

# NUTRITION DEMAND, SUBSISTENCE FARMING, AND AGRICULTURAL PRODUCTIVITY

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agricultural productivity ↓

## CONTRIBUTIONS

- explore farm-level subsistence, document scale-dependent product choice

► *literature*

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- build model with nutrition demand driven by caloric needs as explanation
- show importance of farm-level subsistence for aggr. agricultural productivity

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DATA

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  - ▶ *food* ▶ *output*
- Rescale HH kcal intake, output, income by HH kcal requirement
  - “per capita” measures, weighted by energy needs

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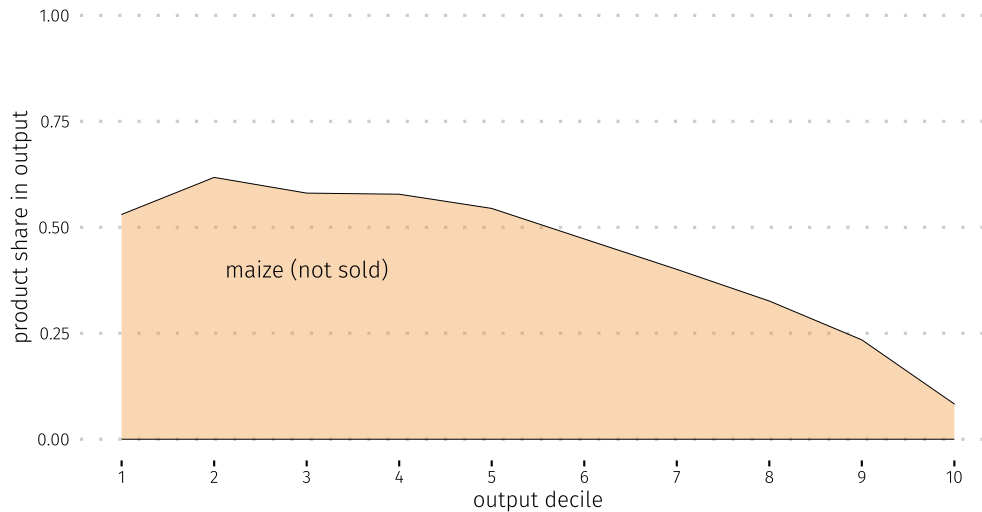
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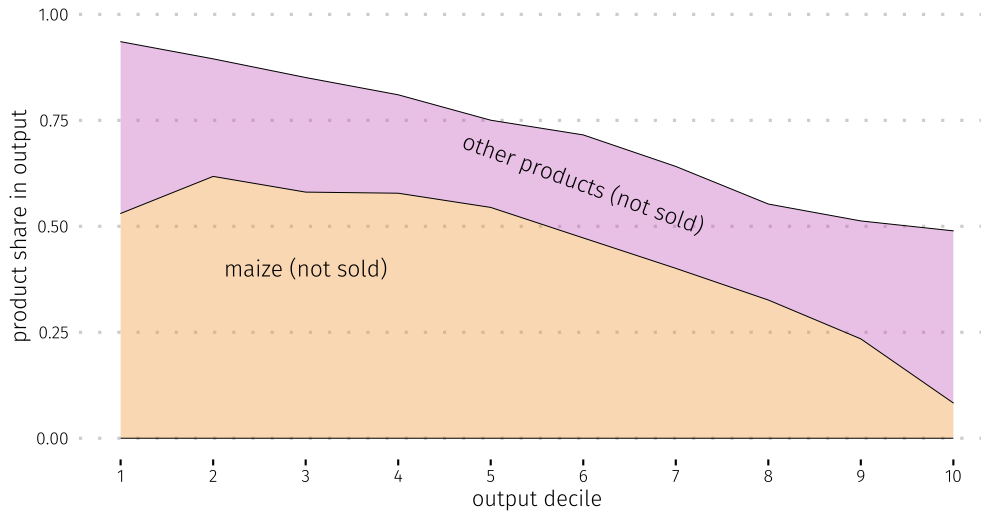
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  - avg share of output sold: 16%

