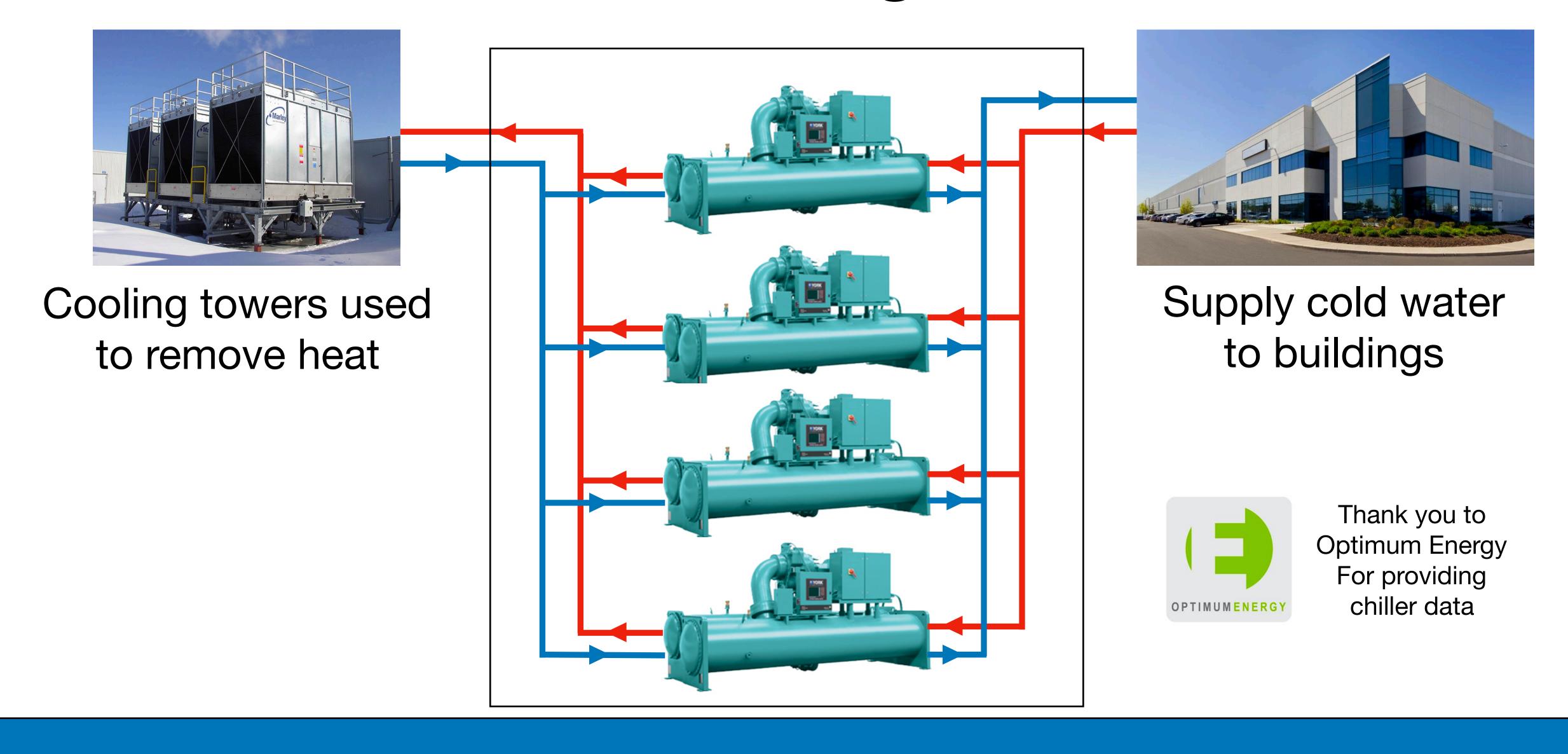
Predicting Industrial Chiller Performance

