

# Dynamic Choice of Information Sources

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Why do we learn?

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Physics is like sex: sure, it may give some practical results, but that's not why we do it.

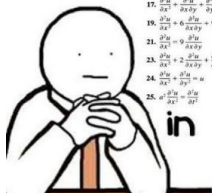
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Richard Feynman

# Why do we learn?

I'm still waiting for the day that I will actually use



$$\begin{array}{ll} 17. \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial x \partial y} + \frac{\partial^2 u}{\partial y^2} = 0 & 18. 3 \frac{\partial^2 u}{\partial x^2} + 5 \frac{\partial^2 u}{\partial x \partial y} + \frac{\partial^2 u}{\partial y^2} = 0 \\ 19. \frac{\partial^2 u}{\partial x^2} + 6 \frac{\partial^2 u}{\partial x \partial y} + 9 \frac{\partial^2 u}{\partial y^2} = 0 & 20. \frac{\partial^2 u}{\partial x^2} - \frac{\partial^2 u}{\partial x \partial y} - 3 \frac{\partial^2 u}{\partial y^2} = 0 \\ 21. \frac{\partial^2 u}{\partial x^2} = 0 & 22. \frac{\partial^2 u}{\partial x \partial y} - \frac{\partial^2 u}{\partial y^2} + 2 \frac{\partial u}{\partial x} = 0 \\ 23. \frac{\partial^2 u}{\partial x^2} + 2 \frac{\partial^2 u}{\partial x \partial y} + \frac{\partial^2 u}{\partial y^2} + \frac{\partial u}{\partial x} - 6 \frac{\partial u}{\partial y} = 0 & \\ 24. \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = u & \\ 25. a^2 \frac{\partial^2 u}{\partial x^2} = \frac{\partial^2 u}{\partial t^2} & 26. k \frac{\partial^2 u}{\partial x^2} = \frac{\partial u}{\partial t}, \quad k > 0 \end{array}$$

in real life

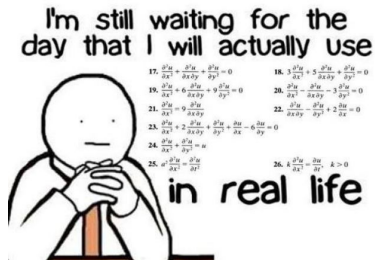
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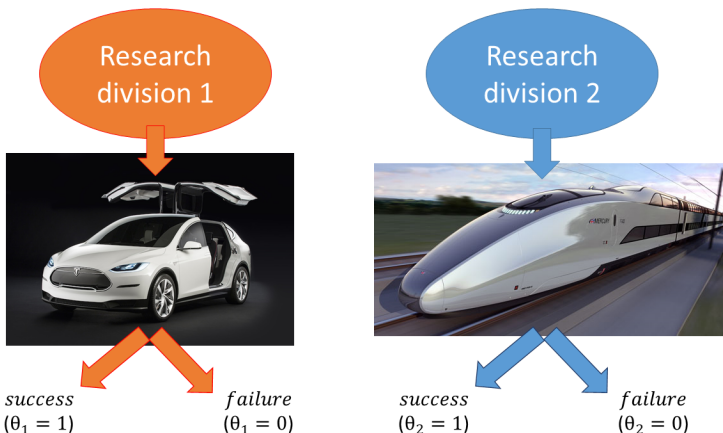
Richard Feynman

## Motivation

Formulating the goal of learning is itself a hard task. *How much do we need to know about the goal to learn optimally?*

## Example

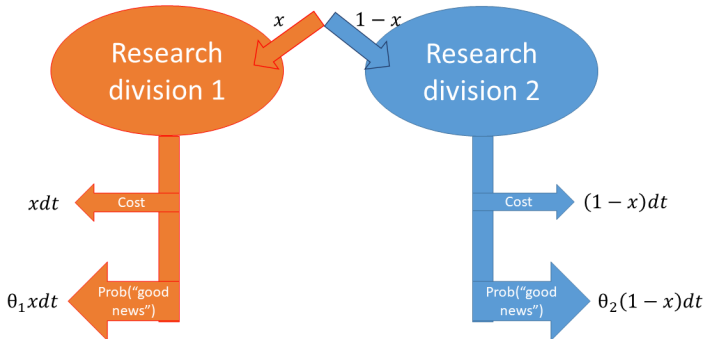
- ▶ A company has **two research divisions**
- ▶ Each division  $k = 1, 2$  investigates the profitability of a certain project, which can be either **success** ( $\theta_k = 1$ ) or **failure** ( $\theta_k = 0$ )
  - ▶  $\theta_1$  and  $\theta_2$  are not necessarily independent



