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SUSTAINABLE ENERGY

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RESOURCE EVALUATION AND DEPLETION ANALYSES



WAYS OF ESTIMATING ENERGY RESOURCES

- Monte Carlo
- “Hubbert” Method Extrapolation
- Expert Opinion (Delphi)

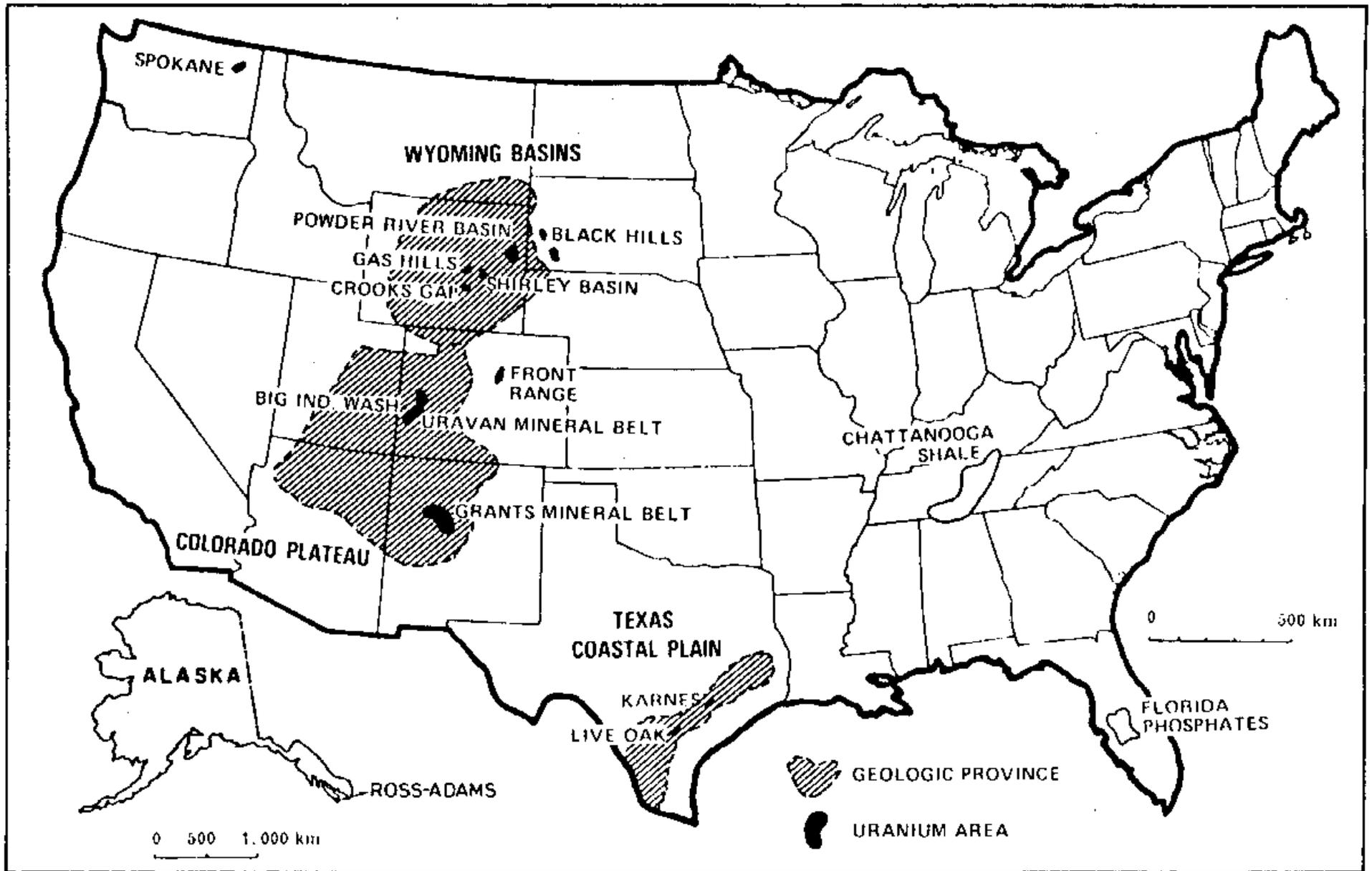


FACTORS AFFECTING RESOURCE RECOVERY

- Nature of Deposit
- Fuel Price
- Technological Innovation
 - Deep drilling
 - Sideways drilling
 - Oil and gas field pressurization
 - Hydrofracturing
 - Large scale mechanization



URANIUM AREAS OF THE U.S.



Courtesy of U.S. Atomic Energy Commission.



MAJOR SOURCES OF URANIUM

Class 1 – Sandstone Deposits

	Share	U ³ O ⁸ Concentration (Percent)	Tons U ³ O ⁸
New Mexico	.49	0.25	Total
Wyoming	.36	0.20	315,000
Utah	.03	0.32	\$ \$10/lb
Colorado	.03	0.28	
Texas	.06	0.28	
Other	.03	0.28	

Class 2 – Vein Deposits

7,100

Class 3 – Lignite Deposits

0.01-0.05

1,200

Class 4 – Phosphate Rock

0.015

Class 5 – Phosphate Rock Leached Zone (Fla.)

0.010

54,600

Class 6 – Chattanooga Shale

0.006

2,557,300

Class 7 – Copper Leach Solution Operations

0.0012

30,000

Class 8 – Conway Granite

0.0012-Uranium

1x10⁶

0.0050-Thorium

4x10⁶

Class 9 – Sea Water

0.33x10⁻⁶

4x10⁹



ESTIMATES OF URANIUM AVAILABILITY FROM GEOLOGICAL FORMATIONS AND OCEANS IN THE U.S.