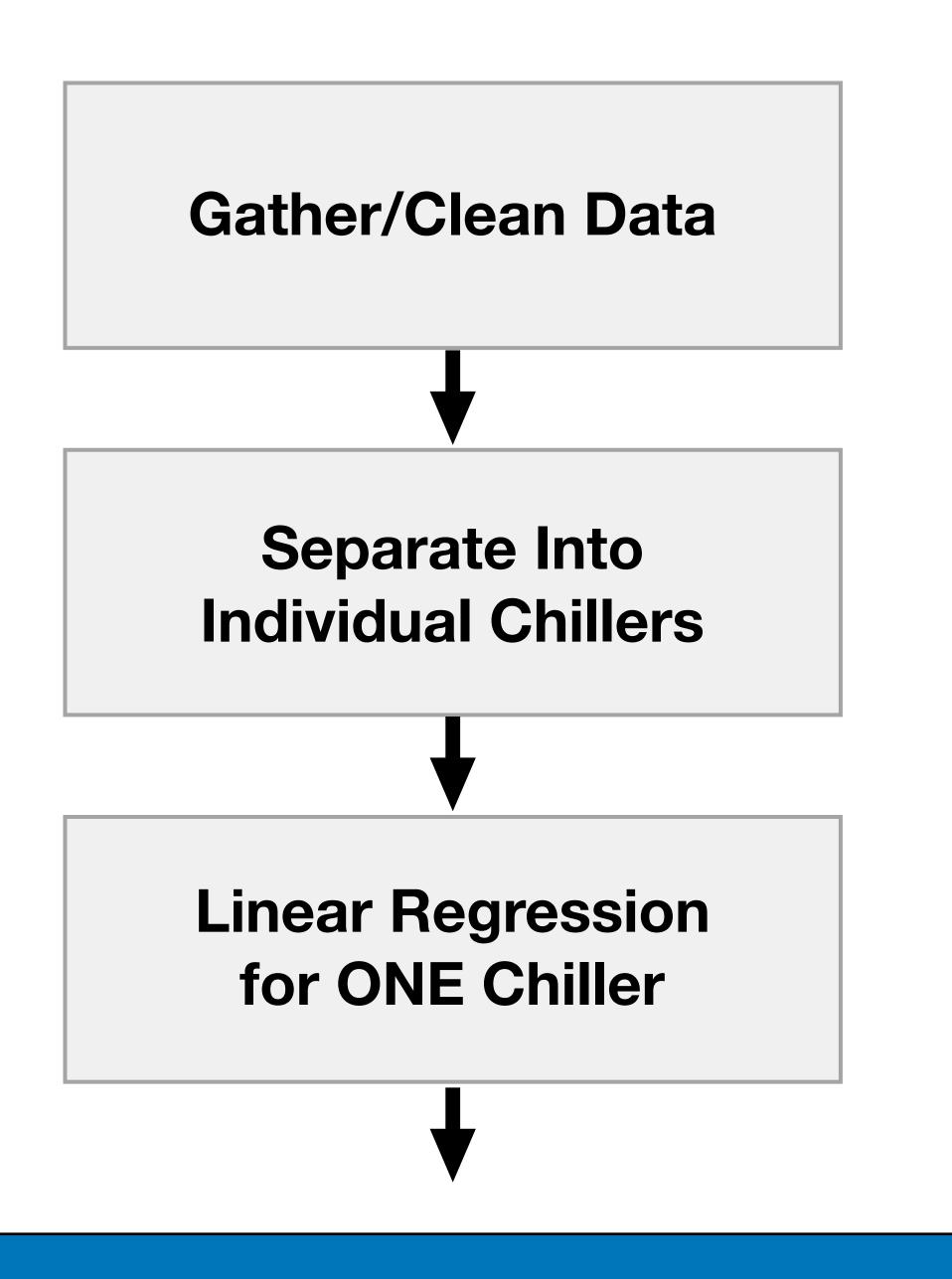
Dana Lindquist, PhD

Chiller = Heat Exchanger



Chiller Plants

- Chiller manufacturers provide some limited design specifications when they sell a Chiller (usually end up collecting dust in a filing cabinet).
- Most Chiller Plant operators do not think about efficiency.
- Optimum Energy collects data on Chiller Plants.