## Example

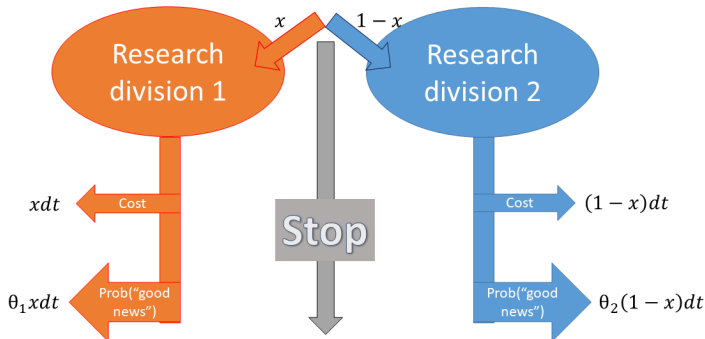
- ▶ At each instance of time, the company allocates a **unit of resources** between two divisions
- ▶ Each division  $k$  spends its resources  $x_k \in [0, 1]$  to **search for a proof** that its project is successful  $\theta_k = 1$

	$\theta_k = 1$	$\theta_k = 0$
Prob("good news")	$x_k dt$	0
Prob("no news")	$1 - x_k dt$	1



## Example

- ▶ The company decides when to stop research and make the decision



Which project(s) to invest in (if any)?



# General Framework

## Learning Stage

- ▶ four states of the world:  
 $(\theta_1, \theta_2) \in \{(1, 0), (0, 1), (1, 1), (0, 0)\}$
- ▶ two information sources:
  - ▶ source 1 = search for a proof that  $\theta_1 = 1$
  - ▶ source 2 = search for a proof that  $\theta_2 = 1$

## Decision Making Stage

- ▶  $\mathcal{A} = \{a_1, \dots, a_n\}$ : set of actions to choose from
- ▶  $u_i(a)$ : payoff from choosing action  $a$

	(1, 0)	(0, 1)	(1, 1)	(0, 0)
$a_1$	$u_1(a_1)$	$u_2(a_1)$	$u_3(a_1)$	$u_4(a_1)$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$a_n$	$u_1(a_n)$	$u_2(a_n)$	$u_3(a_n)$	$u_4(a_n)$

# General Framework

## Learning Stage *How to learn?*

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## Decision Making Stage *Why to learn?*

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$a_n$	$u_1(a_n)$	$u_2(a_n)$	$u_3(a_n)$	$u_4(a_n)$

# Solution

Optimal strategy balances the trade-off:

- ▶ *belief-based incentives*: use the source that leads to a higher  $\text{Prob}(\text{"good news"})$ 
  - ▶ depends on current **beliefs**
- ▶ *payoff-based incentives*: learn about the project that leads to a higher profit if successful
  - ▶ depends on **payoff** matrix

## Solution

All strategies can be divided into two classes:

1. No matter what happens in the future, at most **one source** is used.
2. **Both sources** can potentially be used (not necessarily simultaneously).

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- ▶ but the learning strategy simple! 😊
  - ▶ choose the source once
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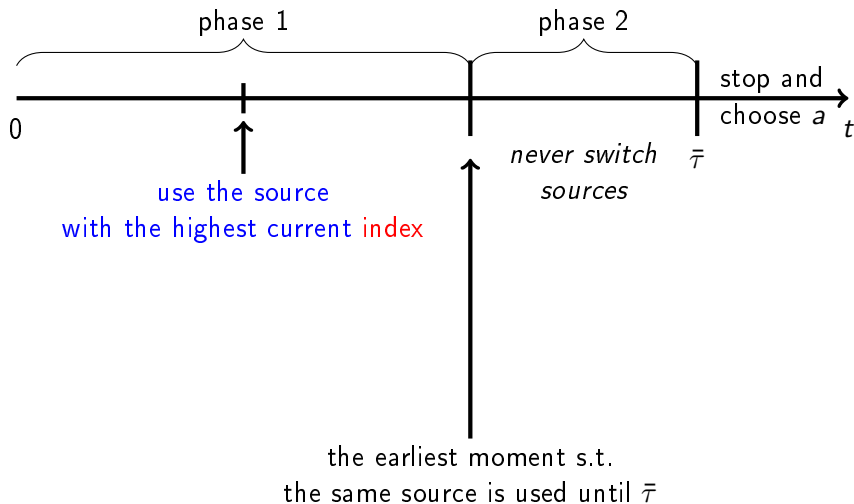
- ▶ the choice of the source depends on everything ☹
- ▶ but the learning strategy simple! ☺
  - ▶ choose the source once
  - ▶ then solve the optimal stopping problem

If the optimal strategy belongs to the second class, then

- ▶ choose the source with the highest index
- ▶ main result: expression for the index, how it depends on the payoff matrix

## Optimal Strategy

In the absence of "good news"





## Phase 1

Fix current beliefs

$(p_1 = \text{Prob}(\{1, 0\}), p_2 = \text{Prob}(\{0, 1\}), p_3 = \text{Prob}(\{1, 1\}))$ .