## FARM SIZE $\uparrow \rightarrow$ SHIFT FARM FROM MAIZE TO DIVERSITY, COMMERCIALIZE

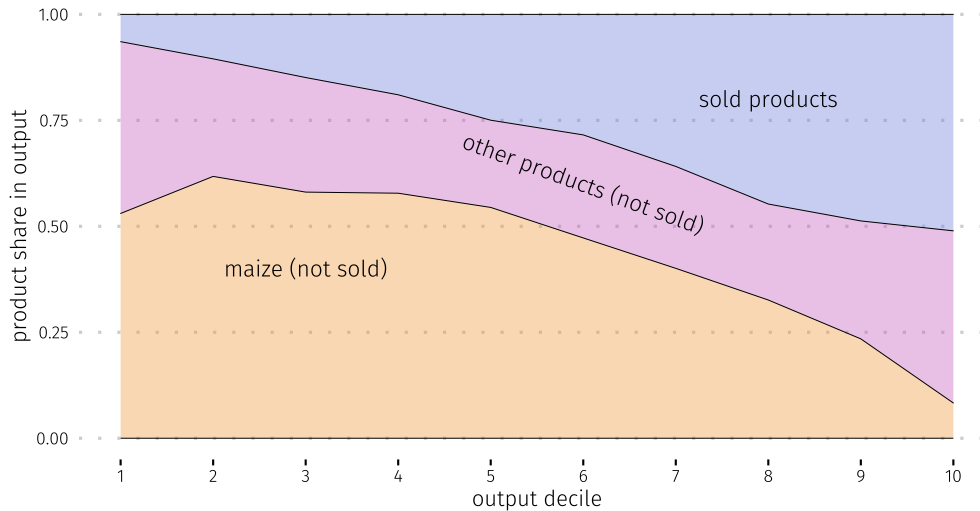




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MODEL

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## MODEL: HH PROBLEM

- heterogeneous households and agricultural products, solve GE ► *details*
- HH  $h$  consumes  $n$  foods  $\{c_{h,i}\}_{i=1}^n$  and a manufactured good  $c_{h,m}$

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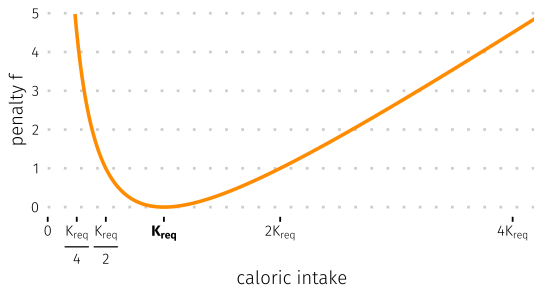
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## CALORIC DEVIATION PENALTY $f$

- caloric deviation penalty fn (► *properties*):

$$f\left(\sum_i c_{h,i} k_i, K_{req,h}\right) = \psi\left(\frac{\sum_i c_{h,i} k_i - K_{req,h}}{K_{req,h}}\right)^2 \frac{K_{req,h}}{\sum_i c_{h,i} k_i}$$



## FARM BEHAVIOR IN MODEL AND DATA

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## FARM SIZE $\uparrow$ $\rightarrow$ SHIFT FROM DIETARY ENERGY TO DIVERSITY: MODEL & DATA

|  | log kcal intake | food diversity $\blacktriangleright$ <i>def</i> $\blacktriangleright$ <i>nutrients</i> |
|--|-----------------|--|
| log output                                   |                 |  |
| log non-farm income                          |                 |  |
| N  |                 |  |
| Adj. $R^2$                                   |                 |  |
| * $p < 0.1$ , ** $p < 0.05$ , *** $p < 0.01$ |                 |  |



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|------------------------|--------------------|--|
|                        | model:<br>CES-only |  |
| log output             | 0.732<br>(0.001)   |  |
| log non-farm<br>income | 0.289<br>(0.001)   |  |
| N                      | 35,520             |  |
| Adj. $R^2$             | 0.937              |  |

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

- CES-only ( $\psi = 0$ ): relative consumptions invariant to size/income
  - kcal intake  $\uparrow$  proportionally to total shadow income, diversity constant

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|------------------------|--------------------|--|
|                        | model:<br>CES-only | model:<br>CES-only   |
| log output             | 0.732<br>(0.001)   | -0.061<br>(0.001)  |
| log non-farm<br>income | 0.289<br>(0.001)   | 0.031<br>(0.002)   |
| N                      | 35,520             | 35,520   |
| Adj. $R^2$             | 0.937              | 0.054  |