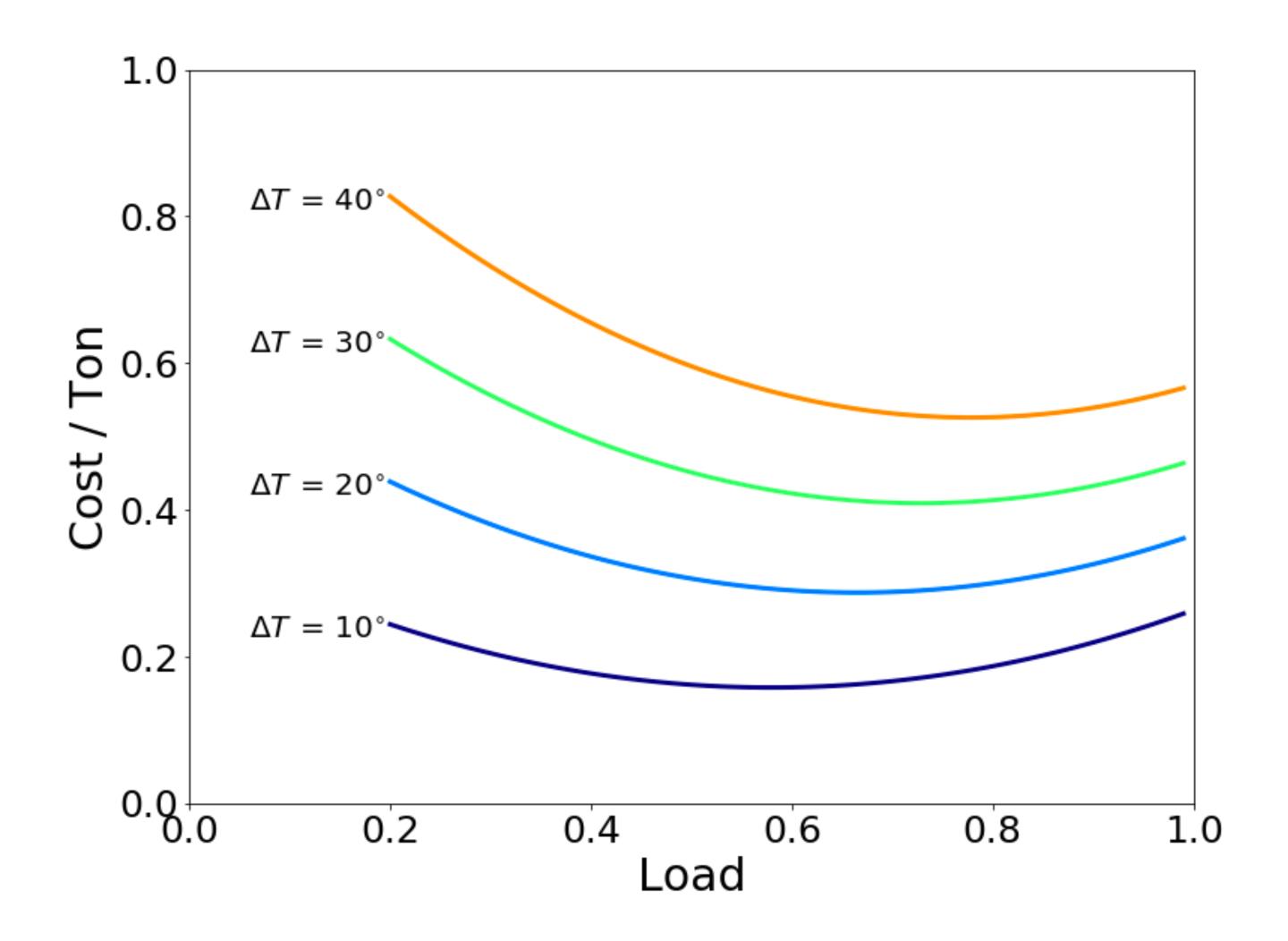


Remove data when Chiller in ALARM or just starting or stopping

30 distinct but similar Chillers 8 different sizes & configurations

Scikit-learn Linear Regression with LASSO Regularization

Just One Chiller



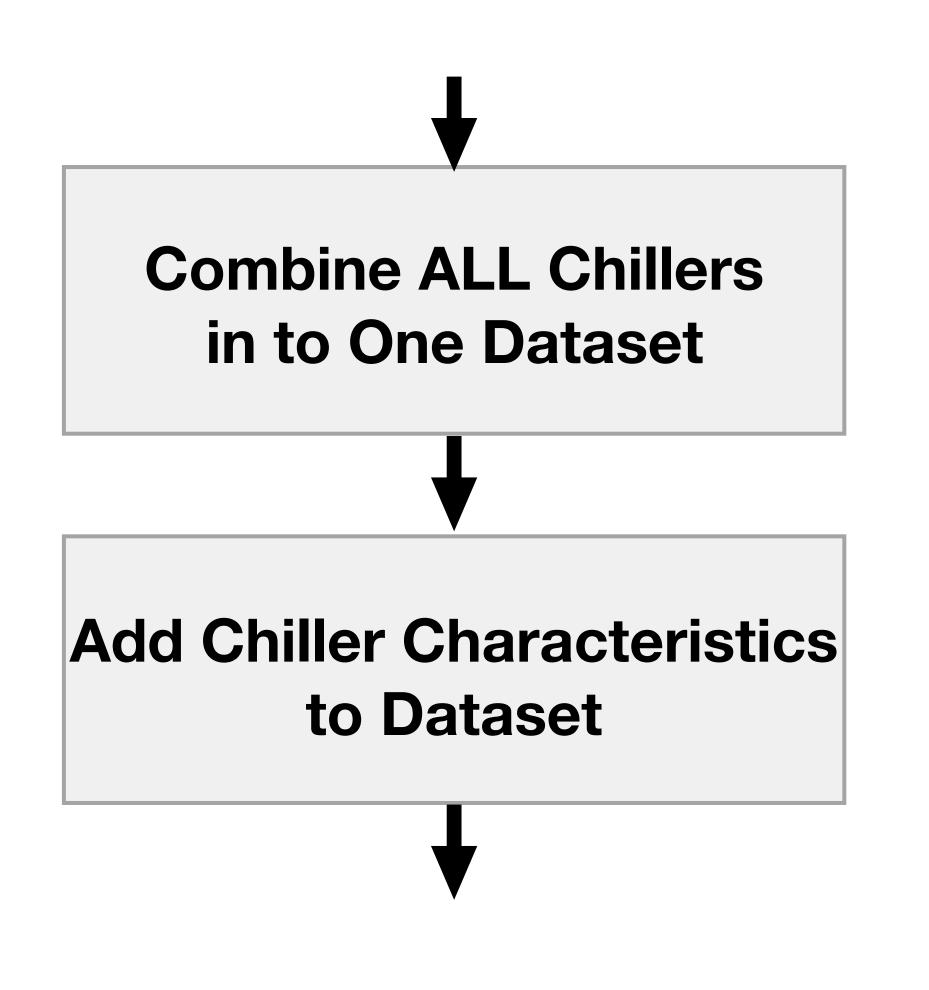
Ton = unit of refrigeration

Load = how hard the
Chiller is working
(fraction of max power)

Quadratic in Load

Cost/Ton =
$$(A + B \Delta T)$$
 + $(C + D \Delta T)$ Load + $(E + F \Delta T)$ Load²

Want to predict A, B, C, D, E & F based on chiller characteristics



1 million data points

Rated Tons, Variable Speed or not, . . .



