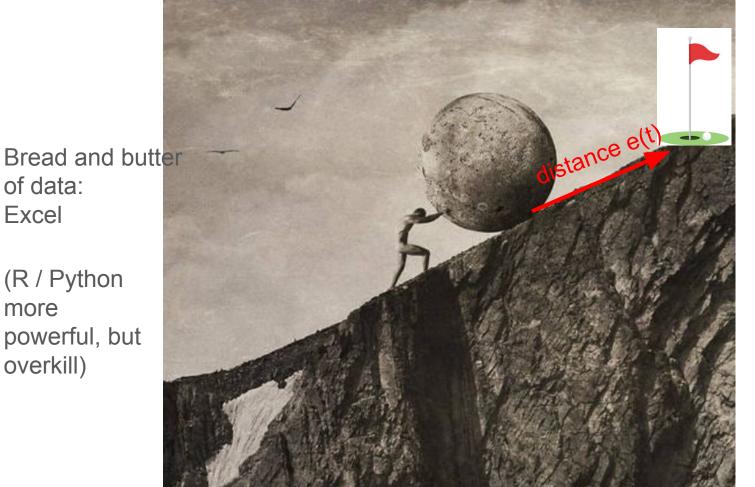
Crash Course on HVAControl Theory by Ben

(Ben's a math guy)



Bread and butter of controls: PID

(State-Space model more powerful, but overkill)

(R / Python more powerful, but overkill)

of data:

Excel

You are the caddy to a blindfolded golfer! What's your strategy?

Options for strategic guidance:

Golf Clubs:

Tee: Driver

Fairway: Woods, Irons, Wedges

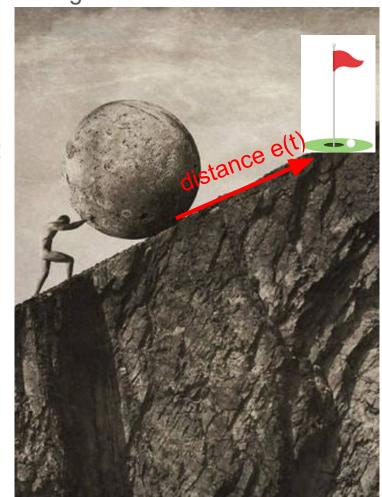
Green: Putter

Controls:

P: Proportional

I: Integral

D: Derivative



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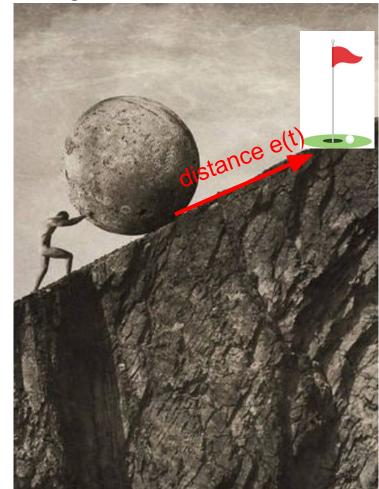
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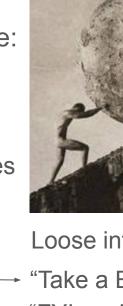
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stance e(t)

Bread and butter of controls:

PID

Loose intuition:

→ "Take a BIG swing!"

→ "FYI you've been stuck in sand trap a while"

→ "Hey someone's moving the flag back"

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Controls: Signal to process:

P: Proportional $\rightarrow e(t)$ I: Integral $\rightarrow \int_0^t e(\tau)d\tau \rightarrow \frac{de(t)}{dt}$ D: Derivative $\rightarrow \frac{de(t)}{dt}$

Only accessible present information!

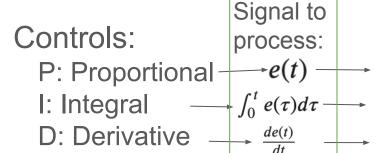
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Decision!