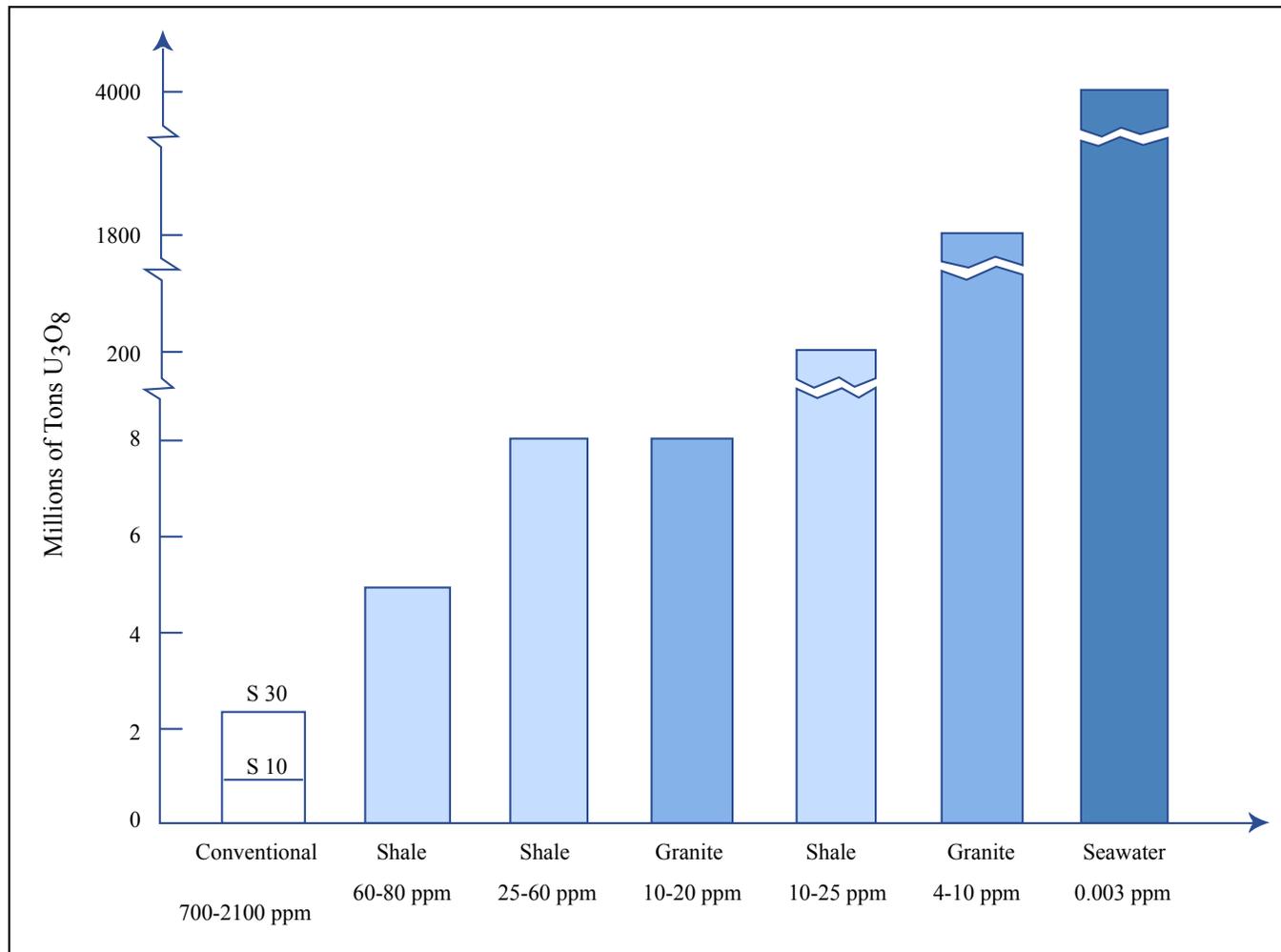


Image by MIT OpenCourseWare.



DECLINE IN GRADE OF MINED COPPER ORES SINCE 1925

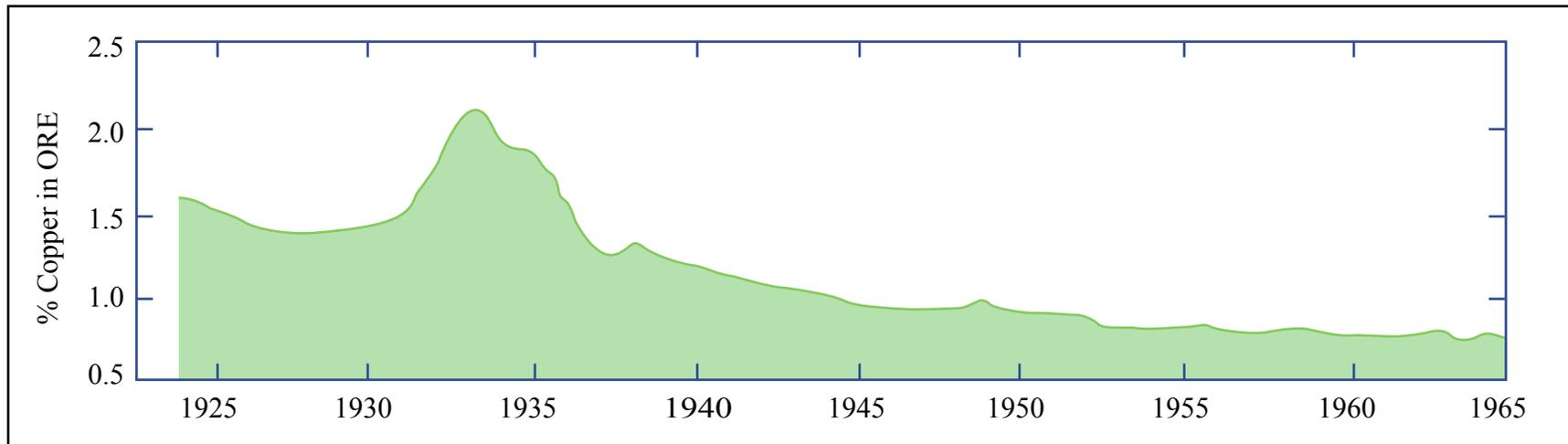
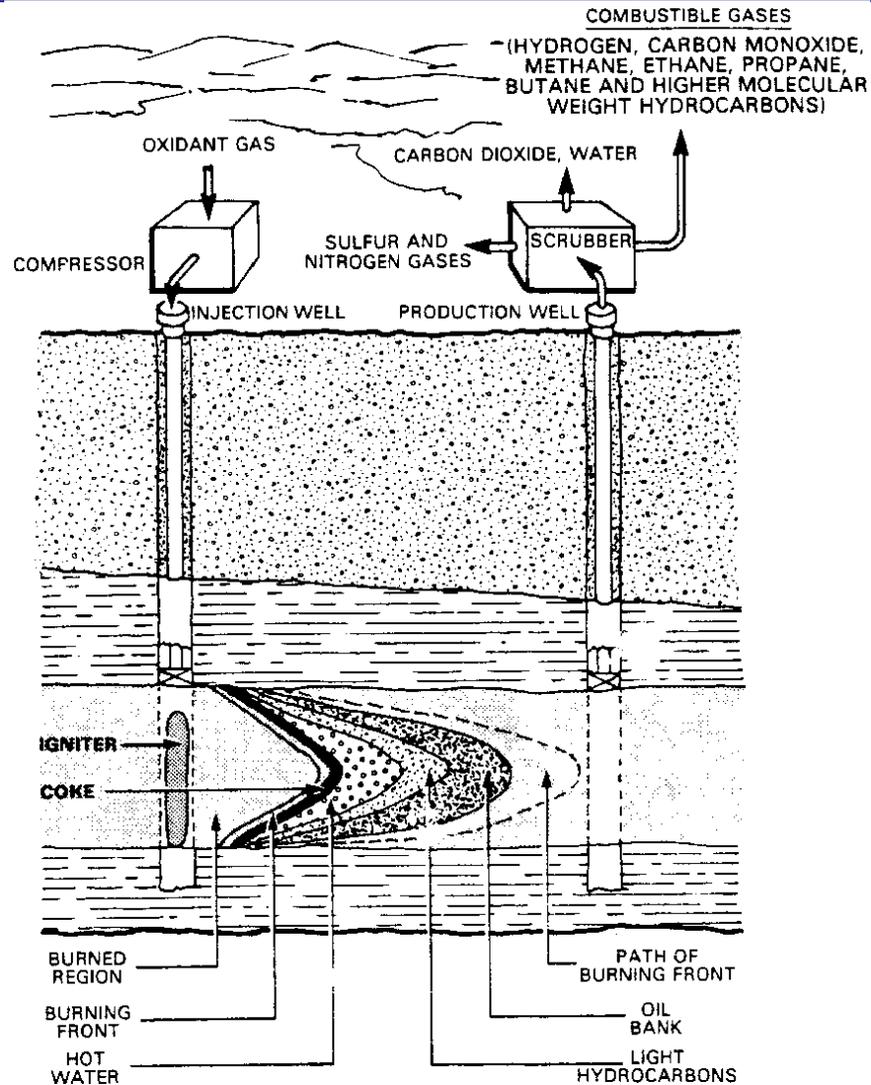


