# Black Boxes or Rube Goldberg Machines? Neural Networks as Cybernetic Regulators

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#### Overview

- Explanatory Demographic
  - Math vs. Concepts
- 2 Black Boxes vs. Rube Goldberg Machines
- Cybernetic Regulators
  - Ashby Regulator
- 4 Error-Controlled Regulator

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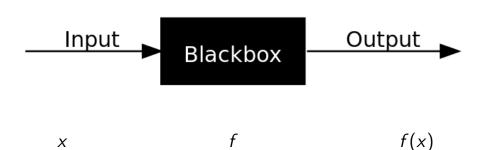
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  - No. Important for explaining to demographic 2.





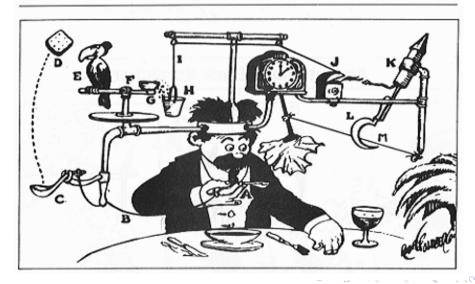
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### Self-Operating Napkin



• In principle transparent, by design convoluted with many mechanisms and parameters.

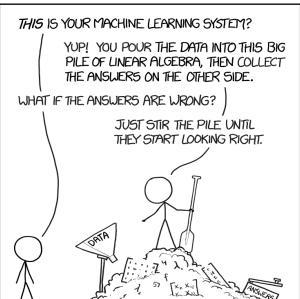
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- Don't need to understand each part/mechanism... but general idea of what kind of object we are dealing with.

#### Pile of Math?



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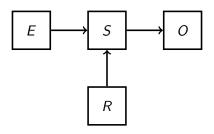
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- Explanation for the people: go back to theoretical and conceptual foundations: cybernetic regulators.

## Cybernetic Regulator

A cybernetic regulator is a complex system which controls environmental inputs by appropriate actions, resulting in a state aligning with a regulatory goal.

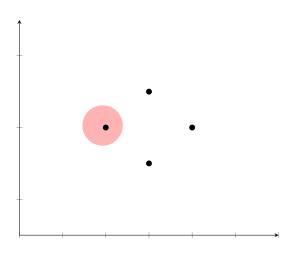


## Ashby Regulator (Ashby, 1958)

		R	
Ε	$\alpha$	β	$\gamma$
	β	$\alpha$	$\gamma$
	$\gamma$	$\alpha$	$\delta$
	$\delta$	$\epsilon$	$\gamma$
	$\gamma$	$\delta$	$\epsilon$

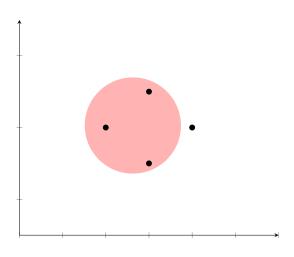
Table: Game in which E goes first. R plays for some outcome (e.g.  $\gamma, \delta$ ). Broadly, Ashby's Law of Requisite Variety says that "only variety in R can force down the variety due to E".

# Another Game (VC Dimension/Shattering)



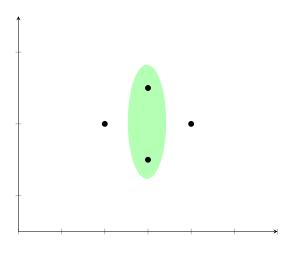
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### Another View: VC Dimension/Shattering



$$\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$$

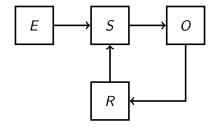
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- ANNs are Error-Controlled Regulators.

### Error-Controlled Regulator



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  Forgetting, Dropout...
- Developing Intuitions about the kinds of objects ANN models are:
  Cybernetic Regulators, complex systems like RGMs.

Thanks!

#### References

Ashby, W. R. (1958). An Introduction to Cybernetics. Chapman and Hall.