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|                        | model:<br>CES-only | model:<br>benchmark | model:<br>CES-only   |
| log output             | 0.732<br>(0.001)   | 0.124<br>(0.001)    | -0.061<br>(0.001)  |
| log non-farm<br>income | 0.289<br>(0.001)   | 0.084<br>(0.001)    | 0.031<br>(0.002)   |
| N                      | 35,520             | 33,613              | 35,520   |
| Adj. $R^2$             | 0.937              | 0.393               | 0.054  |

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| log output             | 0.732<br>(0.001)   | 0.124<br>(0.001)    | -0.061<br>(0.001)  | 0.428<br>(0.002)    |
| log non-farm<br>income | 0.289<br>(0.001)   | 0.084<br>(0.001)    | 0.031<br>(0.002)   | 0.396<br>(0.002)    |
| N                      | 35,520             | 33,613              | 35,520   | 33,613              |
| Adj. R <sup>2</sup>    | 0.937              | 0.393               | 0.054  | 0.762               |

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- $\triangleright$  details  $\triangleright$  comparison to Stone-Geary

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| log output             | 0.732<br>(0.001)   | 0.124<br>(0.001)    | 0.091***<br>(0.005) | -0.061<br>(0.001)  | 0.428<br>(0.002)    |
| log non-farm<br>income | 0.289<br>(0.001)   | 0.084<br>(0.001)    | 0.063***<br>(0.004) | 0.031<br>(0.002)   | 0.396<br>(0.002)    |
| N                      | 35,520             | 33,613              | 8,674               | 35,520   | 33,613              |
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| log output             | 0.732<br>(0.001)   | 0.124<br>(0.001)    | 0.091***<br>(0.005) | -0.061<br>(0.001)  | 0.428<br>(0.002)    | 0.395***<br>(0.034) |
| log non-farm<br>income | 0.289<br>(0.001)   | 0.084<br>(0.001)    | 0.063***<br>(0.004) | 0.031<br>(0.002)   | 0.396<br>(0.002)    | 0.857***<br>(0.033) |
| N                      | 35,520             | 33,613              | 8,674               | 35,520   | 33,613              | 8,675               |
| Adj. R <sup>2</sup>    | 0.937              | 0.393               | 0.063               | 0.054  | 0.762               | 0.131               |

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## SELLING BEHAVIOR: MODEL & DATA

### LARGE FARMS ARE MORE ACTIVE SELLERS ► *details*

- **MODEL** & **DATA**: farm size  $\uparrow \rightarrow$  sell bigger fraction of output
- **model mechanism**: size  $\uparrow \rightarrow$  reallocate cons. to diversity, manuf.  $\rightarrow$  need revenue

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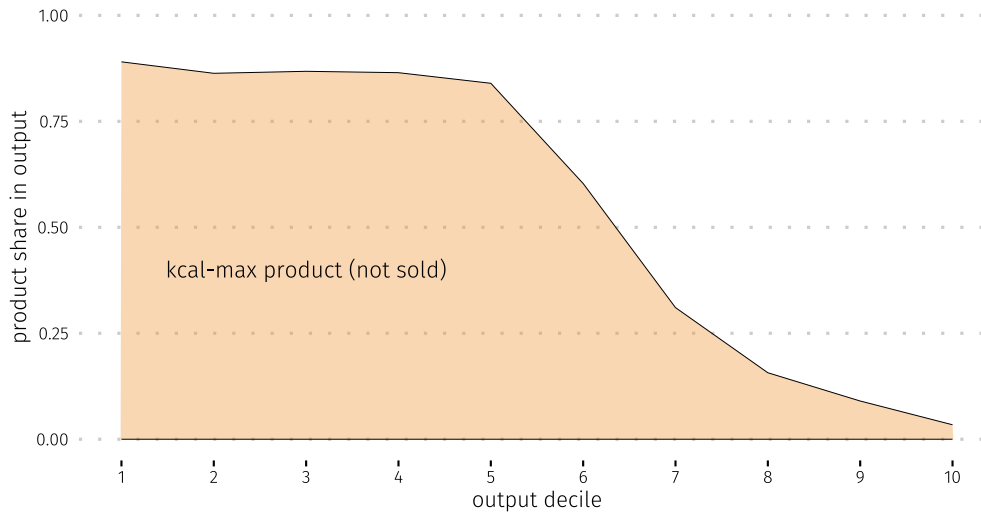
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### SALES ARE SPECIALIZED ► *details*

