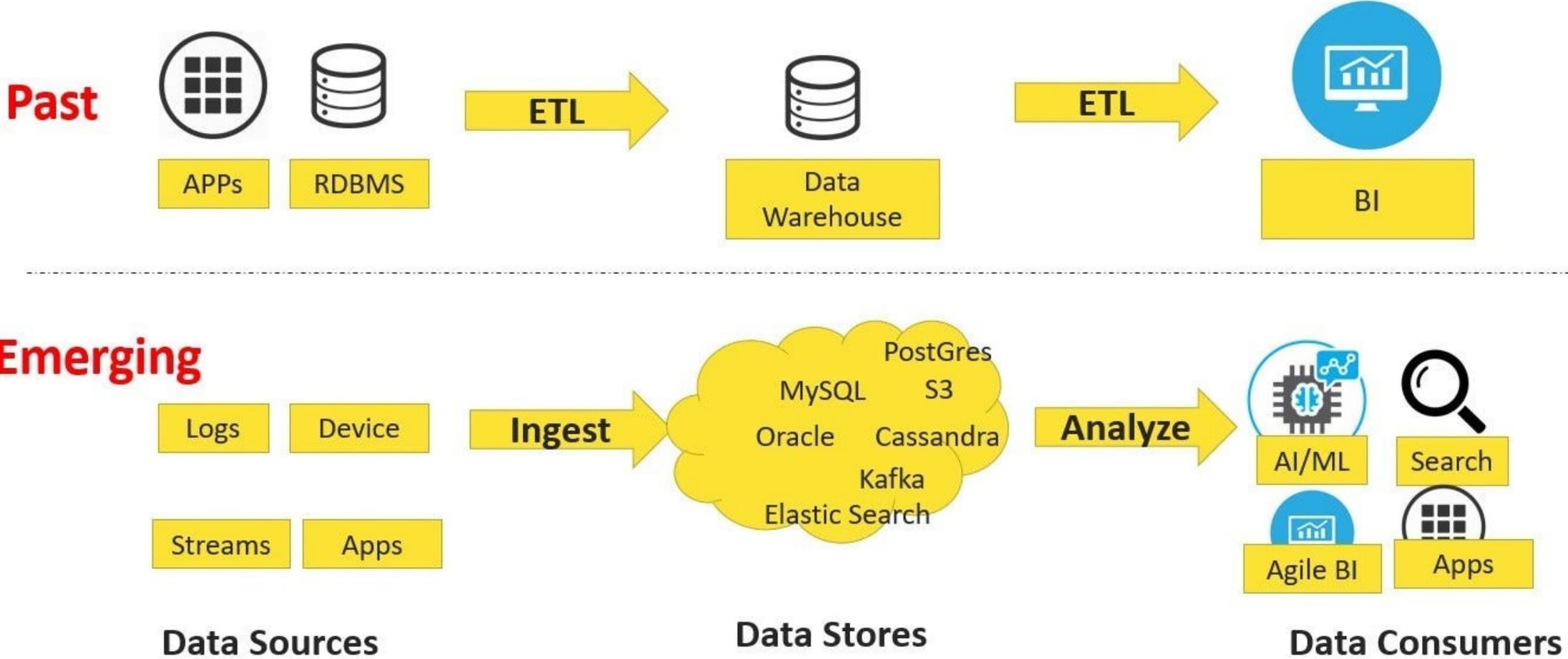


Image Credits: Various sources on the Internet

Data Pipelines : Challenges

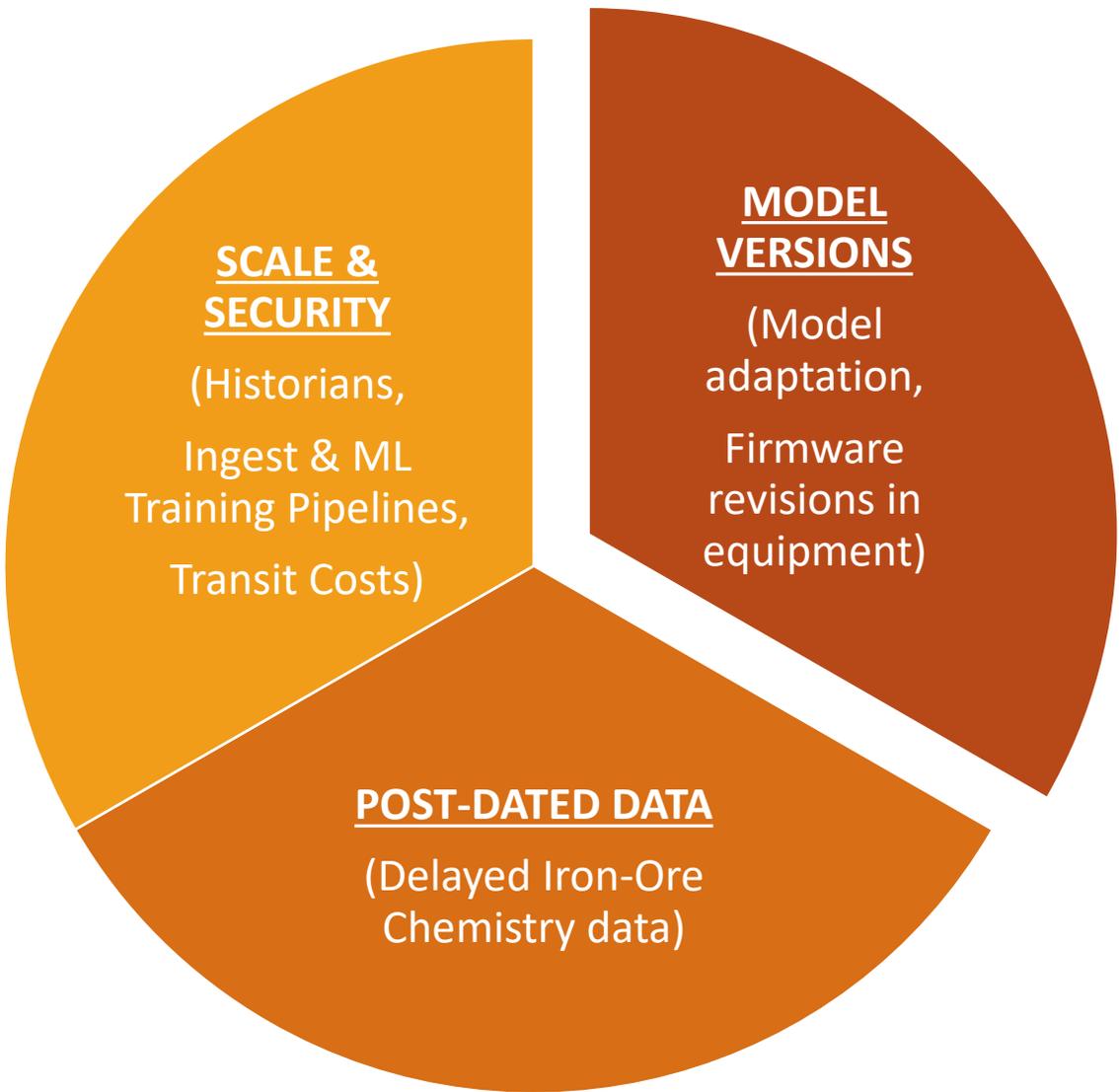


Image Credits: Various sources on the Internet

Mitigating Data Pipeline Challenges



Scale & Security

Multi Tenancy Design,
SOC2, RBAC & Credentials
Design,
Common Data Clouds,
Priority channels



Model Versions

Standardized and Versioned
(Internal) Data Models,
firmware and equipment
versioning,
Data abstraction



Post-Dated Data

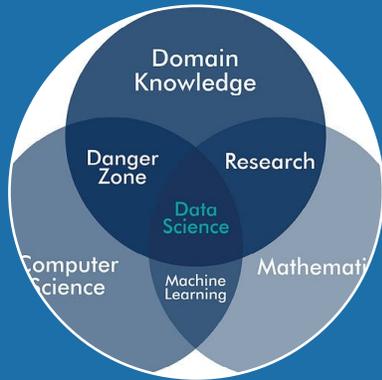
Impute post-dated data
based off domain expertise
e.g. previous batch
chemistry



Scale across different Cloud Deployments, Different Plants

Image Credits: Various sources on the Internet

Sinter Plant ML Models : Can It/Will It Work ?