Determine constants in Quadratic based on Chiller Characteristics

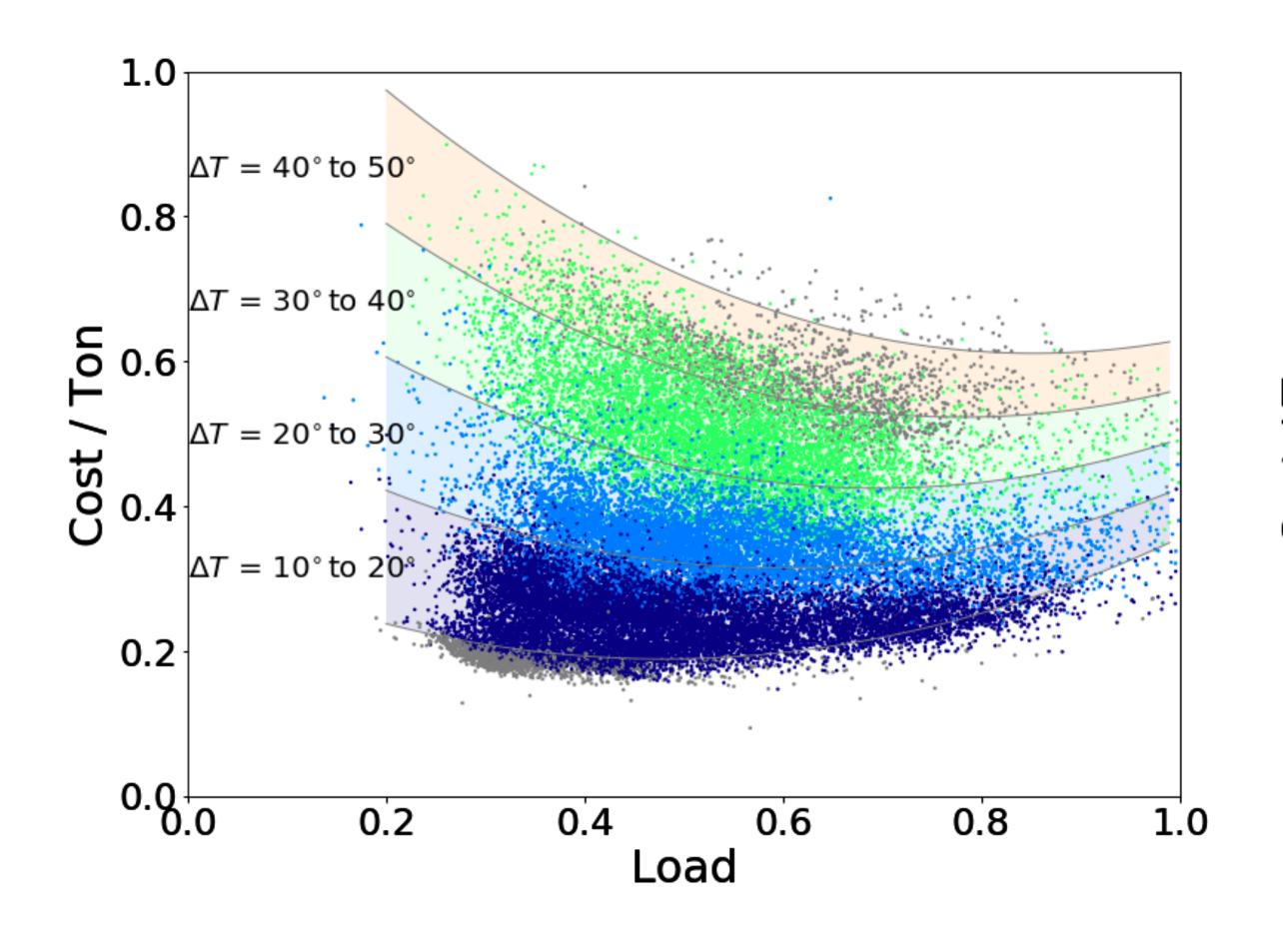
Predict quadratic constants

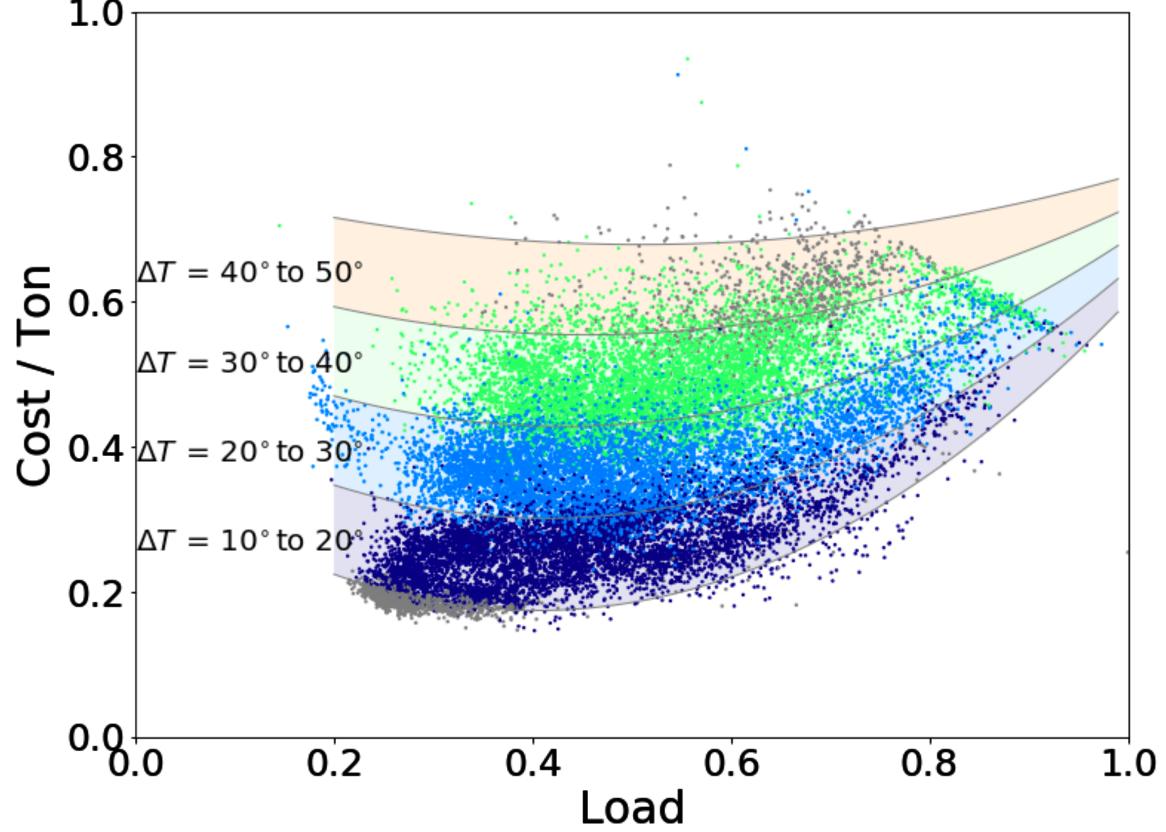
- 1. Linear Regression
- 2. Bayesian Model using PYMC3