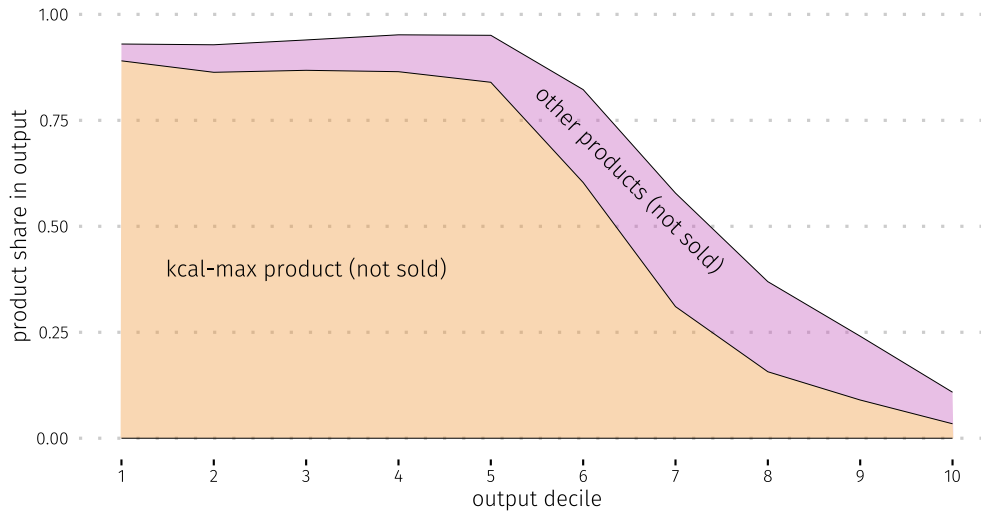
- **MODEL** & **DATA**: sales are specialized compared to overall production
- **model mechanism**: sell only the most revenue-productive good, but can produce others for own consumption



