Applied Production Analysis

SOK-3011—Part 1

Course content

- Focus on production theory & empirical applications
  - Theoretical readings: Varian (1992)
  - Empirical readings: Henningsen (2024, R-based)

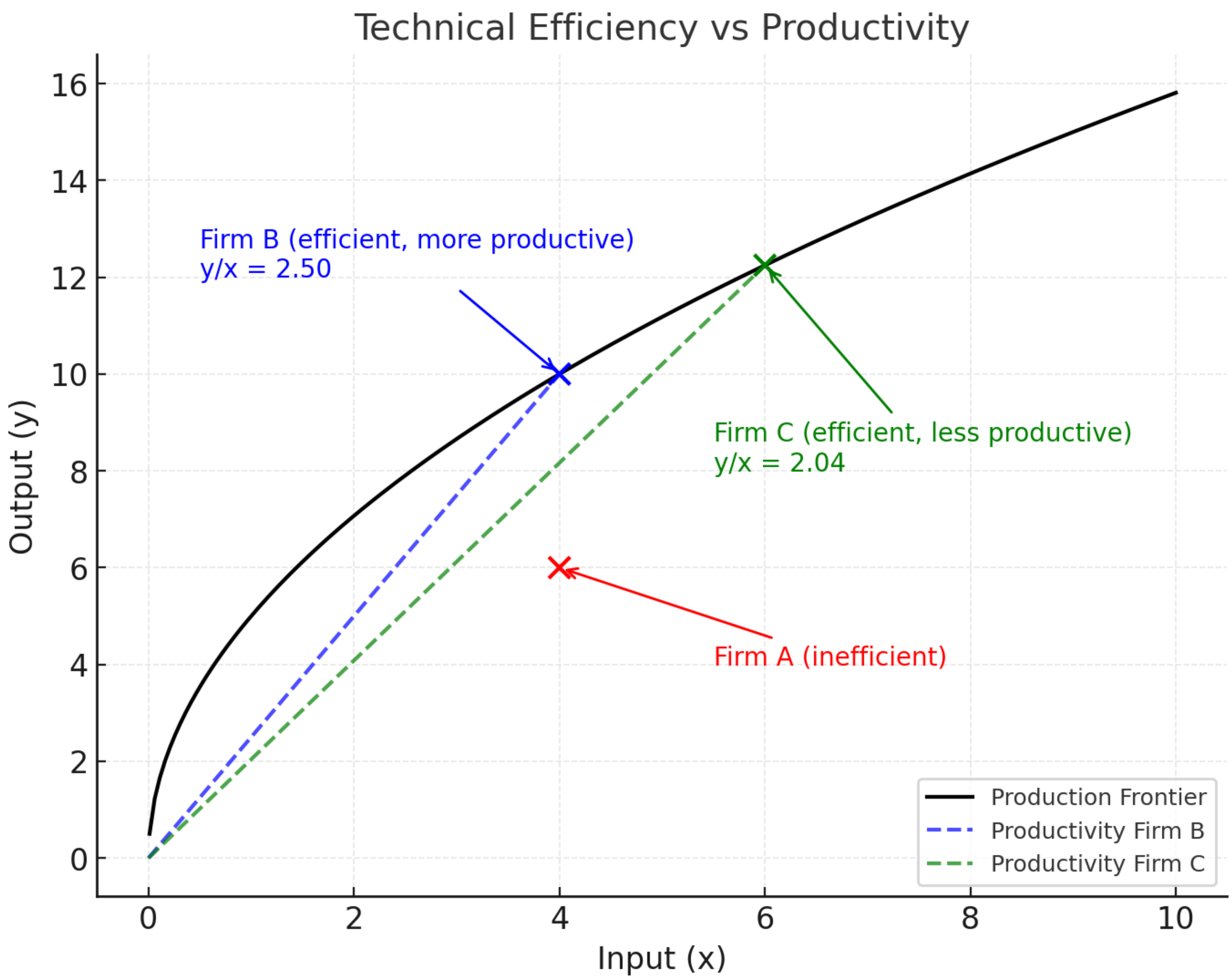
Objectives

- Estimating production technology from firm-level data
- Analyzing profit-maximizing, cost-minimizing behavior
- Applications:
  - profit firms, non-profits, industries (agriculture, manufacturing, services)

Method of Analysis

- Least-square method (LS)
- Total factor productictity (TFP)
- Data envelopment analysis (DEA)
- Stochastic frontier analysis (SFA)
    - We will focus on LS and TFP (which assume firms are technically efficient)
    - DEA and SFA can provide measures of relative efficiency

Technical efficiency vs productivity



Data

- I will primarily follow *appleProdFr86* data set, which can be found in Henningsen's *micEcon* R package
  - Additional datasets can be found in other R packages, for example, *sfaR*, *micEcon*, *rDEA*, *deaR*, *Benchmarking*, or others.
  - Another resourceful website: https://vincentarelbundock.github.io/Rdatasets/datasets.html