Domain Knowledge

Process expertise ?

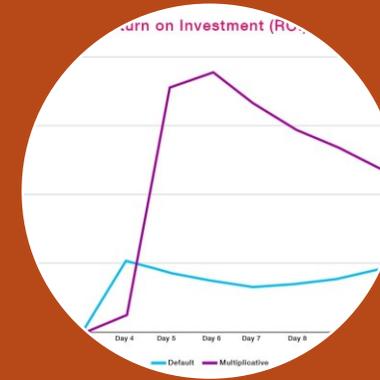
Data semantics expertise ?



Explainability

Can AI/ML surpass 30-40 years of process experience ?

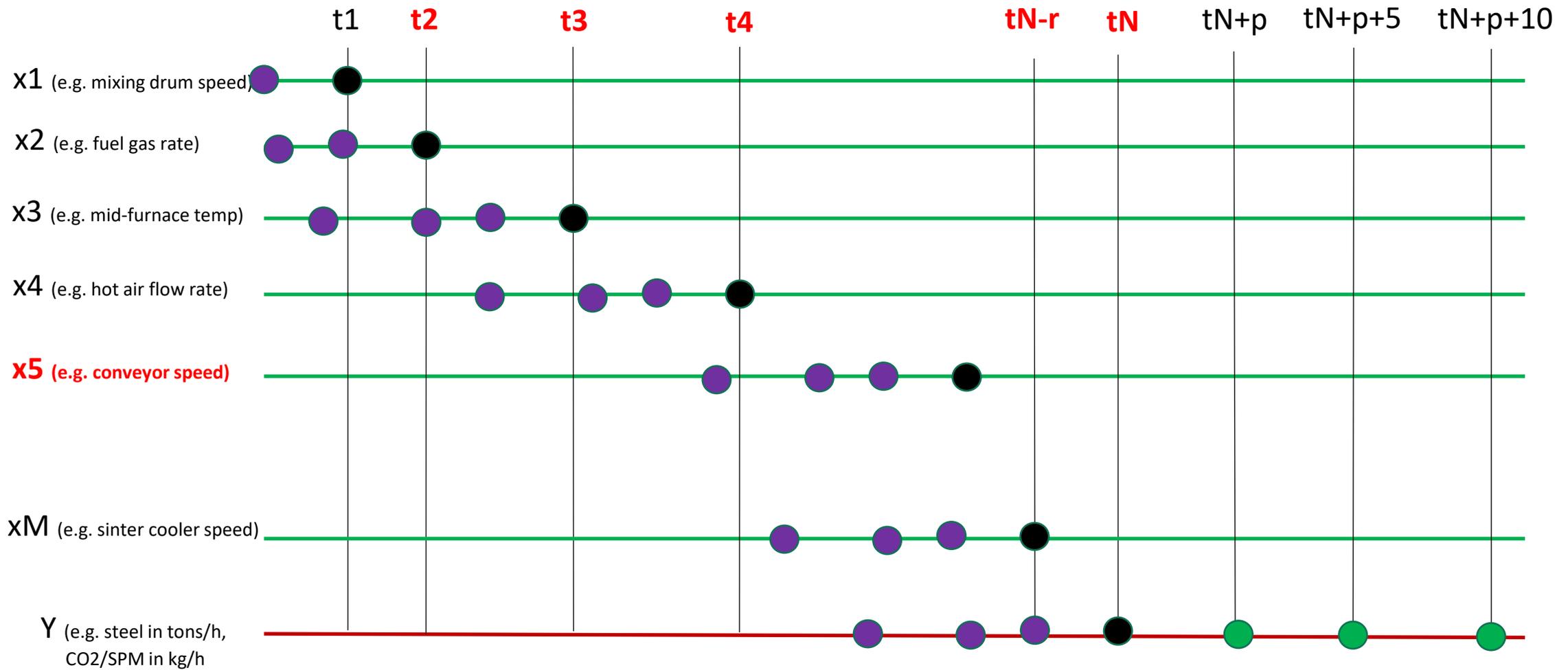
How do I explain results, demonstrate repeatability ?