Has the ability to extend this analysis to more chiller characteristics

Results

Two DIFFERENT 900 Ton Chillers



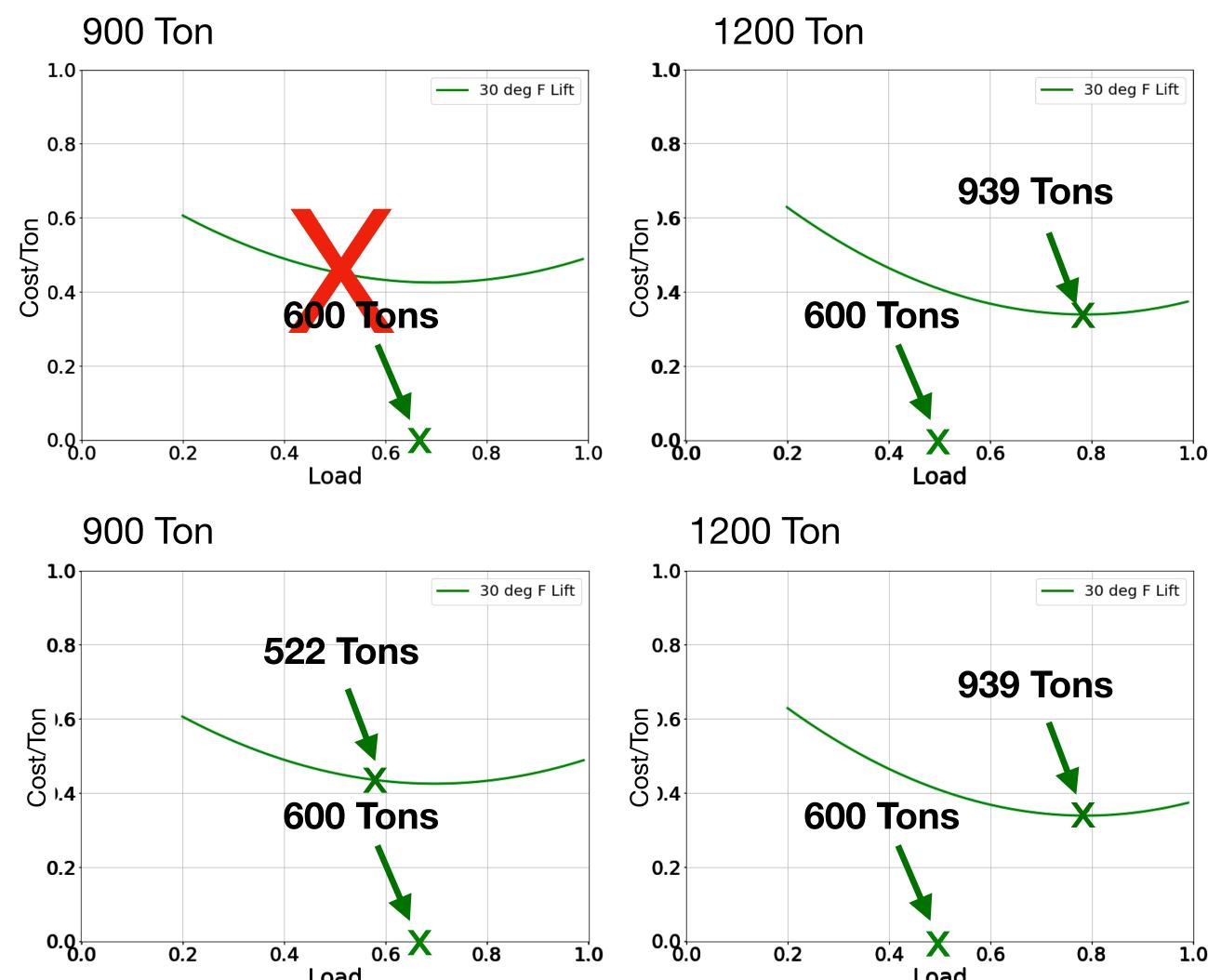


How to Run the Plant Efficiently?

