## Lesson 13. Introduction to Stochastic Dynamic Programming

## 1 Motivation

- In the dynamic programs we have studied so far, the transitions from one state to the next are deterministic
- For example, the knapsack problem:
- Suppose we are in stage $t$ and state $n$ (deciding whether to take metal $t$ with $n \mathrm{~kg}$ of space remaining)
- If we decide to take metal $t$ in stage $t$, we know exactly what state we will be in stage $t+1$ :
-What if the transitions between states are subject to some randomness or stochasticity?


## 2 A production and inventory problem with stochastic demand

Example 1. The Dijkstra Brewing Company is planning production of its new limited run beer, Primal Pilsner, over the next 2 months. Based on some market analysis studies, the company has determined that the demand for the new beer in each month will be:

| Demand (batches) | Probability |
| :---: | :---: |
| 0 | $1 / 4$ |
| 2 | $3 / 4$ |

Each batch of beer costs $\$ 3,000$ to produce. Batches can be held in inventory at a cost of $\$ 1,000$ per batch per month. Each month, the company can produce either 0 or 1 batches, due to capacity limitations. In addition, the size of the company's warehouse restricts the ending inventory for each month to at most 2 batches. The company has 1 batch ready to go in inventory.
Due to contractual obligations, there is a penalty of $\$ 5,000$ for each batch of demand not met. Any batches produced that cannot be stored in the company's warehouse gets thrown away, and cannot be used to meet future demand.

The company wants to find a production plan that will minimizes its total production and holding costs over the next 3 months.

### 2.1 Modeling the problem visually

- Let's think about the decision-making process starting at month 1
- Let:
- Node $t_{n}$ represent month $t$ with $n$ batches in inventory
- $x_{t}$ represent the number of batches to produce in month $t$
- $d_{t}$ represent the number of batches in demand in month $t$
- We can draw the following diagram (that looks like a graph) that models the decision-making process

$$
x_{1}=1
$$

$1_{1}$
$2{ }_{1}$


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- We can diagram the entire 2-month process in a similar fashion:

$$
x_{1}=1
$$

$1_{2}$

$$
x_{1}=0
$$

$\begin{array}{ll}2 & \left.\begin{array}{l}2 \\ \hline\end{array}\right)\end{array}$

$$
x_{1}=1
$$

$\omega$
$1_{1}$

$$
x_{1}=0
$$

$$
x_{2}=1
$$

$$
\begin{array}{|c|}
\hline 2 \\
\hline
\end{array}
$$

$$
3_{1}
$$

$$
x_{1}=1
$$

$$
x_{2}=1
$$

$1_{0}$

$$
x_{1}=0
$$

$$
2
$$

$$
3_{0}
$$

- Consider the following production policy:
- In month 1, produce 1 batch
- In month 2 :
$\diamond$ If there are 2 batches in inventory, produce 0 batches
$\diamond$ If there are 0 batches in inventory, produce 1 batch
- What is the expected cost of this policy?
- Working backwards:
- Expected cost in month 2 with 2 batches in inventory (node $2_{2}$ ):
$\square$
- Expected cost in month 2 with 0 batches in inventory (node $2_{0}$ ):
- Expected cost in month 1 (node $1_{1}$ ):


### 2.2 Things to think about

- The policy above gives contingency plans
- The diagram we drew on page 3 sort of looks like a shortest path problem, but it's not!
- We cannot solve this example as a shortest path problem, since the edges "are random"
- We can, however, still write a recursion to represent this example problem, and others like it
- We'll explore this next...


### 2.3 Writing down the model

- What we really want: a production policy with minimum expected cost
- Let's write down the recursive representation of our model/diagram
- We can then solve this recursion by working backwards and determine the minimum expected cost and associated optimal policy
- Stages:
$\square$
- States:
- Transition probability $p\left(m \mid n, t, x_{t}\right)$ of moving from state $n$ to state $m$ in stage $t$ under decision $x_{t}$ :
$\square$
- Contribution $c\left(m \mid n, t, x_{t}\right)$ of moving from state $n$ to state $m$ in stage $t$ under decision $x_{t}$ :
- Allowable decisions $x_{t}$ at stage $t$ and state $n$ :
- In words, the value-to-go $f_{t}(n)$ at stage $t$ and state $n$ is:
- Boundary conditions:
- Value-to-go recursion:
- Desired value-to-go function value:


### 2.4 Interpreting the value-to-go function

- We can solve this recursion just like with a deterministic DP: start at the boundary conditions and work backwards
- For this problem, we get the following value-to-go function values $f_{t}(n)$ for $t=1,2$ and $n=0,1,2$, as well as the decision $x_{t}^{*}$ that attained each value:

| $t$ | $n$ | $f_{t}(n)$ | $x_{t}^{*}$ |
| :---: | :---: | :---: | :---: |
| 1 | 0 | 13.125 | 1 |
| 1 | 1 | 8.875 | 1 |
| 1 | 2 | 5.875 | 0 |
| 2 | 0 | 7 | 1 |
| 2 | 1 | 3.5 | 1 |
| 2 | 2 | 0.5 | 0 |

- Based on this, what should the company's policy be?
- What is the company's total expected cost?


## 3 Stochastic dynamic programs, more generally

## Stochastic dynamic program

- Stages $t=1,2, \ldots, T$ and states $n=0,1,2, \ldots, N$
- Allowable decisions $x_{t}$ at each stage $t$ and state $n$
- Transition probability $p\left(m \mid n, t, x_{t}\right)$ of moving from state $n$ to state $m$ in stage $t$ under decision $x_{t}$
- Contribution $c\left(m \mid n, t, x_{t}\right)$ for moving from state $n$ to state $m$ in stage $t$ under decision $x_{t}$

- Value-to-go function $f_{t}(n)$ at each stage $t$ and state $n$
- Boundary conditions on $f_{T}(n)$ for each state $n$
- Recursion on $f_{t}(n)$ at stage $t$ and state $n$

$$
\begin{aligned}
f_{t}(n) & =\min _{x_{t} \text { allowable }}\left\{\sum_{m=0}\left\{\left(m \mid n, t, x_{t}\right)\left[c\left(m \mid n, t, x_{t}\right)+f_{t+1}(m)\right]\right\}\right. \\
& \text { for } t=1,2, \ldots, T-1 \text { and } n=0,1, \ldots, N
\end{aligned}
$$

- Desired value-to-go, usually $f_{1}(m)$ for some state $m$


## 4 A precision manufacturing problem

Example 2. The Hit-and-Miss Manufacturing Company has received an order to supply one item of a particular type. However, manufacturing this item is difficult, and the customer has specified such stringent quality requirements that the company may have to produce more than one item to obtain an item that is acceptable.
The company estimates that each item of this type will be acceptable with probability $1 / 2$ and defective with probability $1 / 2$. Each item costs $\$ 100$ to produce, and excess items are worthless. In addition, a setup cost of $\$ 300$ must be incurred whenever the production process is setup for this item. The company has time to make no more than 3 production runs, and at most 5 items can be produced in each run. If an acceptable item has not been obtained by the end of the third production run, the manufacturer is in breach of contract and must pay a penalty of $\$ 1600$.

The objective is to determine how many items to produce in each production run in order to minimize the total expected cost.

### 4.1 Warm up

- Suppose the manufacturer produces $x$ items in a single production run.
- What is the probability that at least one of these items is acceptable?
- What is the expected number of acceptable items?


### 4.2 Modeling the problem

- Stages:
$\square$
- States:
- Allowable decisions $x_{t}$ at stage $t$ and state $n$ :
$\square$
- Sketch of basic structure:
- When the state $n=1$ :

- When the state $n=0$ :

- In words, the value-to-go $f_{t}(n)$ at stage $t$ and state $n$ is:
- Boundary conditions:
- Value-to-go recursion:

$$
f_{t}(n)=\min _{x_{t} \text { allowable }}^{\max }\left\{\sum_{m \text { state }} p\left(m \mid n, t, x_{t}\right)\left[c\left(m \mid n, t, x_{t}\right)+f_{t+1}(m)\right]\right\} \quad \text { for stages } t \text { and states } n
$$

$\square$

- Desired value-to-go function value:


### 4.3 Interpreting the value-to-go function

- Solving the recursion, we get the following value-to-go function values $f_{t}(n)$ for $t=1,2,3$ and $n=0,1$, as well as the decision $x_{t}^{*}$ that attained each value:

| $t$ | $n$ | $f_{t}(n)$ | $x_{t}^{*}$ |
| :---: | :---: | :---: | :---: |
| 1 | 0 | 0 | 0 |
| 1 | 1 | 675 | 2 |
| 2 | 0 | 0 | 0 |
| 2 | 1 | 700 | 2 |
| 3 | 0 | 0 | 0 |
| 3 | 1 | 800 | 3 |

- Based on this, what should the company's policy be?
- What is the company's total expected cost?