Returns

Quantify the returns?
What will happen when process technology matures?

Sinter Plant ML Models: Time Series Nuances

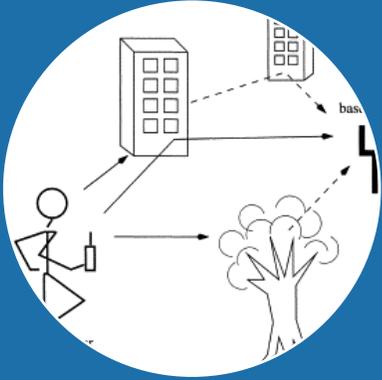


$Y(t_N) = f(X(t_1, t_{N-r}))$: estimator for Throughput & Emissions

$Y(t_{N+p}, t_{N+p+10}) = g(X(t_1-Z, t_{N-r}))$: forecaster for Throughput & Emissions based of Z samples of X

$ty = h(x_5)$ for $1 < y \leq N$: h is a perturbation for each $t_2 \dots t_N$

Sinter Plant ML Models : Approach



Account for variable tY's ?

Analyse $X(t)$ vs $Y(t)$ to determine best $\{tY\}$

Analyse $x5(t)$ with Sinter SOP design to determine best $x5$ and hence best $\{tY\}$

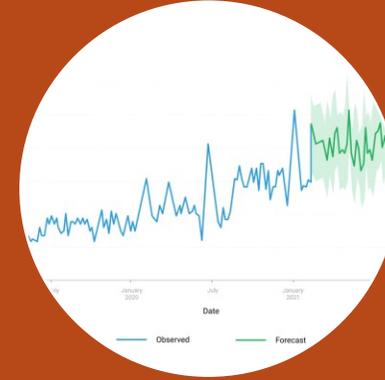


Estimator $f(x)$

Dataframe realignment: i.e. $\{xI\}$ with $\{tY\}$

SVM, DNN, ...

Temporal models: humidity, temperature, ore H₂O-content



Forecaster $g(X)$

T future timeslots from "Z"
 $X(t)$ collections

LSTM, B-LSTM, ...

No single all-in-one model for $f(X)$ or $g(X)$

Sinter Plant ML Models : Approach (cont'd)



Recommendations ?

Recalibrate $g(X)$ to a specific $tN+F$ future timeslot

Leverage PSO optimizer for optimal X values for given Y objectives

X variables cannot be adjusted in one step. Need to follow Sinter SOP for step-increments

“F” and SOP introduce brittleness owing to per plant/process variations and management objectives

AI/ML for Steel (Sinter) Plants : Lessons & Questions

- ❑ Domain knowledge: Data Semantics, Physical Process Layout and Sinter SOPs is a Must
- ❑ Enterprise grade focus on Errors, Support/Troubleshooting, Diagnostics, HA, Recovery and Security.
- ❑ Robustness of AI-based algorithms to be validated via FMEA design
- ❑ Need many models (temporal & geo-based) even for the same process and customer
- ❑ Is “Transfer Learning” possible across sinter plants – jury is still out?
- ❑ Approach is reusable across different processes – how do I reduce domain and training time?
- ❑ Closed-loop control: Insufficient Data Sets to reliably deploy for critical operations



THANK YOU !

Narayanan Subramaniam

LinkedIn: <http://www.linkedin.com/in/cnsubramaniam>

WebSite: <https://vanashri.com>, <https://climate350.com>