Requirement: 2400 Tons of cooling 30° F ΔT

Goal: Lowest Cost

Barvialy \$t2j000s (at660)0pTemseaach



Conclusion

Given the data, we can predict Chiller performance

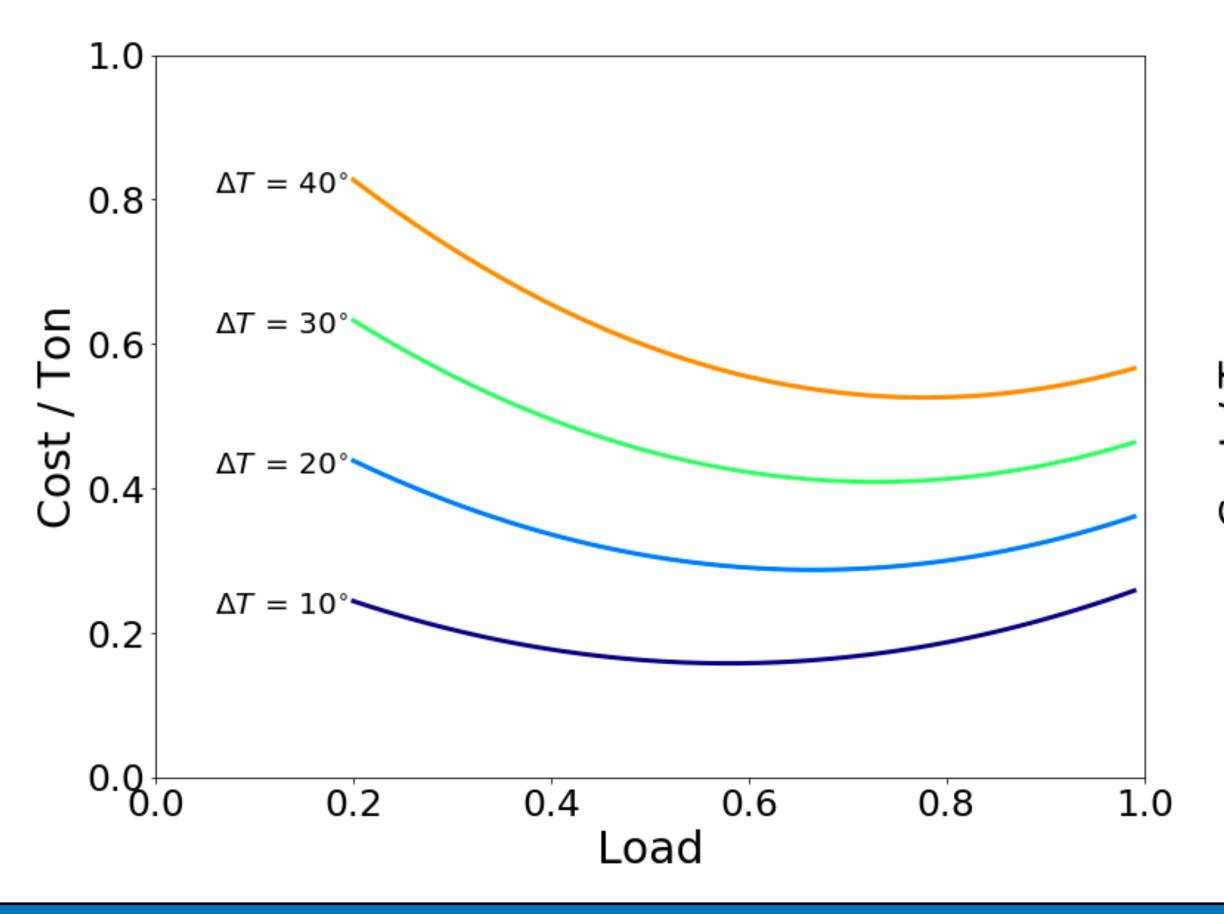
With a model for the system, we can make intelligent decisions on how to operate the system.