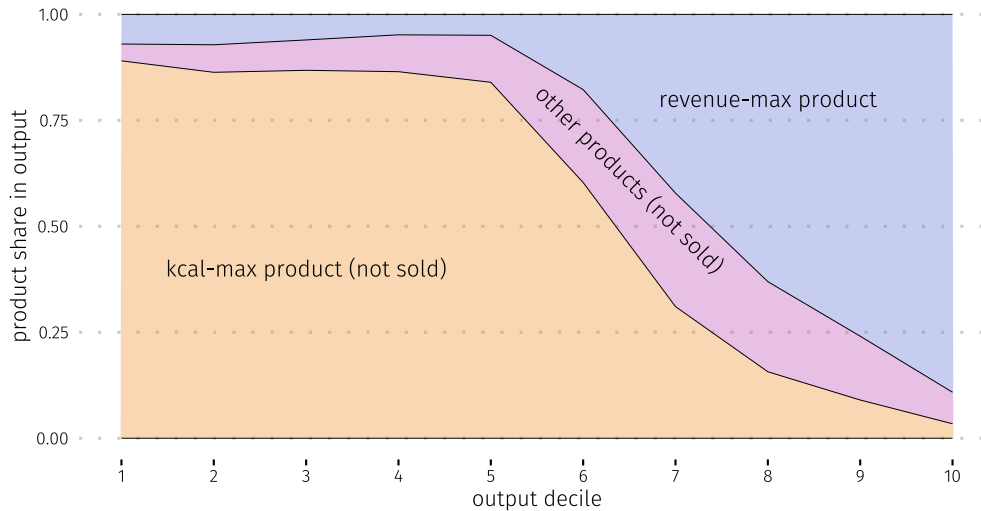
## SMALL SPECIALIZE, MEDIUM DIVERSIFY, LARGE COMMERCIALIZE: MODEL



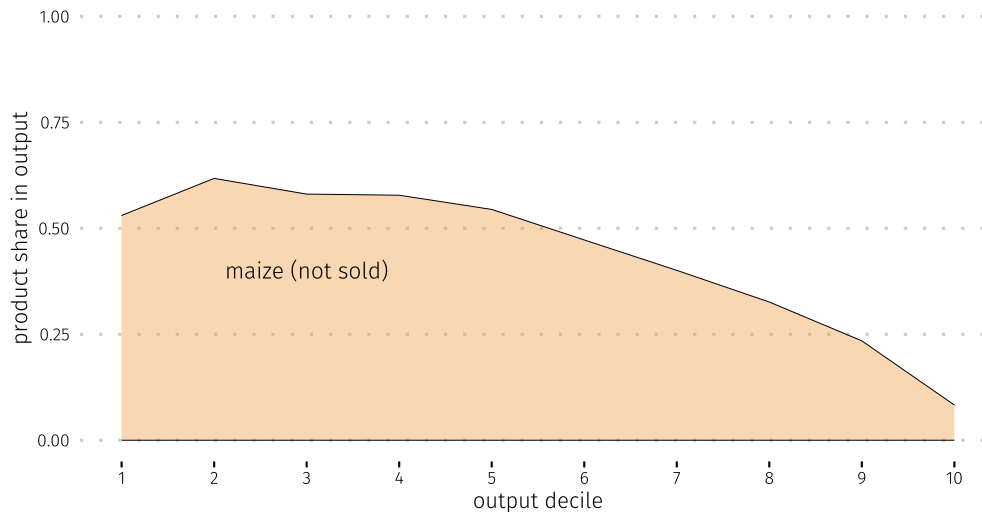
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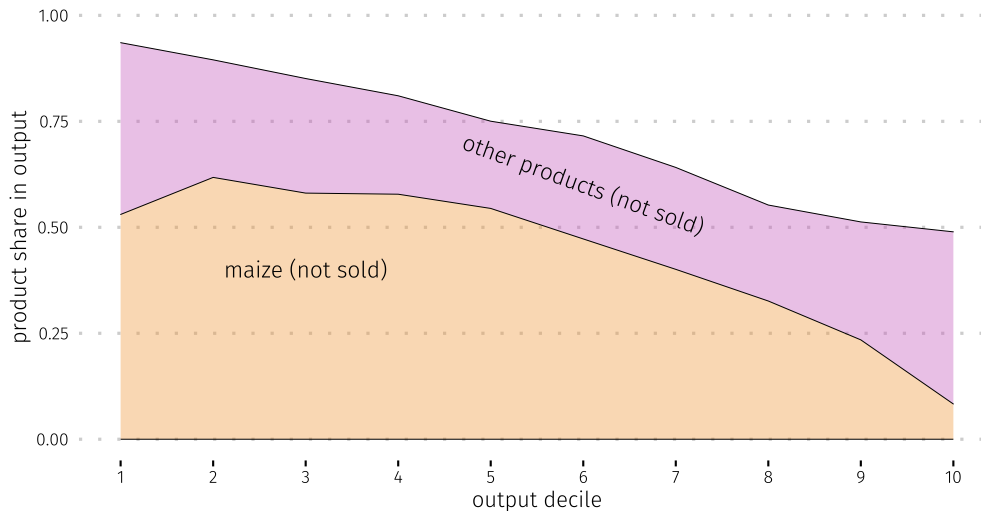
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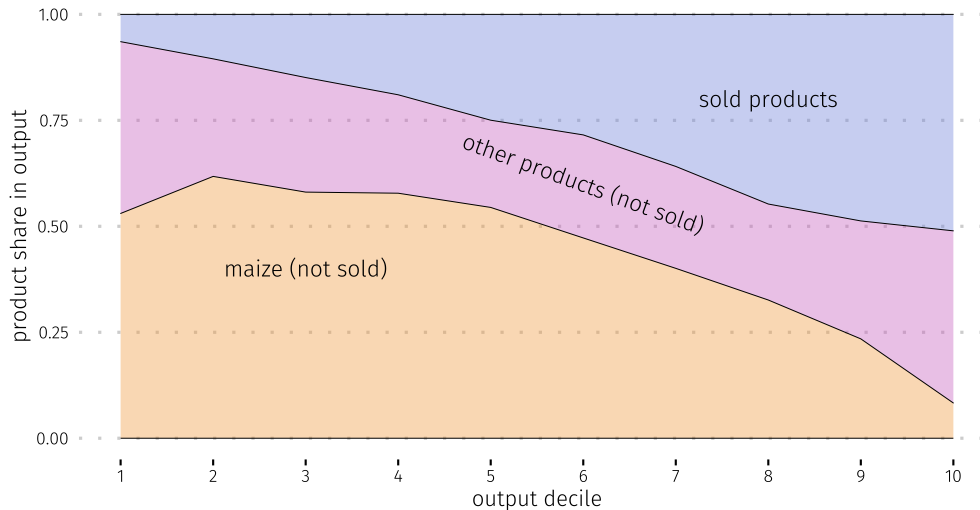
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# AGGREGATE PRODUCTIVITY