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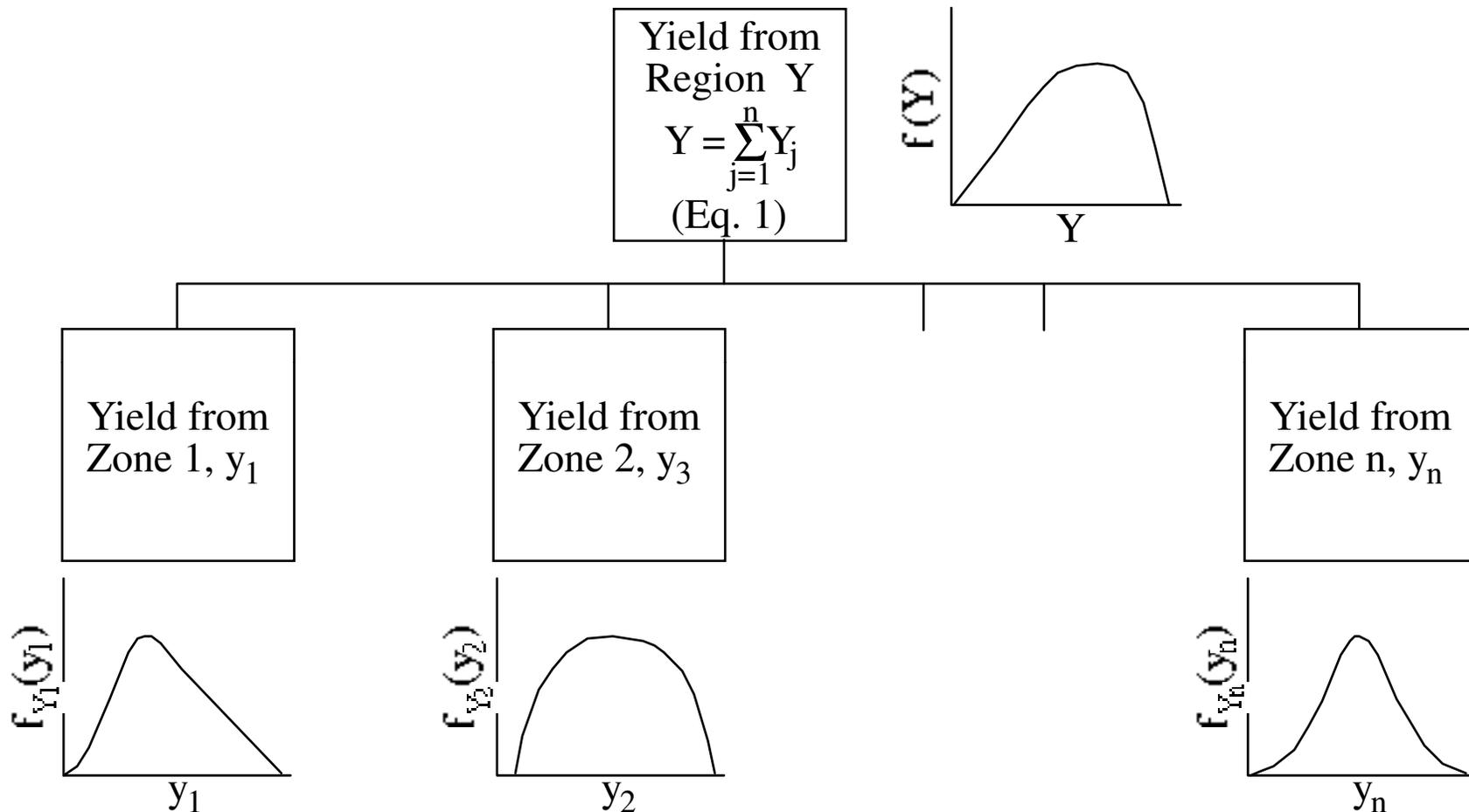
RECOVERY BY IN-SITU COMBUSTION



Source: U.S. Department of Energy, "Fossil Energy Research and Development Program of the U.S. Department of Energy, FY 1979," DOE/ET-0013(78), March 1978.



MONTE CARLO ESTIMATION



Probability density functions are obtained subjectively, using information about deposit characteristics, fuel price, and technology used.



MONTE CARLO ESTIMATION OF THE PROBABILITY DENSITY FUNCTION OF A FUNCTION OF A SET OF RANDOM VARIABLES, AS

$$G = G(\bar{Z}), \text{ where} \quad (\text{Eq. 1})$$

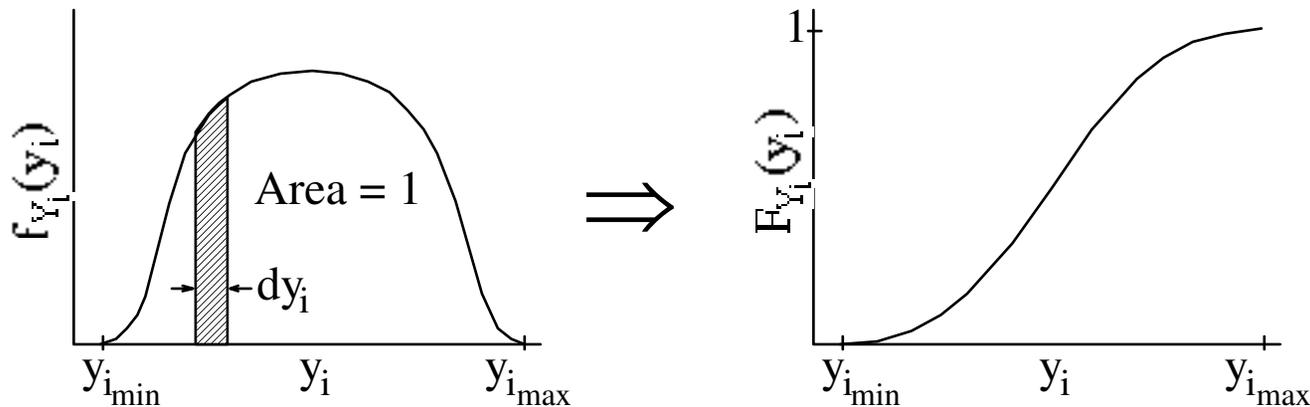
$$\bar{Z} = [y_1, y_2, \dots, y_n] \text{ and}$$

Y_i is a random variable ($i = 1, n$)

Note that \bar{Z} and G are also random variables.



MONTE CARLO ESTIMATION OF PROBABILITY DENSITY AND CUMULATIVE DISTRIBUTION FUNCTIONS



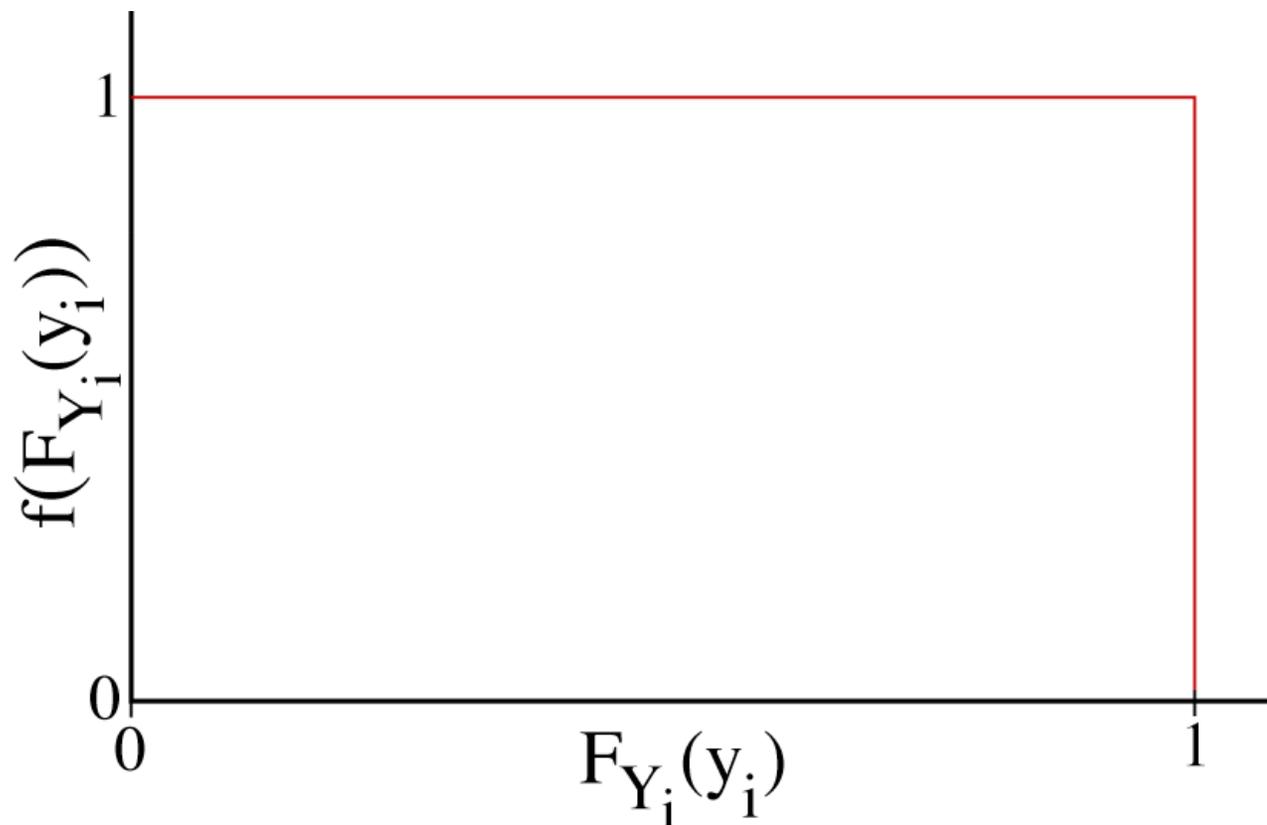
$$\text{Prob.}(y_i < Y_i < y_i + \delta y_i) = f_{Y_i}(y_i) dy_i \quad (\text{Eq. 2}) \quad \text{Prob.}(Y_i < y_i) = F_{Y_i}(y_i) = \int_{y_{i,\min}}^{y_i} f_{Y_i}(y'_i) dy'_i \quad (\text{Eq. 3})$$