Dataset: appleProdFr86

- R code:

```
library(micEcon); library(psych); library(lmtest); library(car); library(miscTools)
options(scipen = 999)
data( "appleProdFr86", package = "micEcon" )
dat <- appleProdFr86
describe(dat)
```

	vars	n	mean	sd	median	trimmed	mad
vCap	1	140	102576.24	79992.28	84114.50	89202.88	55160.87
vLab	2	140	237199.39	194867.78	175871.00	199077.29	71095.12
vMat	3	140	201250.06	208054.52	136291.50	160457.37	92486.81
qApples	4	140	3.07	5.46	1.37	1.87	1.68
qOtherOut	5	140	1.50	1.32	1.07	1.29	0.95
qOut	6	140	2649825.38	3300778.29	1773989.17	2005998.00	1440430.19
pCap	7	140	1.30	0.79	1.11	1.20	0.56
pLab	8	140	1.01	0.20	0.96	1.00	0.20
pMat	9	140	6.77	2.64	6.25	6.55	2.75
pOut	10	140	1.01	0.53	0.83	0.91	0.31
adv	11	140	0.52	0.50	1.00	0.53	0.00

Dual approach to production analysis

- Duality is a fundamental concept in optimization, particularly in linear programming and game theory.
  —Every optimization problem (referred to as the primal problem) can be associated with a corresponding dual problem, where the solution to one provides bounds to the solution of the other.

- In the context of producer behavior, the primal approach involves studying how firms can optimally decide on the input mix for a given production technology to achieve an objective, such as minimizing expenses to produce a certain volume of output. The resulting optimal expense is referred to as the cost function.

- Duality tells us that the cost function is sufficiently informative, allowing us to confidently trace back the production technology under mild conditions.

Dual approach to production analysis

- Prior to 1970, economists mostly followed Samuelson's classic treatment of profit-maximizing firms, where firms face technological constraints, typically modeled with a smooth production function, and standard optimization techniques are used to infer producer responses to price perturbations.

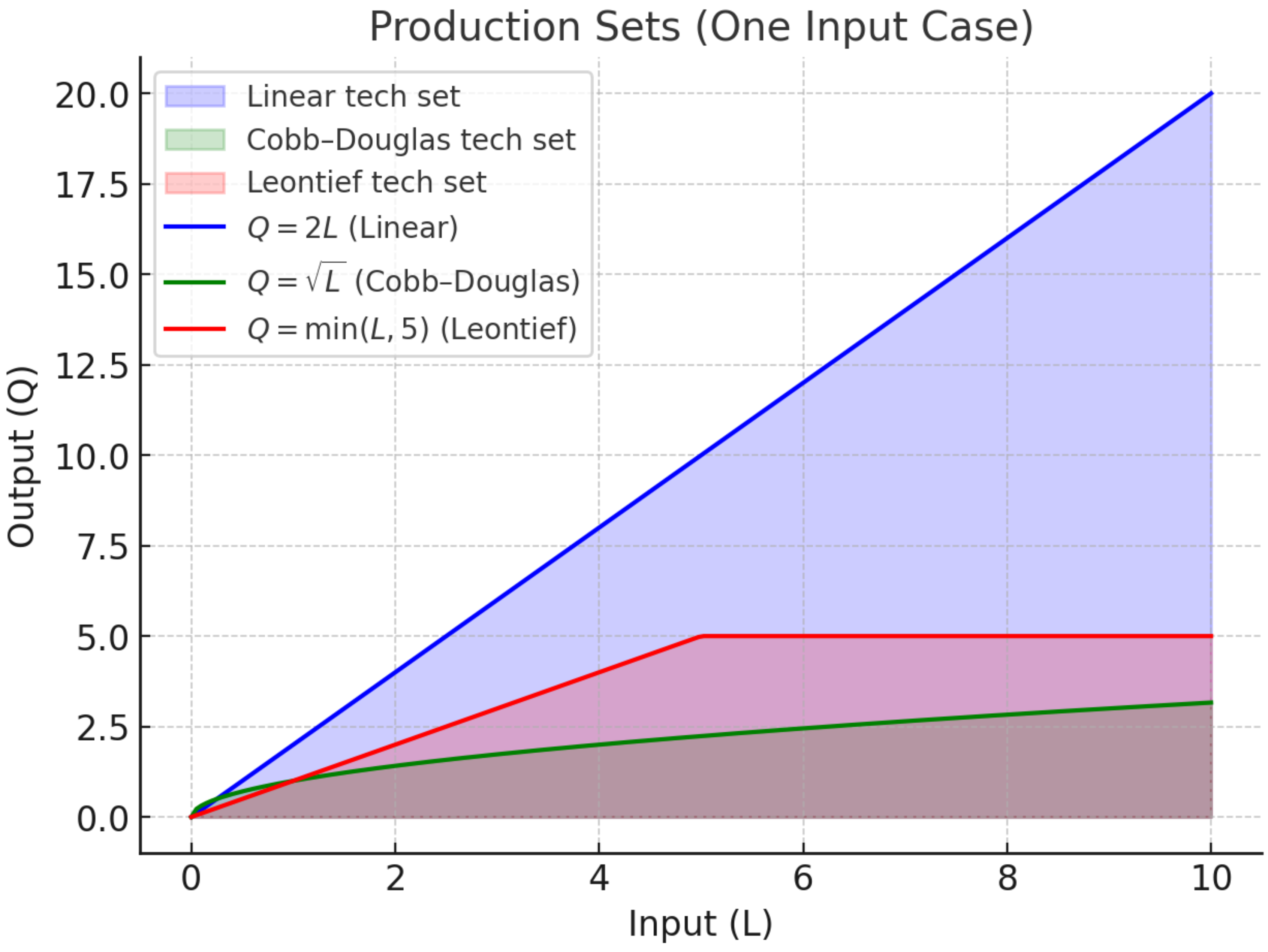
- This approach is often referred to as the primal approach.