---

## TRADE COSTS ↓ → AGGREGATE PRODUCTIVITY ↑

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  - scale-dependence of consumption, production, selling



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  - $\rightarrow$  subsistence matters more for macro, nutrition matters for micro

## CONCLUSION

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# CONCLUSION

Subsistence farmer nutrition demand



Farm production decisions



Aggregate agricultural productivity ↓

- smallest farms specialize in calories
- medium farms diversify diet & production
- largest farms become market-oriented
- if partially leave subsistence → agric. productivity ↑
- half because improved product choice
- calories matter less than subsistence itself

## FUTURE RESEARCH

- Analyze government programs targeting smallholder farmers

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- Analyze **government programs** targeting smallholder farmers
  - smallholder farmer support is central to public policy in poor countries
  - existing & proposed policies: encourage staples, biodiversity, or cash crops?
  - framework well suited for predicting nutritional, economic outcomes

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back

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*nutrition literature:*

## SUBSISTENCE FARMING & NUTRITION

- *Jones (2017), Sibhatu et al. (2015)*
  - smallholder farm biodiversity related to dietary diversity
  - especially with poor market access
- farm characteristics matter for nutritional outcomes

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back

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  - HH-product consumption in past week

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## OUTPUT & INCOME

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[back](#)

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- Non-farm income
  - income from employment and non-farm enterprises

[back](#)

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$$\sum_{i=1}^n x_{h,i}^p p_i d + p_m c_{h,m} \leq \sum_{i=1}^n x_{h,i}^s \frac{p_i}{d} + w N_h$$

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## $f(\sum_i c_i k_i, K_{req})$ PROPERTIES

Properties:

1.  $f(bK_{in}, bK_{req}) = f(K_{in}, K_{req})$

(homogeneity of deg. 0)

2.  $f(bK_{req}, K_{req}) = f\left(\frac{K_{req}}{b}, K_{req}\right)$

(symmetry around  $K_{req}$  in ratios)

3.  $\min_{K_{in} > 0} f(K_{in}, K_{req}) = f(K_{req}, K_{req}) = 0$

(minimum and zero if eat  $K_{req}$ )

4.  $f_{11}(K_{in}, K_{req}) = \frac{2\psi K_{req}}{K_{in}^3} > 0$

(convex in intake)

back

## CALORIES SKEW CONSUMPTION: MODEL

- Consider the problem of a household
- Suppose  $\psi = 0$  (CES-only)

$$MU_i^{\text{CES}}(c_i) = MC_i \quad (c_i \text{ FOC})$$

## CALORIES SKEW CONSUMPTION: MODEL

- Consider the problem of a household
- Suppose  $\psi > 0$  (benchmark)

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## CALIBRATION

### Agricultural Goods

- 6 agricultural goods commonly produced and consumed
  - ▶ *list*

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▶ *Calibration: parameters & moments*

## AGRICULTURAL GOODS USED IN CALIBRATION

- Selected goods:
  1. maize
  2. pigeonpea
  3. groundnut
  4. tomato
  5. soybean
  6. tobacco
- These goods account for, on average,
  - 70% of HH output market value
  - 43% of HH food consumption market value

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## GENERAL EQUILIBRIUM: AGRICULTURAL GOODS

- Solve for agricultural prices  $\{p_i\}_i$  s.t. edible good markets clear:

$$\frac{1}{d} \sum_h x_{h,i}^s = d \sum_h x_{h,i}^p \quad \forall i$$



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  - data: tobacco accounts for 60% of Malawi's exports
  - **tobacco** traded internationally at exogenous  $\bar{p}_t$
  - some **manufactured good** is imported to balance the trade:

$$\underbrace{\bar{p}_{\text{tobacco}} \left( \frac{1}{d} \sum_h x_{h,\text{tobacco}}^s - d \sum_h x_{h,\text{tobacco}}^p \right)}_{\text{tobacco exports}} = \underbrace{p_m \left( \sum_h c_{h,m} - Y_m \right)}_{\text{manuf. good imports}}$$

## FARM SALES ARE SPECIALIZED: MODEL & DATA

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back

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  - on avg, top good accounts for 91% in sales but 67% in output

back

## LOWER TRADE COSTS → ALL SPECIALIZE: MODEL & DATA

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  - below some cutoff  $\tilde{d}_h$ , HH  $h$  only produces the revenue-maximizing good