Consider Y_i to be a random variable within $[y_{i,\min}, y_{i,\max}]$



MONTE CARLO ESTIMATION, Continued

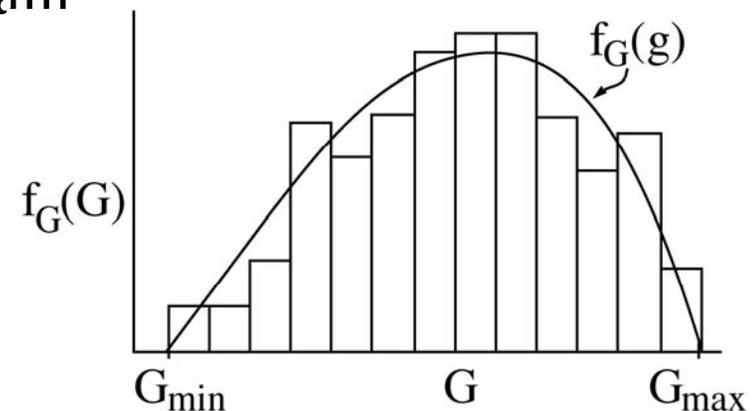
Note: $F_{Y_i}(y_i)$ is uniformly distributed within $[0, 1]$





MONTE CARLO ESTIMATION, Continued

1. Utilize a random number generator to select a value of $F(y_i)$ within range $[0, 1] \Rightarrow$ corresponding value of y_i (Eq. 3).
2. Repeat step 1 for all values of i and utilize selected values of $\bar{Z}_1 = [y_{1_1}, y_{2_1}, \dots, y_{n_1}]$ to calculate a value of \bar{Z}_1 (Eq. 1) (note \bar{Z} is also a random variable).
3. For the k -th set of selected values of $\bar{Z}_K = [y_{1_K}, y_{2_K}, \dots, y_{n_K}]$ can obtain the corresponding value of $G_K = G(\bar{Z}_K)$
4. Repeat step 2 many times and obtain a set of values of vector \bar{Z} , and corresponding value of G_k .
5. Their abundance distributions will approximate those of the pdfs of the variables \bar{Z} and $G(\bar{Z})$ as





M. KING HUBBERT'S MINERAL RESOURCE ESTIMATION METHOD

ASSUMED CHARACTERISTICS OF MINERAL RESOURCE EXTRACTION

- As More Resource Is Extracted The Grade Of The Marginally Most Attractive Resources Decreases, Causing
 - Need for improved extraction technologies
 - Search for alternative deposits, minerals
 - Price increases (actually, rarely observed)



M. KING HUBBERT'S MINERAL RESOURCE ESTIMATION METHOD, Continued

POSTULATED PHASES OF MINERAL RESOURCE EXTRACTION

- Early: Low Demand, Low Production Costs, Low Innovation
- Growing: Increasing Demand And Discovering Rate, Production Growing With Demand, Start of Innovation
- Mature: Decreasing Demand And Discovery Rate, Production Struggling To Meet Demand, Shift To Alternatives
- Late: Low Demand, Production Difficulties, Strong Shift To Alternatives (rarely observed)