- Later, the dual approach gained prominence, where exploring cost, profit, or revenue functions allows us to trace back the technological constraints.

Production technology

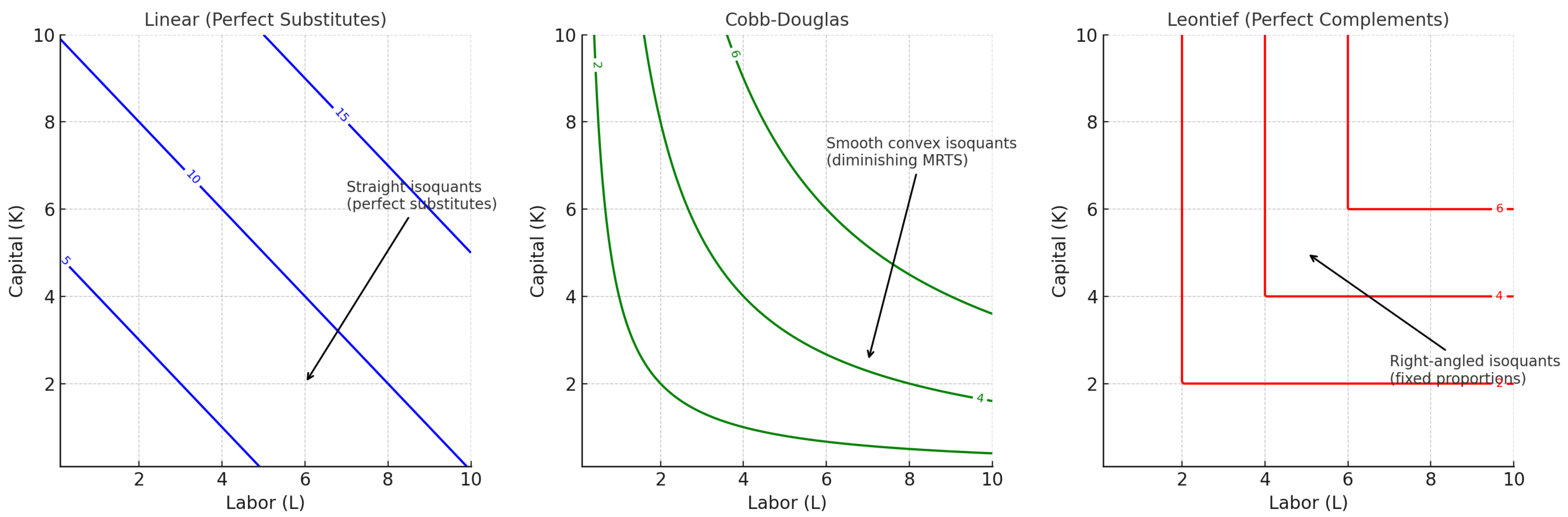
- The set of all combinations of inputs and outputs that comprise a technologically feasible way to produce is called a production (possibility) set.

- One-input case:



Production technology

- Two inputs—Presentation via isoquants
- The isoquants move in the top-right direction as y goes up, since we need more inputs to produce more output.
- The top-right section of the isoquants, and including the points on the isoquants, are often referred to as the input requirement set.



Production technology

- Convex technology—A technology is called convex if the input requirement set is convex.

  - For a convex technology, a convex combination of input choices increases the output volume.

- Monotone technology—A technology is called monotone if its input requirement set satisfies the monotonicity property, which suggests that for any input vector x belonging to the input requirement set, all input vectors weakly greater than x must belong to the input requirement set.

- A general representation of multi-output and multi-input production possibility is given by a transformation function T such that T (x, q) = 0 represents a relationship where an input vector x is used to produce an output vector q.