Thank You

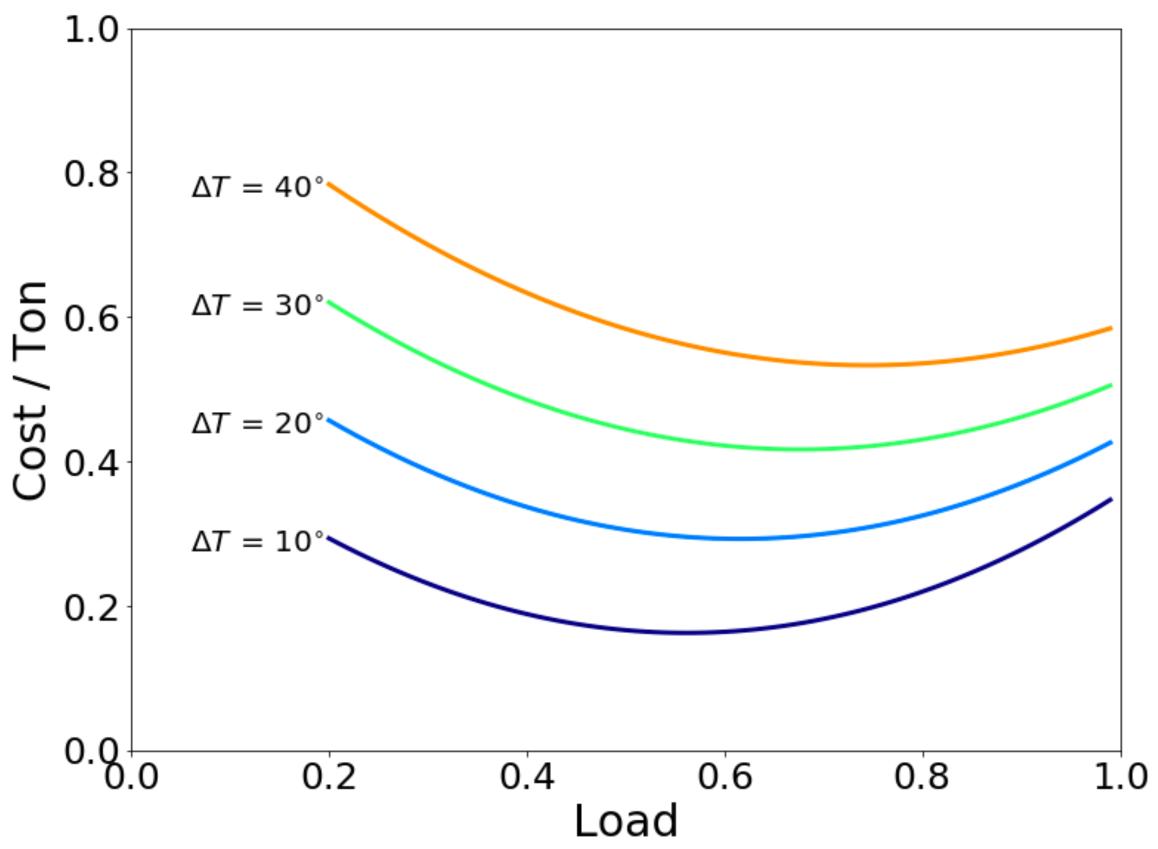
Dana Lindquist, PhD

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- www.linkedin.com/in/danalindquist
- github.com/Stitchmaker

Scikit-learn Linear Regression



PyMC3



 $Cost/Ton_{obs} = (A + B\Delta T) + (C + D\Delta T)Load + (E + F\Delta T)Load^{2}$

 $Cost/Ton = \mathcal{N}(Cost/Ton_{obs}, \mu, \sigma)$

 $A, B, C, D, E \& F = \mathcal{N}(\mu, \sigma)$

```
with Model() as model:
    sigma = HalfCauchy('sigma', beta=10)
    inter = Normal('Intercept', 0, sd=.01)
    D_coef = Normal('Load', 0, sd=.01)
     D2\_coef = Normal('Load2', 0, sd=.01)
    DT_coef = Normal('LoadLift', 0, sd=.01)
    D2T_coef = Normal('Load2Lift', 0, sd=.01)
    T_coef = Normal('Lift', 0, sd=.01)
     # Define likelihood
     likelihood = Normal('kWTon', mu=inter +
                 D_coef * X['Load'] + D2_coef*X['Load^2'] + T_coef*X['DTLift']
                 + DT_coef*X['Load^2*DTLift'] +
                 D2T_coef*X['Load^2*DTLift'],
                 sd=sigma, observed=y)
```