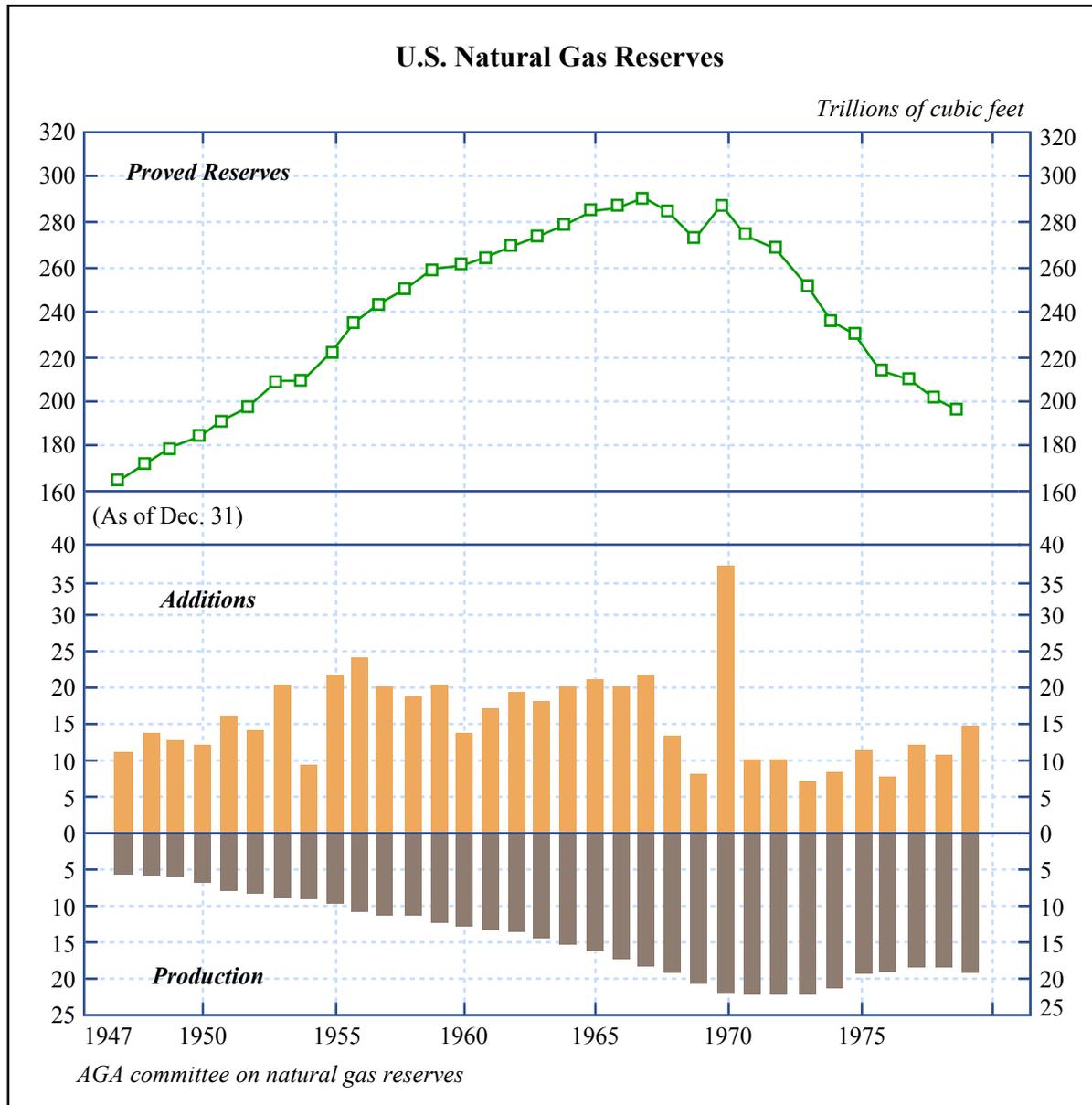
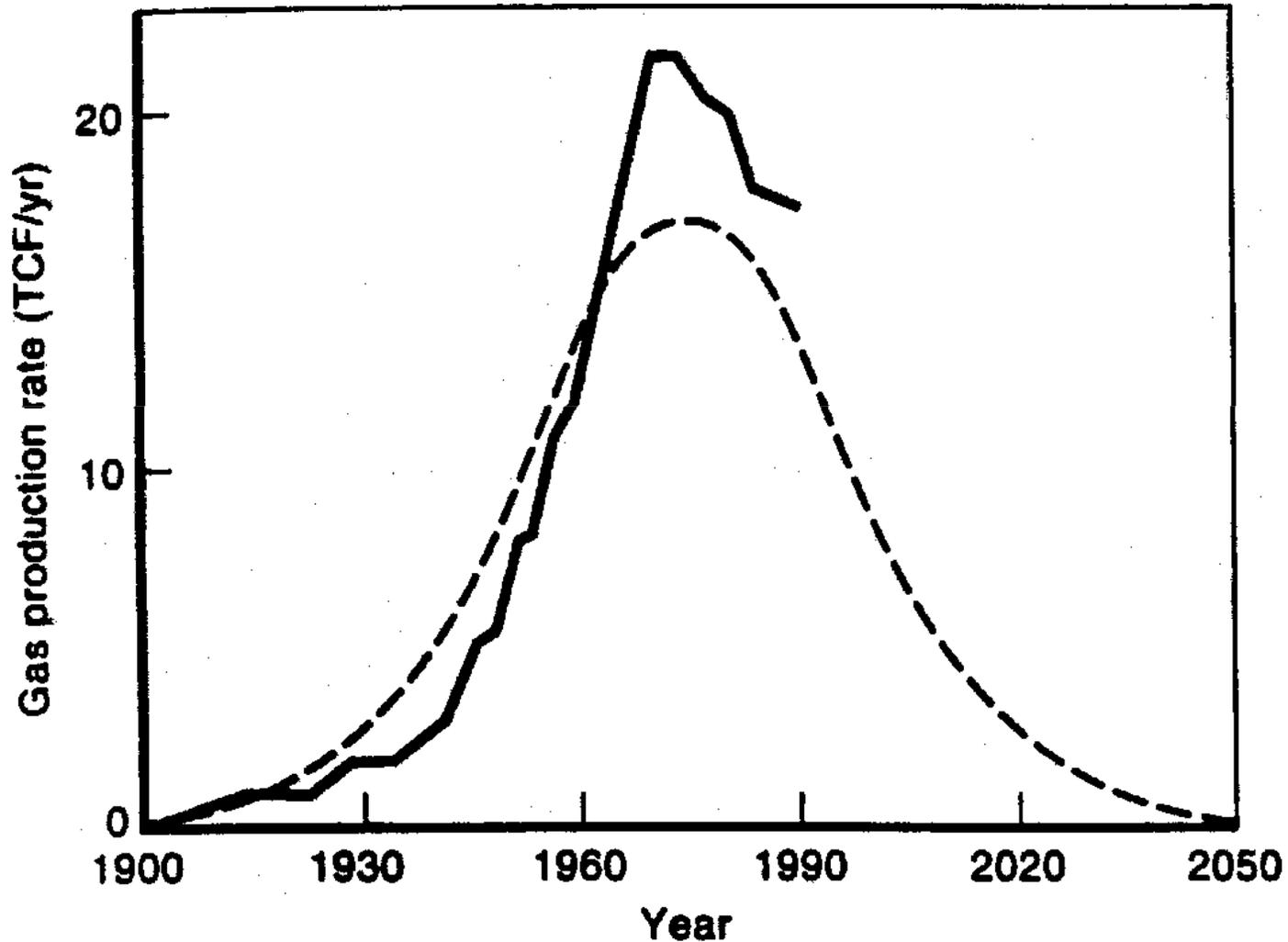


Image by MIT OpenCourseWare. Adapted from the American Gas Association.



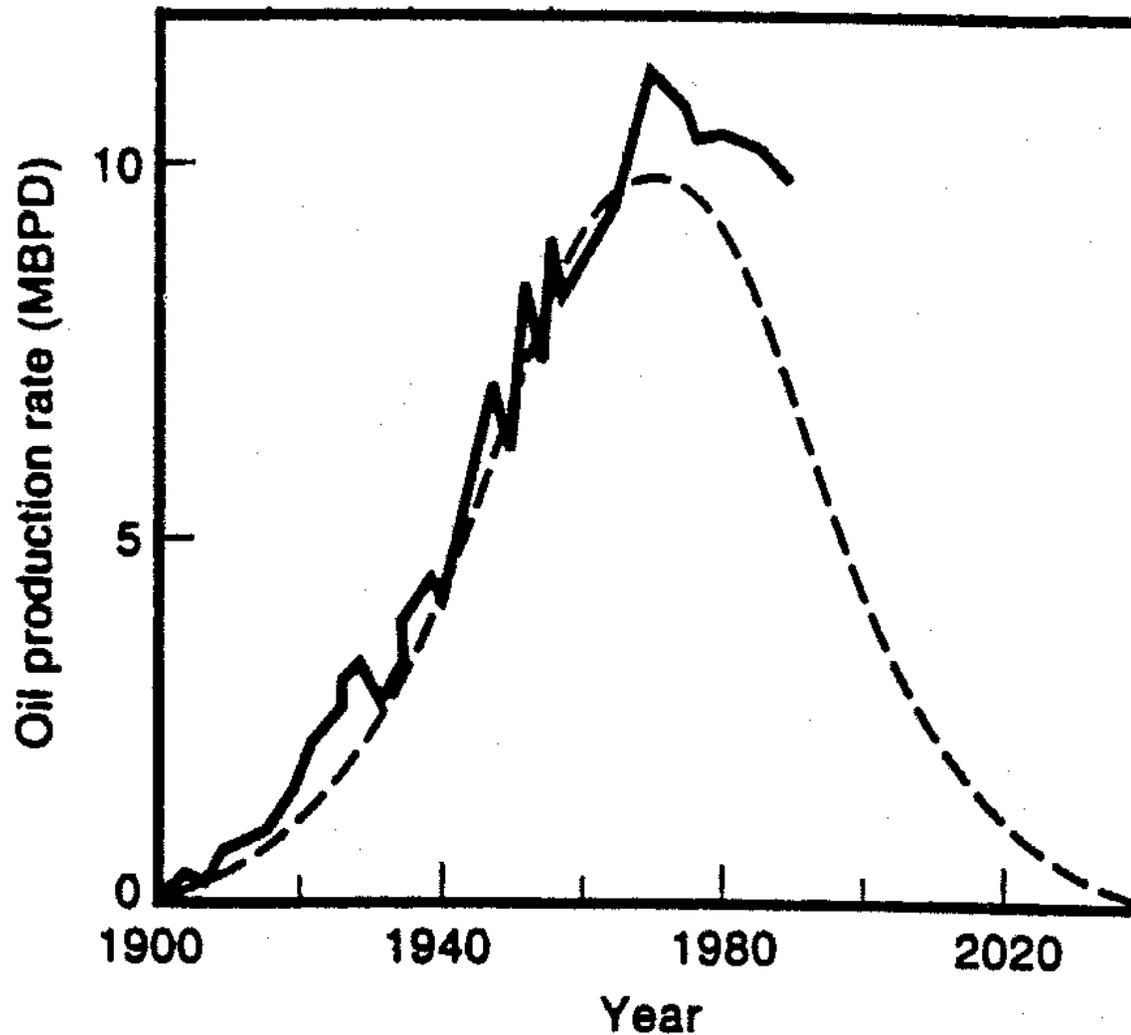
U.S. NATURAL GAS PRODUCTION



Comparison of estimated (Hubbert) production curve and actual production (solid line).