Production technology

- Some examples:

Linear: $y = \beta_0 + \sum_{i=1}^{N} \beta_i x_i$

Cobb-Douglas: $y = \beta_0 \prod_{i=1}^{N} {x_i}^{\beta_i}$, or equivalently, $\ln y = \beta_0 + \sum_{i=1}^{N} \beta_i \ln x_i$

Leontief: $y = \min_{i=1}^{N} \{\beta_i x_i\}$

CES: $y = \left[\sum_{i=1}^{N} \beta_i x_i^{\rho}\right]^{\frac{1}{\rho}}$

Quadratic: $y = \beta_0 + \sum_i \beta_i x_i + \frac{1}{2} \sum_i \sum_j \beta_{ij} x_i x_j$

Translog: $\ln y = \beta_0 + \sum_i \beta_i \ln x_i + \frac{1}{2} \sum_i \sum_j \beta_{ij} \ln x_i \ln x_j$

Production technology

- Returns to scale—How would the output change if we scale up or down the inputs?

- If the output goes up by the same factor k, we call it a constant returns to scale (CRS) technology. Mathematically, a CRS technology exhibits $f(kx) = kf(x)$.

- If the output increases less than k times, we call it a decreasing returns to scale (DRS) technology. Mathematically, a DRS technology exhibits $f(kx) < kf(x)$.

- If the output increases more than k times, we call it an increasing returns to scale (IRS) technology. Mathematically, an IRS technology exhibits $f(kx) > kf(x)$.

- *Exercise*: Consider a two-input Cobb-Douglas production function. Find conditions under which the technology exhibits different kinds of returns to scale.

Productivity

- Average and marginal product

- Single-input case: Consider a production relationship given by y = f(x) .

  The average productivity of the input x is defined by

$$AP = f(x)/x$$

  The marginal productivity of the input x is defined by

$$MP = \partial f(x)/\partial x$$

- Multi-input case:

$$AP_i = \frac{y}{x_i} = \frac{f(\mathbf{x})}{x_i}$$

$$MP_i = \frac{\partial y}{\partial x_i} = \frac{\partial f(\mathbf{x})}{\partial x_i} = f_i$$

Output elasticity of an input

- The output elasticity of an input measures the percentage changes in output because of a percentage change in input.

$$\varepsilon_i = \frac{\partial f(\mathbf{x})/f(\mathbf{x})}{\partial x_i/x_i} = \frac{MP_i}{AP_i}$$

- The output elasticities are free of the unit of measurement.

- The elasticity of scale is the sum of output elasticities of all input:

$$\varepsilon = \sum_i \varepsilon_i$$

- A technology exhibiting IRS, CRS, and DRS has the elasticity of scale

$$\varepsilon > 1,\ \varepsilon = 1, \text{ and } \varepsilon < 1,$$

respectively.

Total factor productivity (TFP)

- In multi-input production process, it is often desirable to calculate the total factor productivity (TFP) by aggregating inputs into an input index

$$TFP = \frac{y}{X},$$

where X is a quantity aggregating index of all inputs.

Indexing

- Indexing is used for measuring changes in a set of related variables.
- It can be used for comparison over time or space or both.
- Examples include price indices for measuring changes to consumer price, export or import prices, quantity indices measuring changes in output volume by a firm or industry over time or across firms.
- Consider a formula for measuring the change of the value of a basket consisting of n goods between the two period t and s can be measured by

$$X = \frac{\sum_{i=1}^{n} x_{it} p_{it}}{\sum_{i=1}^{n} x_{is} p_{is}}$$

If we fix the prices (either to current or old prices), we get a measure due to changes in quantity, and it then reflects a quantity index. Similarly, if we fix the quantity (either to current or old quantity levels), we will get a price index.

Various indices

- Laspeyres quantity index:

$$X_j^L = \frac{\sum_i x_{ij} p_{i0}}{\sum_i x_{i0} p_{i0}}$$

- Paasche quantity index:

$$X_j^P = \frac{\sum_i x_{ij} p_{ij}}{\sum_i x_{i0} p_{ij}}$$

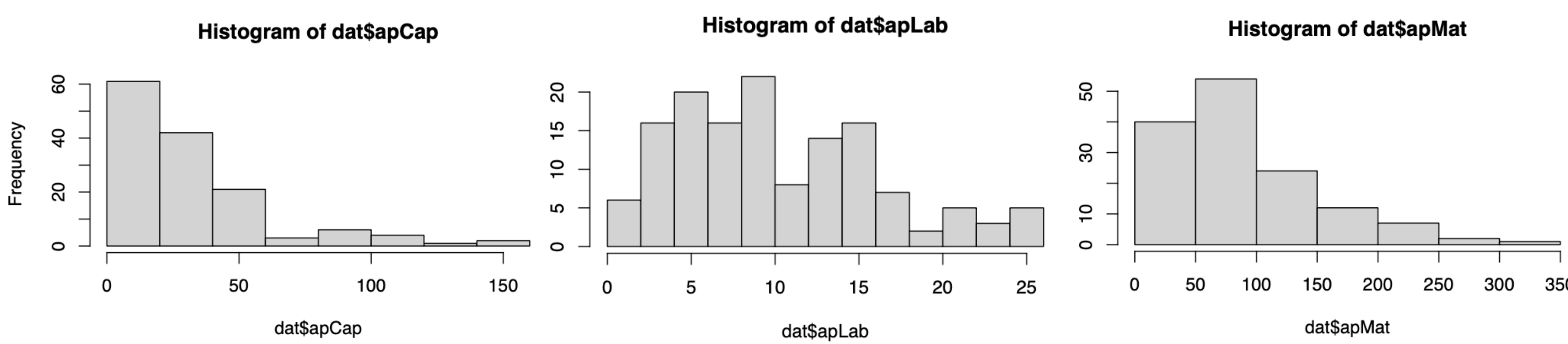
- Fisher's quantity index:

$$X_j^F = \sqrt{X_j^L \times X_j^P}$$

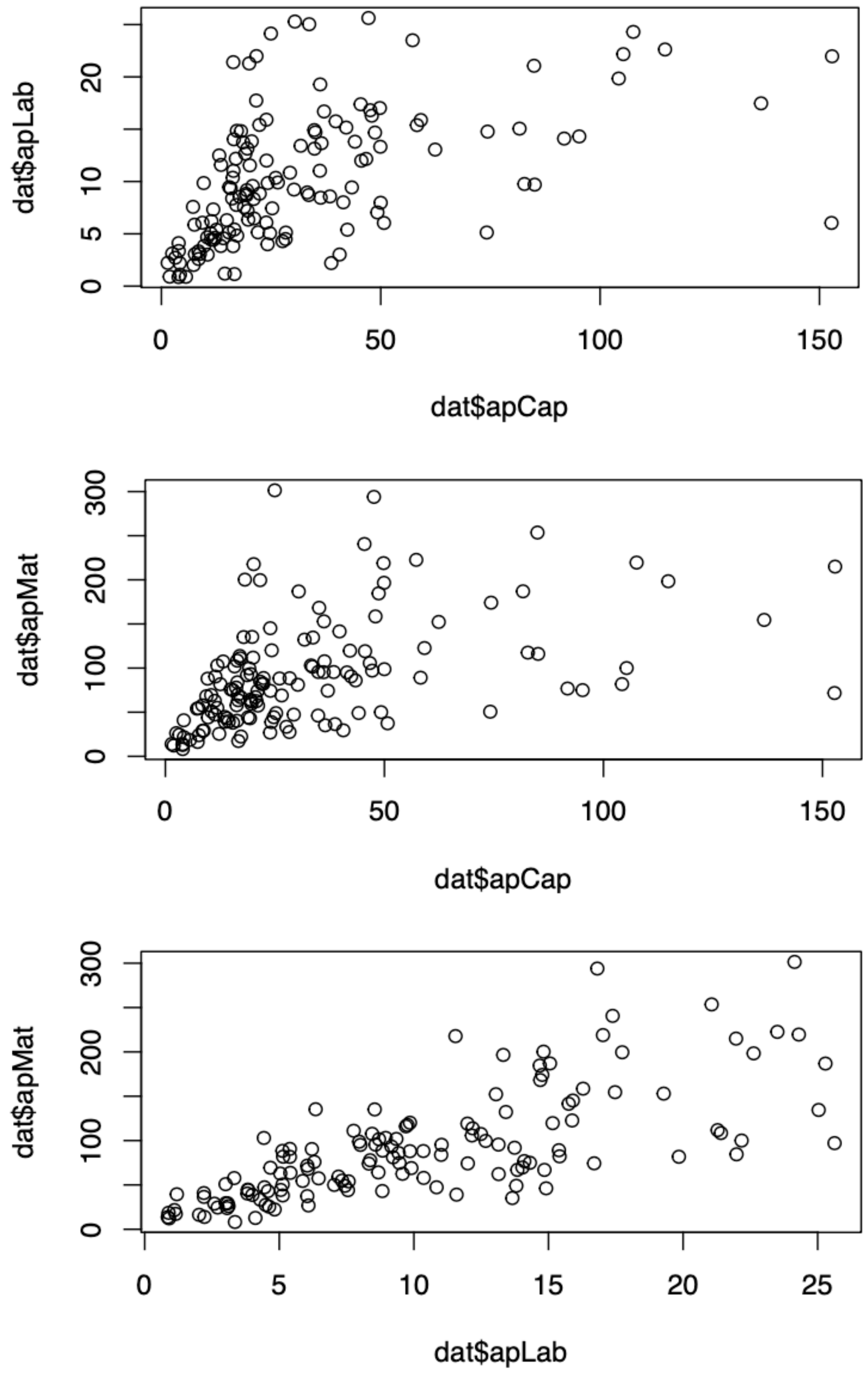
Productivity measures in our data set

- R code

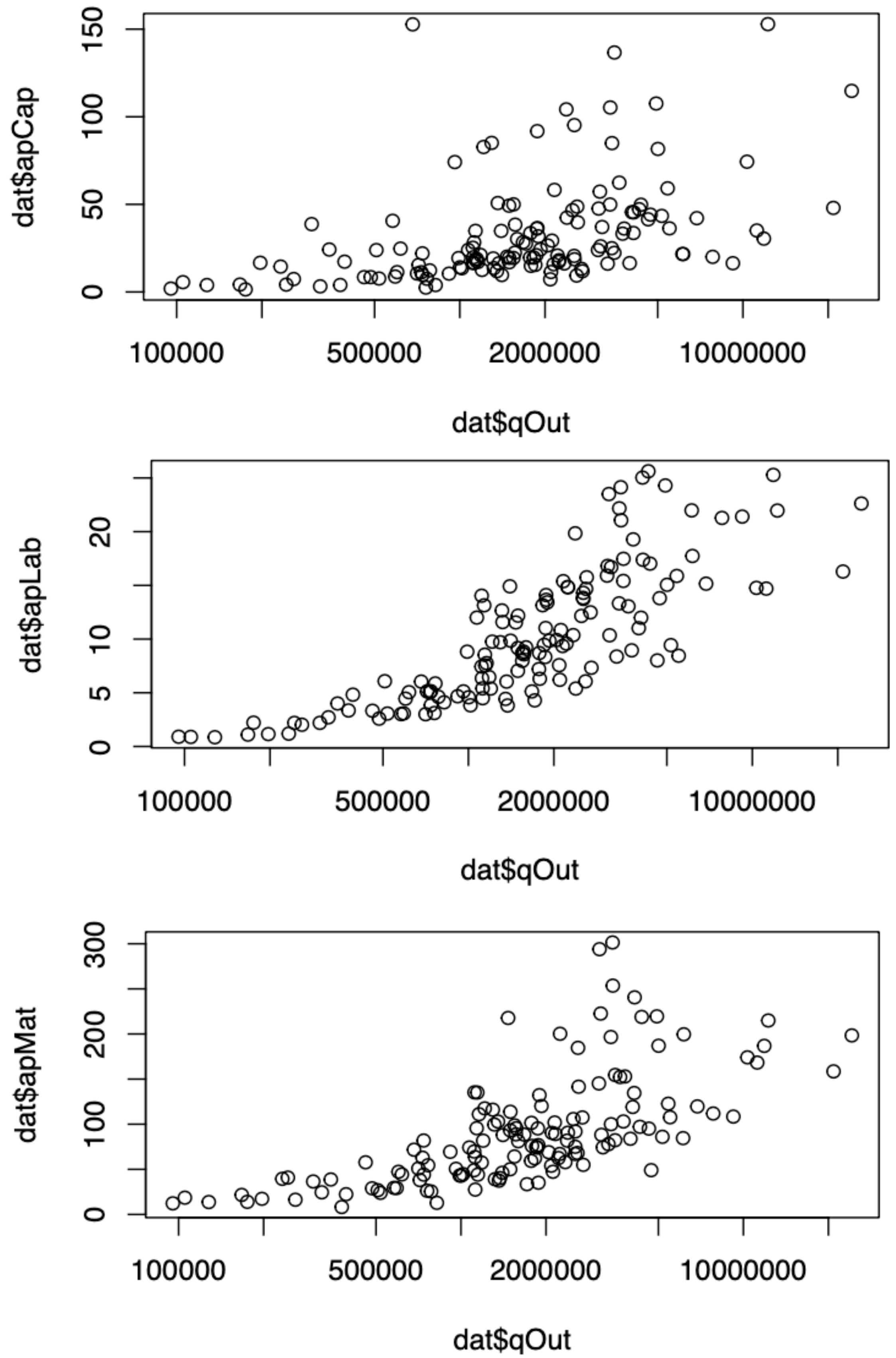
```
# Generate input quantities
dat$qCap <- dat$vCap / dat$pCap
dat$qLab <- dat$vLab / dat$pLab
dat$qMat <- dat$vMat / dat$pMat
# Creating quantity indices
dat$X <- sqrt( dat$XP * dat$XL ) # Fisher Index
# Measuring (partial) average product
dat$apCap <- dat$qOut / dat$qCap
dat$apLab <- dat$qOut / dat$qLab
dat$apMat <- dat$qOut / dat$qMat
```