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- **MODEL:**  $d \downarrow \rightarrow$  specialize production
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- **DATA:**
  - HHs with better market access specialize production
    - ▶ *table*

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## LARGE FARMS SELL MORE: MODEL & DATA

- Larger farms are more active sellers:

| output<br>quartile | <u>sold output share</u> |
|--------------------|--------------------------|
|--------------------|--------------------------|

|   |  |
|---|--|
| 1 |  |
|---|--|

|   |  |
|---|--|
| 4 |  |
|---|--|

back

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$\nearrow$  food diversity  $\uparrow$

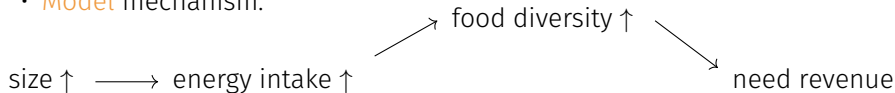
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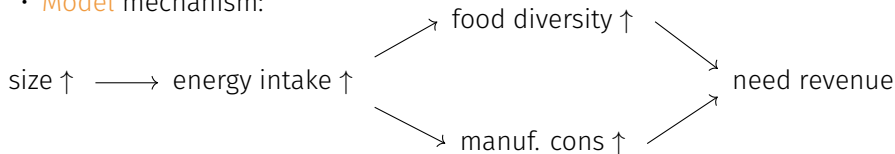
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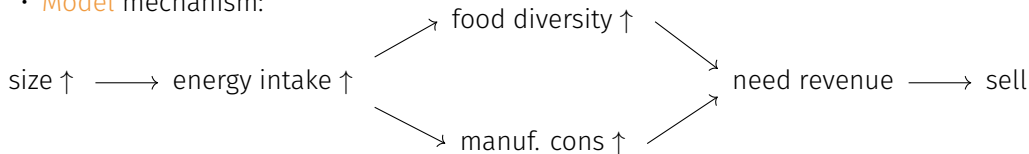
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## CUTOFF TRADE COST $\bar{d}$

$$\bar{d}_h = \sqrt{\frac{\max_i p_i z_{h,i}}{\min_i p_i / k_i \cdot \max_i k_i z_{h,i}}}$$

back



## FOOD DIVERSITY

- Food Diversity = Inverse Simpson Index

$$\text{Food Diversity}_h = \left( \sum_{i=1}^n \left( \frac{\text{food quantity}_{h,i} \times \text{median purchase price}_i}{\sum_{j=1}^n \text{food quantity}_{h,j} \times \text{median purchase price}_j} \right)^2 \right)^{-1}$$

where  $h$  is the HH index,  $n$  is the total number of distinct foods in the dataset.

- Simpson Index: sum of squared food shares within HH's consumption
  - same as HHI
  - interpretation: probability that two random dollars of (shadow) food expenditure come from the same product
- Inverse Simpson Index =  $\frac{1}{S_I}$ , commonly used in measuring species diversity

## NUTRIENT RICHNESS

|                     | NRF9                 |                       | NRF9.3                |                       |
|---------------------|----------------------|-----------------------|-----------------------|-----------------------|
|                     | (1)                  | (2)                   | (3)                   | (4)                   |
| log output          | 17.046***<br>(0.964) | 5.695***<br>(0.724)   | -13.296***<br>(3.326) | -13.400***<br>(3.358) |
| log non-farm income | 10.285***<br>(0.792) | 2.441***<br>(0.603)   | -7.257**<br>(3.898)   | -7.305**<br>(3.548)   |
| log kcal intake     |                      | 124.025***<br>(2.282) |                       | 0.550<br>(26.234)     |
| N                   | 8,675                | 8,674                 | 8,675                 | 8,674                 |
| Adj. R <sup>2</sup> | 0.054                | 0.451                 | 0.002                 | 0.002                 |

\* p < 0.1, \*\* p < 0.05, \*\*\* p < 0.01

- NRF9: sum of daily intakes (relative to recommended level) of 9 nutrients
- NRF9.3: subtracts the relative excessive consumption of 3 disqualifying nutrients

## LOWER TRADE COSTS → ALL SPECIALIZE: DATA

---

production diversity

---

---

N

Adj. R<sup>2</sup>

---

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NOTE. Controls: log output, log non-farm income.