U.S. CRUDE OIL PRODUCTION



Comparison of estimated (Hubbert) production curve and actual production (solid line).



COMPLETE CYCLE OF WORLD CRUDE-OIL PRODUCTION

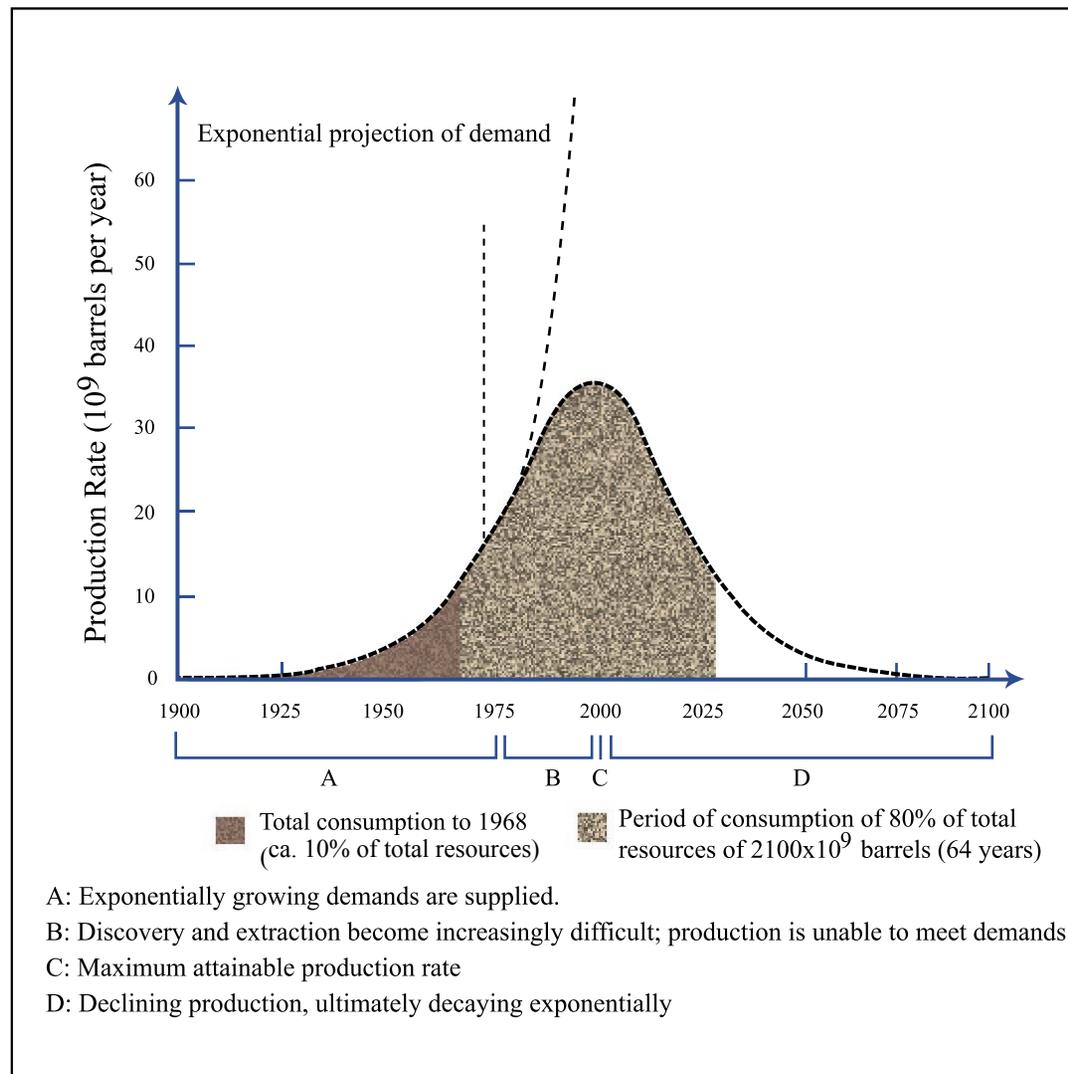
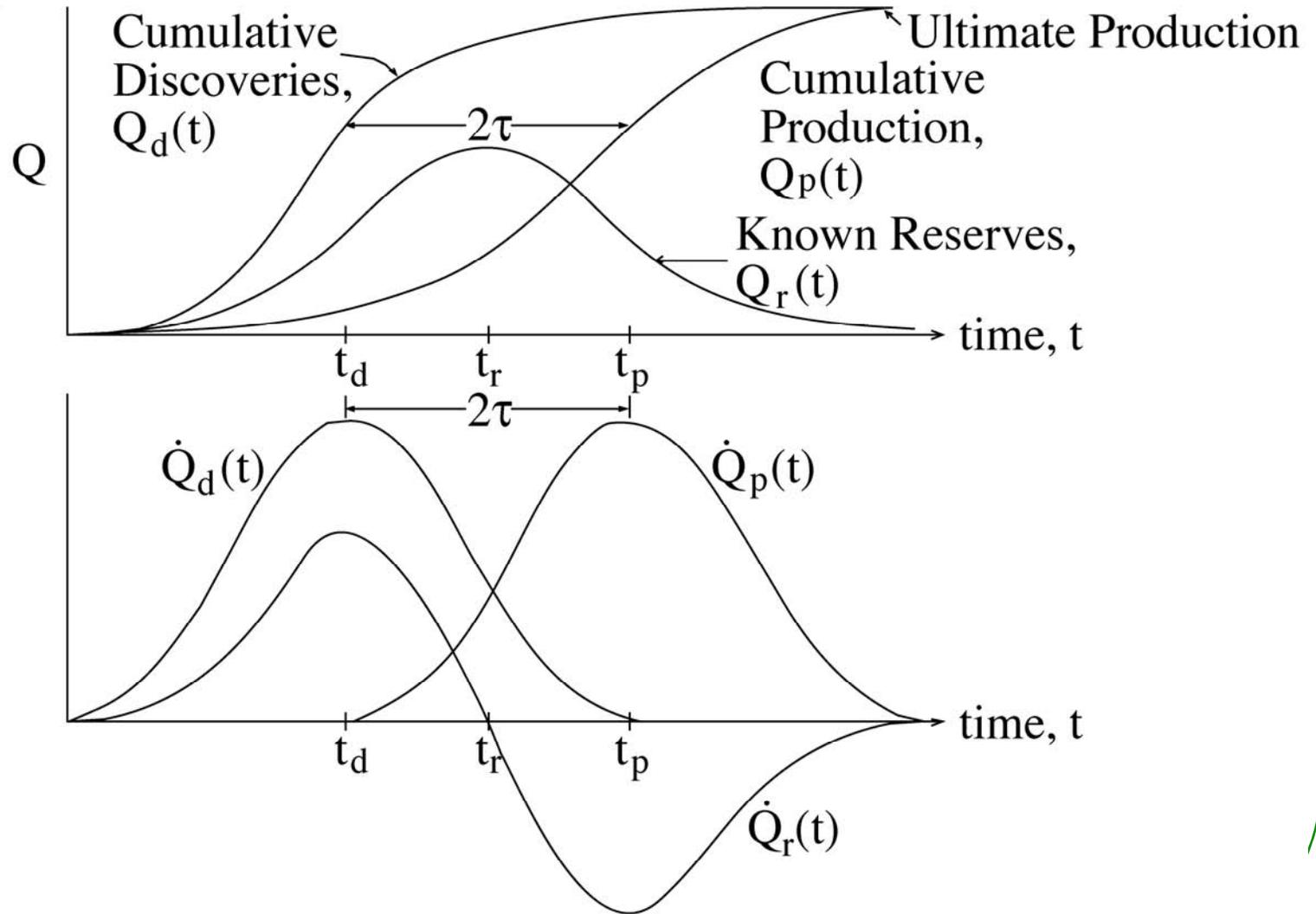


Image by MIT OpenCourseWare.