Average products of the three inputs seem to be positively correlated



Firms producing more also exhibit higher output per unit of input used

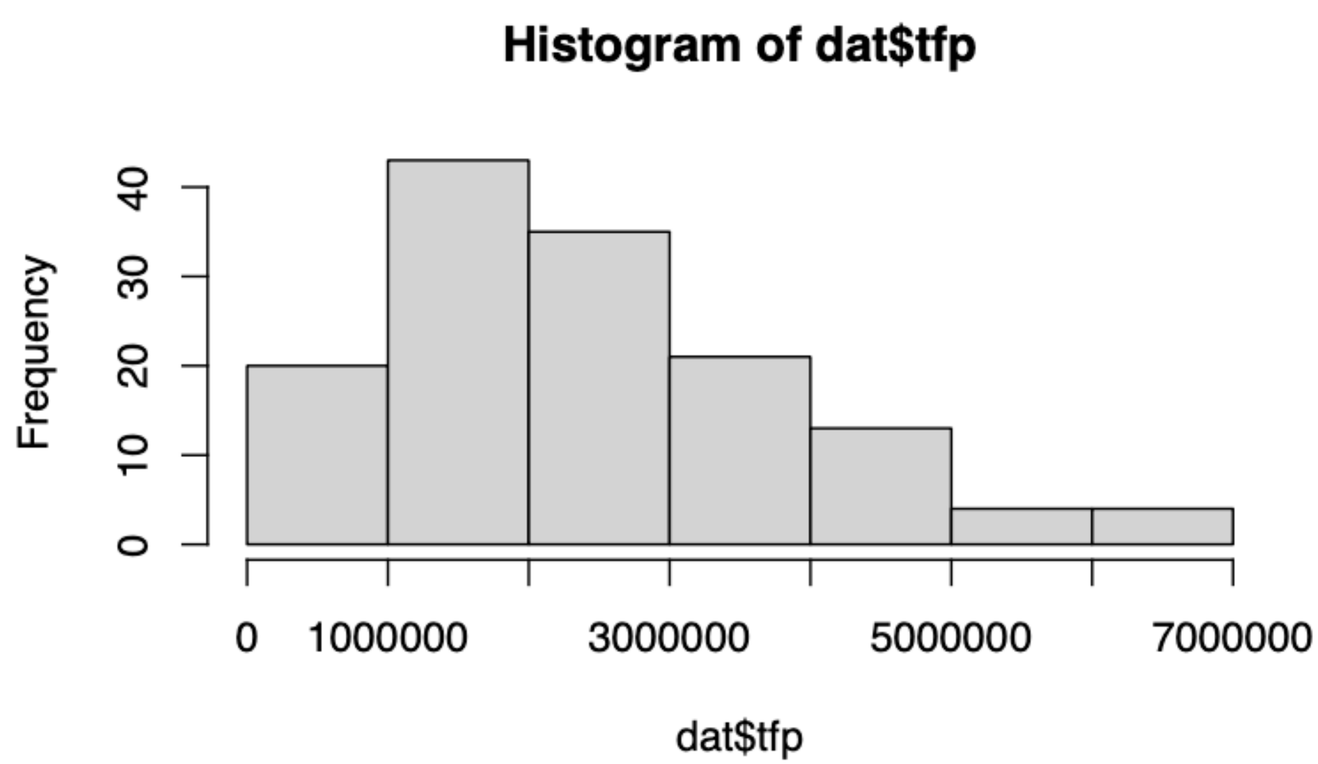


Total factor productivity

- R code

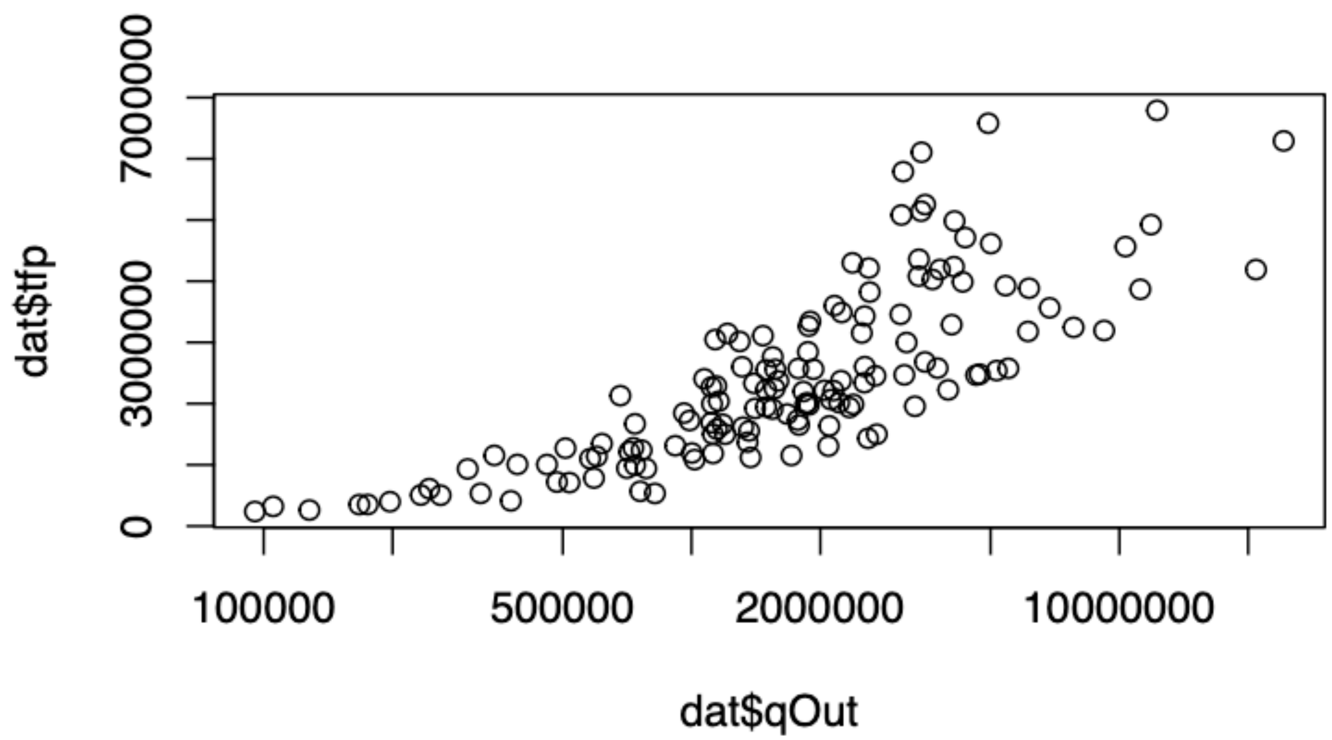
```
# Measuring total factor productivity
dat$tfp <- dat$qOut / dat$X # using Fisher index
dat$tfpP <- dat$qOut / dat$XP # using Paasche Index
dat$tfpL <- dat$qOut / dat$XL # using Laspeyres Index
```