Asymmetry: Ball rolls down, Sisyphus stuck pushing below

heater + no AC = passive cooling only

Big Inertia: Sisyphus's gigantic boulder

- Long while for room to heat up
- Slow vent airflow
- Humidity takes time to alter

No Momentum: everything submerged in molasses



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Though building's entire system of physics unknown... can reasonably assume slow changes only ==>

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Hints at the following surefire strategy: simply make P large, then tweak I to correct any long-term offset biases (intuitively, think about why this should work indeed!)



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Panacea PID game's abstract mathematical form

plant/system dynamics: $\dot{\mathbf{x}} = f(\mathbf{x}, u, t)$ f and h "nicely behaved" black-box: sensor/actuator readoutput: $y(t) = h(\mathbf{x}(t), t)$

Goal: make tracking error zero
$$e(t) = r(t) - y(t)$$
desired reference value

Controller's algebraic form: $u(t) = K_P e(t) + K_I \int_0^t e(\tau) d\tau + K_d \frac{de(t)}{dt}$

3 Degrees of Freedom to Heuristically + Empirically tune

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desired reference Controller's algebraic $u(t) = \underline{K_P}e(t) + K_I \int_0^t e(\tau)d\tau + K_d \frac{de(t)}{dt}$ form:

In practice: fuddle, tinker, experiment, then Guess & Check three constants using all the intuition we've learned up to now!

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behaved" black-box:

3 Degrees of Freedom to Heuristically + Empirically tune

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Can we do better?

 $y(t) = h(\mathbf{x}(t), t)$ In practice: fuddle, tinker, experiment, then Guess & Check three constants using all the intuition we've learned up to now!

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$$u(t) = \underline{K_P}e(t) + \underline{K_I} \int_0^t e(\tau)d\tau + \underline{K_d} \frac{de(t)}{dt}$$

No because the system is a black-box!

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That's ok people still do deep learning + hyperparameter tweaking upon blackboxes!

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$$\dot{\mathbf{x}} = f(\mathbf{x}, u, t)$$
 f and h "nicely behaved" black-box: $y(t) = h(\mathbf{x}(t), t)$

Irony: lack of underlying assumptions behind an unknown black-box makes this the most generalized universal principle and tool

And with simply just the premise of "well-behaved" mathematical functions:

Canonical State Space Reformulation Generalization

$$\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u}$$

$$y = Cx + Du$$

Notion of "well-behavedness" pertinent to PID:

- Exists a localized linear approximation for underlying LTI system
- Convenient framework for mathematical analysis
- Particularly relevant for Laplace domain
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While there are fancy general PID tuning methods such as:

- Cohen-coon
- Ziegler-Nichols
- Astrom-Hagglund
- Tyreus Luyben
- SIMC

But for HVAC where even the D in PID is mostly irrelevant, that brings us back to:

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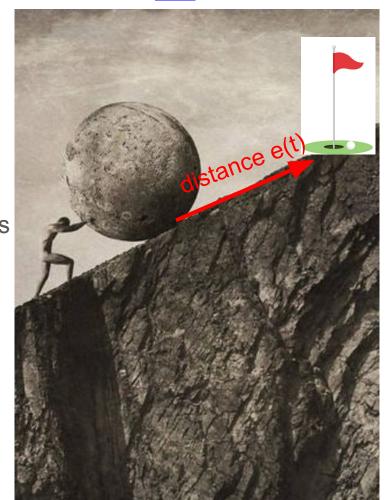
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On a fun brief interdisciplinary note:

They use similar techniques in other fields too:

Economics, for controlling dynamics of economic variables: Consumption, Production, Interest Rate, Inflation, GDP, etc.

Machine Learning, for time-series-type data: Inferring control parameters governing dynamics as opposed to classical probability distribution parametric estimation of underlying feature vectors

Physics, for field theory:
Sensitivity analysis of parameters that control a system is the essence of theoretical physics (Hamiltonian Mechanics, Statistical Mechanics, Principle of Least Action, Chaos)

Questions?

Thanks!