*If a positive signal occurs **now**, what is the optimal strategy?*

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Four possibilities:

1.  $\theta_1 = 1 \Rightarrow$  learn from source 2;  $\theta_2 = 1 \Rightarrow$  learn from source 1
2.  $\theta_1 = 1 \Rightarrow$  stop ;  $\theta_2 = 1 \Rightarrow$  learn from source 1
3.  $\theta_1 = 1 \Rightarrow$  learn from source 2;  $\theta_2 = 1 \Rightarrow$  stop
4.  $\theta_1 = 1 \Rightarrow$  stop ;  $\theta_2 = 1 \Rightarrow$  stop

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▶ phase 1 rule: use source 1
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▶ phase 1 rule: use source 2
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3.  $\theta_1 = 1 \Rightarrow$  learn from source 2;  $\theta_2 = 1 \Rightarrow$  stop  
▶ phase 1 rule: use source 2
4.  $\theta_1 = 1 \Rightarrow$  stop and choose  $a_1$ ;  $\theta_2 = 1 \Rightarrow$  stop and choose  $a_2$   
▶ phase 1 rule: use the source with the highest index

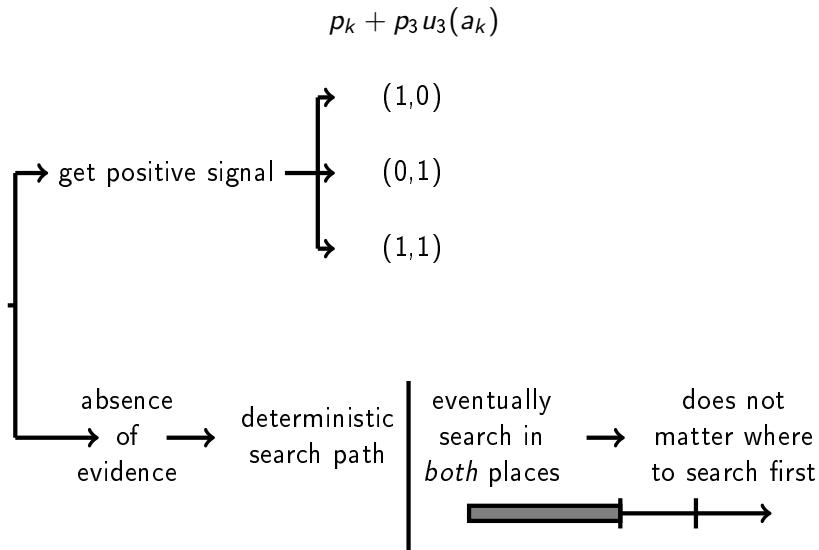
## Theorem

Phase 1 rule: use the source with the highest index

$$p_k + p_3 u_3(a_k)$$

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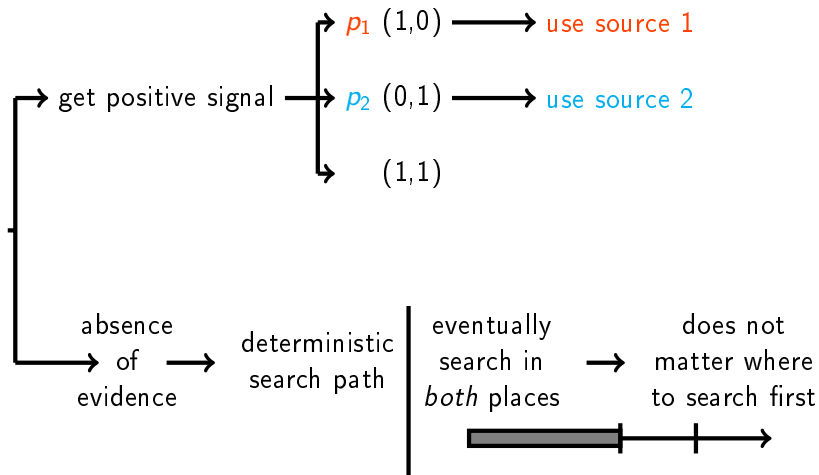