- TFP varies considerably across firms, with the majority falling into the relatively low-TFP range.

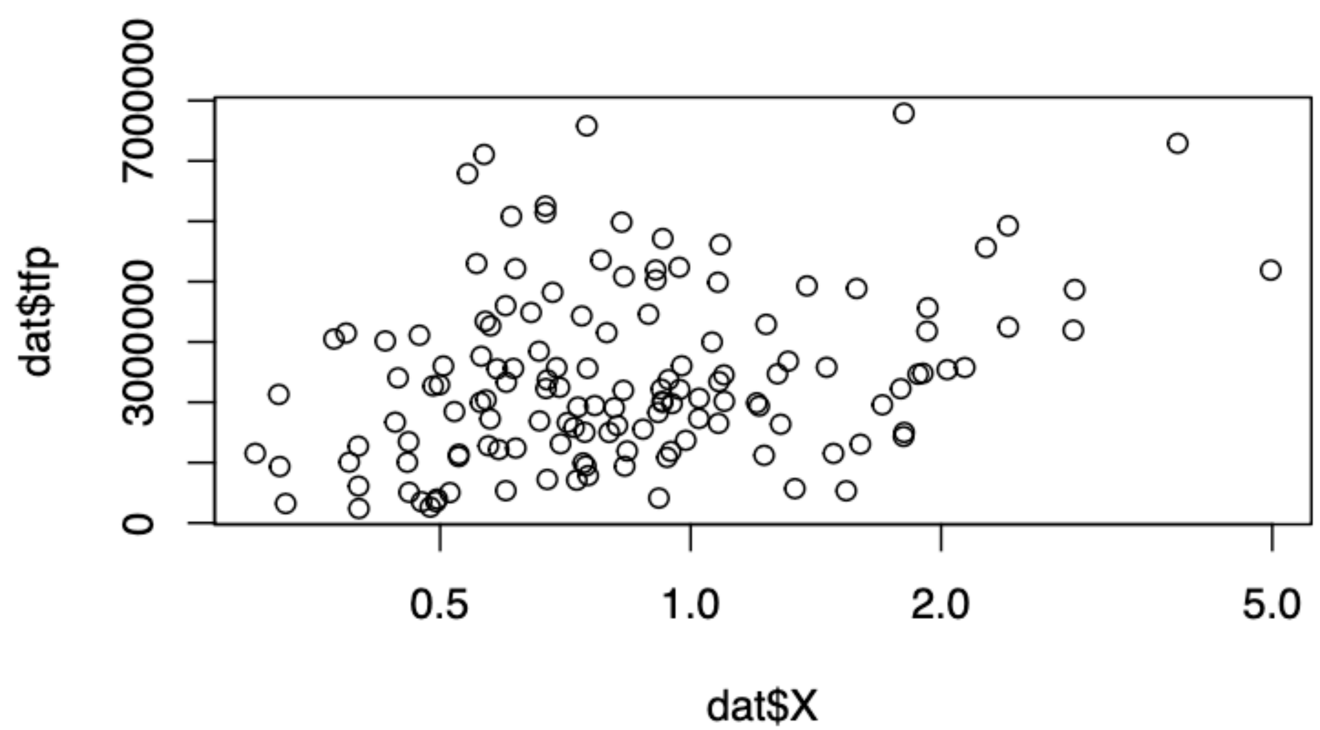


Total factor productivity

- Larger firms, characterized by higher output volumes, are typically associated with greater TFP.



- However, the plot of TFP against the aggregate input index shows only a mild positive association between the two.



- to be continued…