RESOURCE BEHAVIOR UNDER “HUBBERT” ASSUMPTIONS



Timing:

t_d, t_r, t_p
respectively

Q_d, Q_r, Q_p .



LOGISTIC FUNCTION

Hubbert's assumed "logistical" relationships

Rate of Production

$$\dot{Q}_P = \frac{d[Q_P(X)]}{dt} = rQ_P(X) \left[1 - \frac{Q_P(X)}{K} \right]$$

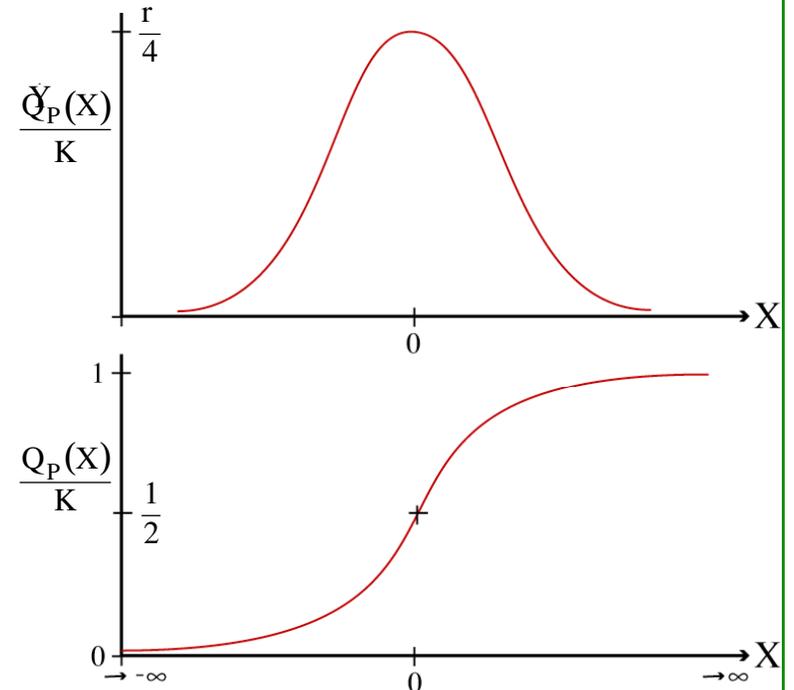
Cumulative Production

$$Q_P(X) = \frac{KQ_{P_0}(X)}{Q_{P_0}(X) - (Q_{P_0}(X) - K)e^{-rX}}$$

where r = relative rate of change
 K = carrying capacity or
ultimate production value

$$Q_{P_0} \equiv Q_P(X=0)$$

$Q_P(X)$ = cumulative production at time, $(t - t_0) = X$





LOGISTIC FUNCTION, continued

$$Q_P(-\infty) = 0 \quad Q_P(0) = \frac{K}{2} = Q_{P_0} \quad P(\infty) = K$$

$$\dot{Q}(X) \equiv \left. \frac{dQ_P(X)}{dt} \right|_{\max} = \left. \frac{dQ_P(X)}{dX} \right|_{X=0} = \frac{rK}{4}.$$

$$Q_P(X) \approx \int_{-\infty}^X \frac{K}{\sqrt{2\pi\sigma}} e^{-1/2\left(\frac{X'}{\sigma}\right)^2} dX', \text{ where } \sigma = \sqrt{\frac{8}{\pi r}}.$$

Let: $t - t_0 \equiv X$.



EQUATIONS

Conservation of Resource:

$$Q_d(t) = Q_r(t) + Q_p(t) \quad (\text{Eq. 4})$$

Rate Conservation:

$$\dot{Q}_d(t) = \dot{Q}_r(t) + \dot{Q}_p(t) \quad (\text{Eq. 5})$$

Approximate Results:

$$t(\dot{Q}_d = 0) - t(\dot{Q}_r = 0) = 2\tau \quad (\text{Eq. 6})$$

$$\tau \approx \begin{cases} (t_r - t_p) \\ (t_d - t_r) \end{cases} \quad (\text{Eq. 7})$$

or

$$t_r \approx \frac{1}{2} (t_d + t_p) \quad (\text{Eq. 8})$$

$$Q_{p \text{ ultimate}} \approx 2Q_d(t_d) \quad (\text{Eq. 9})$$



EQUATIONS, Continued

If we assume Gaussian distributions for $Q_r(t)$, $\dot{Q}_d(t)$ and $\dot{Q}_p(t)$ with each having the same standard deviation, σ , obtain

$$Q_r(t) = \frac{Q_{r_o}}{\sqrt{2\pi\sigma}} \exp \left[-\frac{1}{2} \left(\frac{t-t_r}{\sigma} \right)^2 \right] \quad (\text{Eq. 10})$$

$$\dot{Q}_d(t) = \frac{Q_{d_o}}{\sqrt{2\pi\sigma}} \exp \left[-\frac{1}{2} \left(\frac{t-t_d}{\sigma} \right)^2 \right] \quad (\text{Eq. 11})$$

$$\dot{Q}_p(t) = \frac{Q_{p_o}}{\sqrt{2\pi\sigma}} \exp \left[-\frac{1}{2} \left(\frac{t-t_p}{\sigma} \right)^2 \right] \quad , \quad (\text{Eq. 12})$$

Then, when Q_r is at a maximum $t = t_r$ and $\dot{Q}_r = 0$

$$\dot{Q}_r(t_r) = \frac{-1 \times Q_{r_o}}{\sigma^2} \Rightarrow \sigma^2 = \frac{-1 \times Q_r(t_r)}{\dot{Q}_r(t_r)} \quad , \text{ or} \quad (\text{Eq. 13})$$