## Theorem

Phase 1 rule: use the source with the highest index

source 1:  $p_1$

source 2:  $p_2$

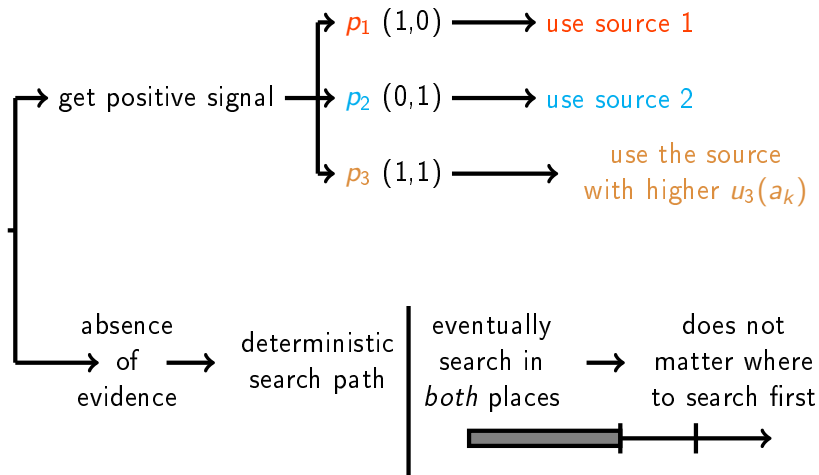




## Theorem

Phase 1 rule: use the source with the highest index

source 1:  $p_1 + p_3 u_3(a_1)$     source 2:  $p_2 + p_3 u_3(a_2)$



## Takeaway

The index depends only on the optimal actions in case of a breakthrough (state  $\theta_k$  is revealed).

Why is it "good news"?

*Examples:*

- ▶ information sources = different diagnostic tests;  
decision problem also specifies possible treatments  
*How much do we need to know about possible treatments  
to know which test to apply?*

## Takeaway

The index depends only on the optimal actions in case of a breakthrough (state  $\theta_k$  is revealed).

## Why is it "good news"?

*Examples:*

- ▶ information sources = different diagnostic tests;  
decision problem also specifies possible treatments  
*How much do we need to know about possible treatments  
to know which test to apply?*
- ▶ city council decides on the fate of some area;  
this area is a critical habitat for a bird and a promising place  
for oil drilling;  
information sources = search for oil / a bird;  
problem: **disagreement** in the council  
*All the council has to agree on is what to do if the bird or oil  
is found in another place*

## Literature

Che&Mierendorff  
(2017, working paper)

Nikandrova&Pancs  
(2018, TE)

This paper

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states	(1,0) and (0,1)	(1,0), (0,1), (1,1), (0,0)	
correlation	$\theta_1 + \theta_2 = 1$	$\theta_1 \perp\!\!\!\perp \theta_2$	arbitrary
actions	2 ("match the state")	2 ("choose successful project")	any number
focus	on a given decision problem ↓ optimal strategy (including chosen action)		on the general form of the learning process ↓ index

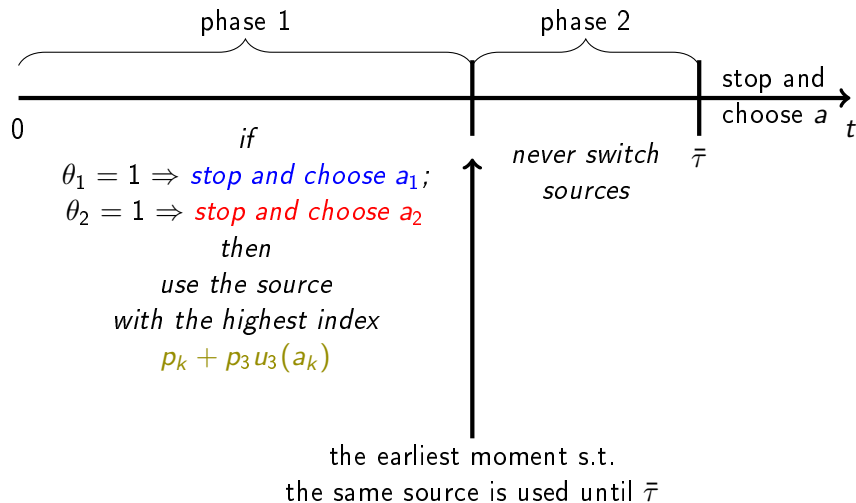
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# Literature

## Existing literature: actions and states are tightly connected

- ▶ Multi-armed bandit problem: actions  $\sim$  states (info about actions directly, not through a state of the world)  
*Pancs and Nikandrova 2018, Che and Mierendorff 2017, Klein and Rady 2011*
- ▶ Search problem
  - ▶ minimize cost of learning conditional on finding the state  
*Ahlswede&Wegener 1987*
  - ▶ or finding the state has a direct consequences to the payoff (treasure hunt)  
*Matros&Smirnov 2016, Fershtman&Rubinstein 1997*
- ▶ *Liang, Mu, and Syrgkanis (2017)* came to the “opposite” conclusion that myopic learning is (almost) optimal
  - ▶ completely different structure of the decision problem; in particular, only one state component,  $\theta_1$ , is payoff relevant

## Optimal Strategy



Reminder:  $p_1 = \mathbb{P}[(1,0)]$ ,  $p_2 = \mathbb{P}[(0,1)]$ ,  $p_3 = \mathbb{P}[(1,1)]$ ,  $u_3$ : payoff at (1,1)