## 5 A small investment problem

Example 3. Suppose you have $\$ 5,000$ to invest. Over the next 3 years, you want to double your money. At the beginning of each of the next 3 years, you have an opportunity to invest in one of two investments: A or B. Both investments have uncertain profits. For an investment of $\$ 5,000$, the profits are as follows:

| Investment | Profit (\$) | Probability |
| :---: | ---: | ---: |
| A | $-5,000$ | 0.3 |
|  | 5,000 | 0.7 |
| B | 0 | 0.9 |
|  | 5,000 | 0.1 |

You are allowed to make at most one investment each year, and can invest only $\$ 5,000$ each time. Any additional money accumulated is left idle. Once you've accumulated $\$ 10,000$, you stop investing.

Formulate a stochastic dynamic program to find an investment policy that maximizes the probability you will have $\$ 10,000$ after 3 years.

### 5.1 Warm up

Consider the following investment policy. What is the probability of having at least $\$ 10,000$ ?


### 5.2 Formulating the stochastic dynamic program

- Stages:

- States:
- Allowable decisions $x_{t}$ at stage $t$ and state $n$ :
- Sketch of basic structure - transition probabilities and contributions:
- In words, the value-to-go $f_{t}(n)$ at stage $t$ and state $n$ is:
- Value-to-go recursion

$$
f_{t}(n)=\min _{x_{t} \text { allowable }} \max \left\{\sum_{m \text { state }} p\left(m \mid n, t, x_{t}\right)\left[c\left(m \mid n, t, x_{t}\right)+f_{t+1}(m)\right]\right\} \quad \text { for stages } t \text { and states } n
$$

$\square$

- Boundary conditions:
- Desired value-to-go function value:


### 5.3 Interpreting the value-to-go function

- Solving the recursion on $f_{t}(n)$, we obtain:

| $t$ | $n$ | $f_{t}(n)$ | $x_{t}^{*}$ |
| :---: | :---: | :---: | :---: |
| 1 | 0 | 0 | no investment |
| 1 | 5000 | 0.757 | B |
| 1 | 10000 | 1 | no investment |
| 2 | 0 | 0 | no investment |
| 2 | 5000 | 0.73 | B |
| 2 | 10000 | 1 | no investment |
| 3 | 0 | 0 | no investment |
| 3 | 5000 | 0.7 | A |
| 3 | 10000 | 1 | no investment |

- Based on this, what should your investment policy be?

- What is your probability of having $\$ 10,000$ ?


## A Problems

Problem 1 (Baytheon). Baytheon has received an order to supply 2 guided missiles. In order to meet stringent quality requirements, the company may have to manufacture more than one missile to obtain an missile that is acceptable. The company has time to make no more than 3 production runs, and at most 2 missiles can be produced in each run. The probability distribution of acceptable missiles in a given run depends on how many missiles are produced:

|  | Probability of acceptable missiles |  |  |
| :---: | :---: | :---: | :---: |
| Number of missiles produced | 0 | 1 | 2 |
| 0 | 1 | 0 | 0 |
| 1 | $1 / 3$ | $2 / 3$ | 0 |
| 2 | $1 / 4$ | $1 / 2$ | $1 / 4$ |

Each missile costs $\$ 100,000$ to produce, and excess missiles are worthless. In addition, a setup cost of $\$ 50,000$ must be incurred whenever the production process is setup for this item. If 2 acceptable missiles have not been obtained by the end of the third production run, Baytheon is in breach of contract and must pay a penalty of $\$ 25,000$. The objective is to determine how many missiles to produce in each production run in order to minimize the total expected cost.

Formulate this problem as a stochastic dynamic program.

Problem 2 (Farkas Investments). You have recently been hired as a junior analyst at Farkas Investments. You have been given $\$ 4$ million to invest over the next 3 years. At the beginning of each of the next 3 years, you can invest in one of two investments: A or B.

| Investment | Cost (\$ millions) | Profit (\$ millions) | Probability |
| :---: | ---: | ---: | ---: |
| A | 3 | 2 | 0.5 |
|  |  | -2 | 0.5 |
| B | 5 | 3 | 0.1 |
|  |  | -1 | 0.9 |

You are allowed to make at most one investment each year. Any additional money accumulated is left idle. You may not borrow money to invest; that is, you cannot buy into an investment if it costs more than you currently have.

Formulate a stochastic dynamic program to find an investment policy that maximizes the probability you will have at least $\$ 10$ million at the end of 3 years.