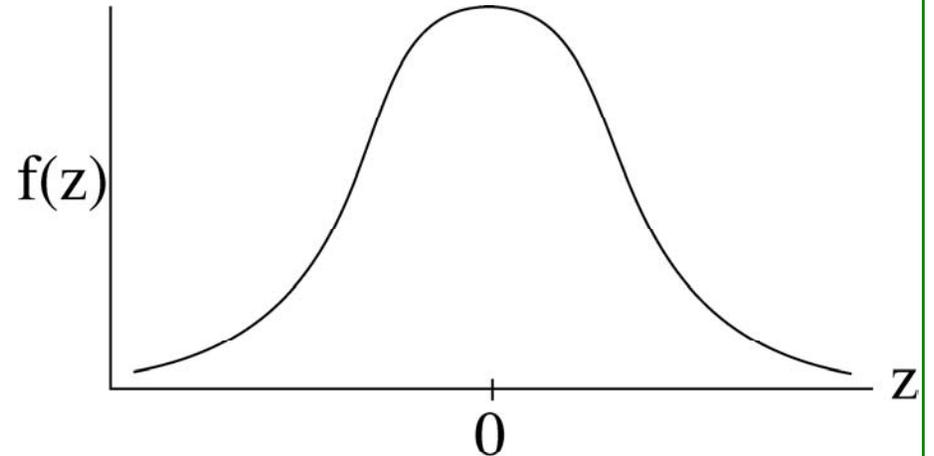


EQUATIONS, Continued

For the normal distribution

$$f(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}$$

$$f(1) = 0.67, \quad z \equiv \frac{t - t_0}{\sigma} = 1$$



$\Rightarrow \sigma \cong 25$ yr. for U.S. petroleum and natural gas

$F(z) = \int_{-\infty}^z f(z') dz'$, Cumulative distribution function

$F(3) = 0.99$, Approximately the state of full depletion

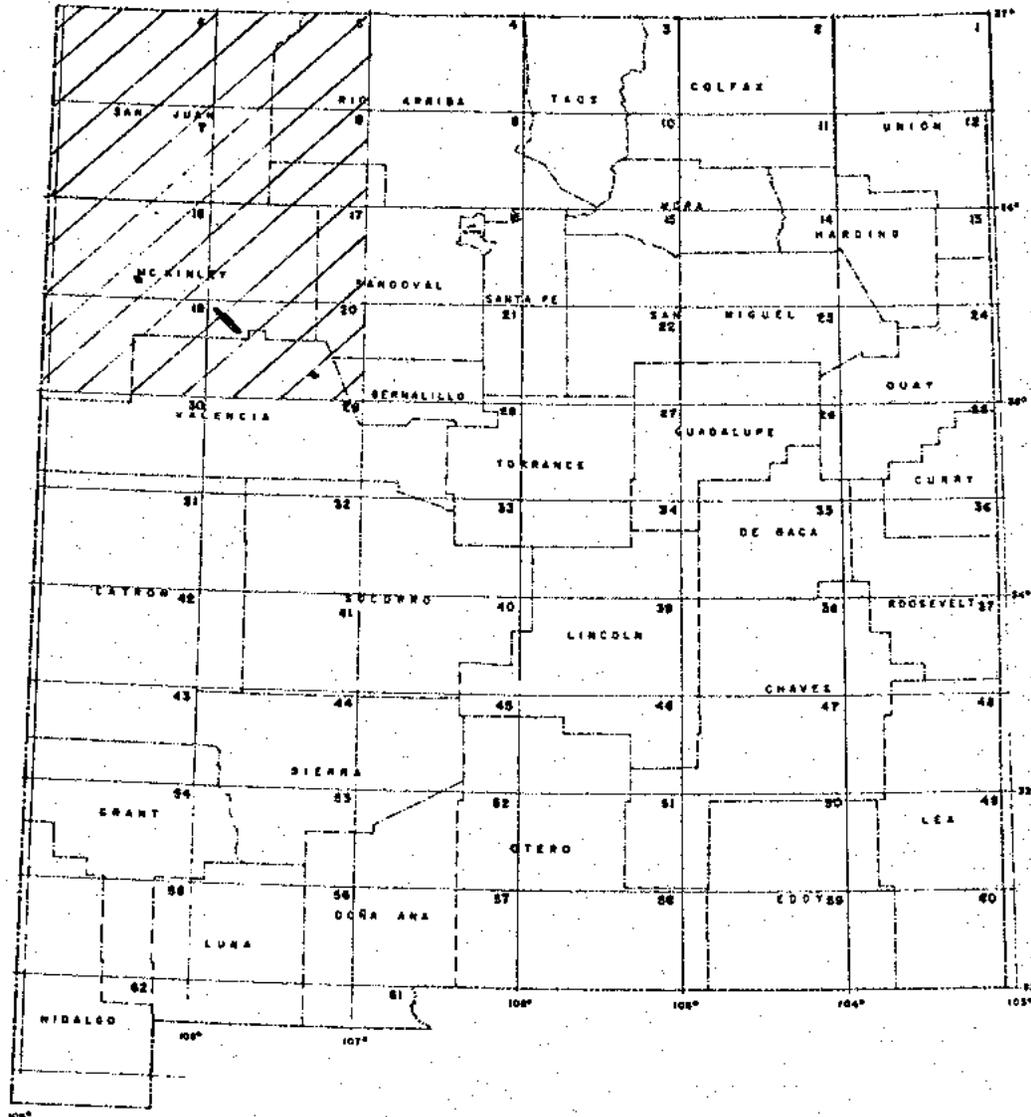
\Rightarrow Time of exploitation $\cong \sigma = 150$ yrs

\Rightarrow End date of major U.S. oil, natural gas production

$$= 1900 + 150 = \underline{\underline{2050}}$$



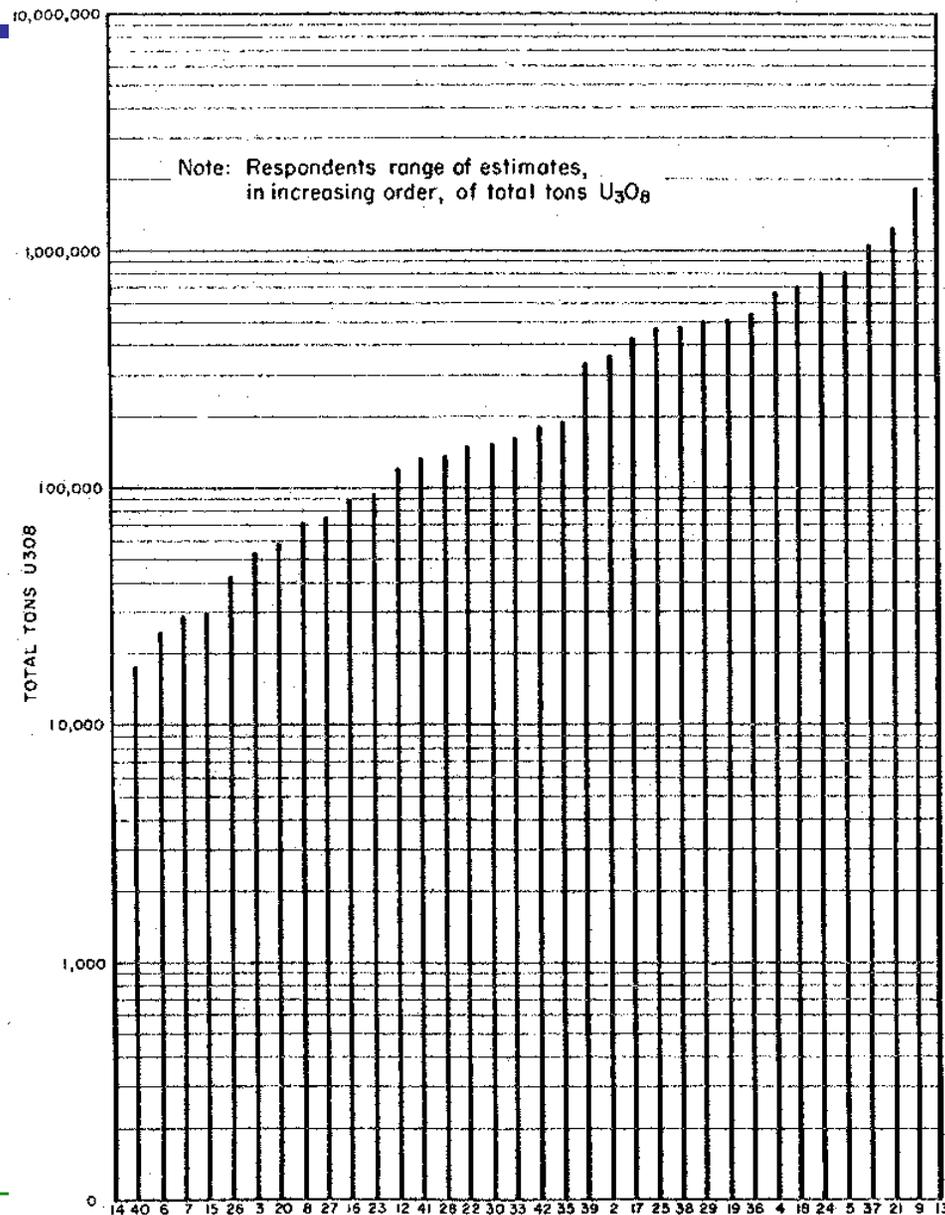
SUBJECTIVE PROBABILITY STUDY - STATE OF NEW MEXICO



Courtesy of U.S. Atomic Energy Commission.



NEW MEXICO SUBJECTIVE PROBABILITY STUDY (AFTER DELPHI)



Courtesy of U.S. Atomic Energy Commission.

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Introduction to Sustainable Energy

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