back

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sold output share

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## LOWER TRADE COSTS → ALL SPECIALIZE: DATA

|                     | production diversity |
|---------------------|----------------------|
| sold output share   | -0.044***<br>(0.016) |
| N                   | 4,042                |
| Adj. R <sup>2</sup> | 0.025                |

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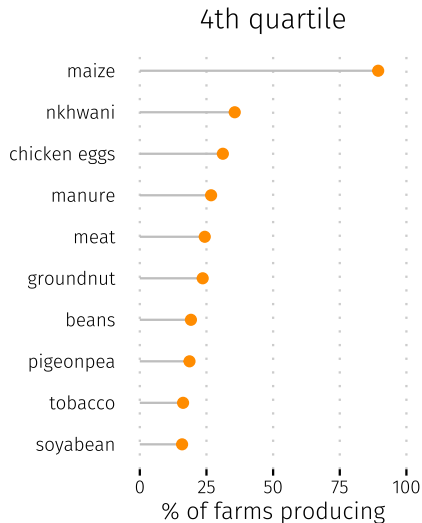
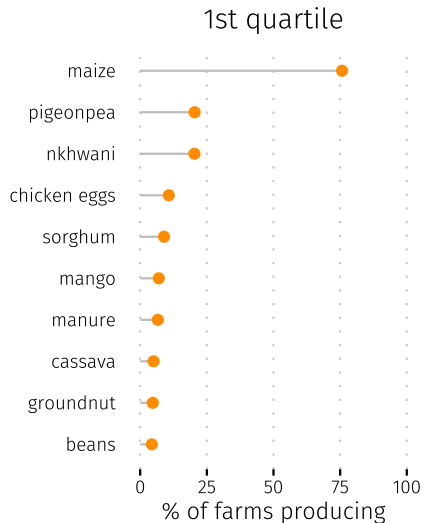
| production diversity |                      |                      |
|----------------------|----------------------|----------------------|
| sold output share    | -0.044***<br>(0.016) |                      |
| 1 [good mkt access]  |                      | -0.164***<br>(0.018) |
| N                    | 4,042                | 8,675                |
| Adj. R <sup>2</sup>  | 0.025                | 0.099                |

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back

## PRODUCT FREQUENCY BY SIZE: DATA





## CALIBRATION

| parameter | value | moment/source | data<br>moment | model<br>moment |
|-----------|-------|---------------|----------------|-----------------|
|-----------|-------|---------------|----------------|-----------------|

Distributions

back

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| parameter              | value | moment/source            | data<br>moment | model<br>moment |
|------------------------|-------|--------------------------|----------------|-----------------|
| <b>Distributions</b>   |       |                          |                |                 |
| $\mathbb{E}(\log L_h)$ | -15   | avg $K_{in,h}/K_{req,h}$ | 1.036          | 0.902           |

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| <b>Distributions</b>   |       |                           |                |                 |
| $\mathbb{E}(\log L_h)$ | -15   | avg $K_{in,h}/K_{req,h}$  | 1.036          | 0.902           |
| $V(\log L_h)$          | 1.5   | $V(\log \text{output}_h)$ | 1.528          | 1.385           |

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| $P(N_h = 0)$           | 0.112 | $P(\text{non-farm income}_h = 0)$ | 0.112          | 0.117           |

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| $V(\log N_h \mid N_h > 0)$ | 2.103 | $V(\log \text{non-farm income}_h)$ | 2.103          | 1.924           |

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| parameter                   | value | moment/source                      | data<br>moment | model<br>moment |
|-----------------------------|-------|------------------------------------|----------------|-----------------|
| <b>Distributions</b>        |       |                                    |                |                 |
| $\mathbb{E}(\log L_h)$      | -15   | avg $K_{in,h}/K_{req,h}$           | 1.036          | 0.902           |
| $V(\log L_h)$               | 1.5   | $V(\log \text{output}_h)$          | 1.528          | 1.385           |
| $P(N_h = 0)$                | 0.112 | $P(\text{non-farm income}_h = 0)$  | 0.112          | 0.117           |
| $V(\log N_h \mid N_h > 0)$  | 2.103 | $V(\log \text{non-farm income}_h)$ | 2.103          | 1.924           |
| <b>Parameters</b>           |       |                                    |                |                 |
| $\sigma$ (EoS across foods) | 0.75  | estimated                          | —              | —               |

## CALIBRATION

| parameter                            | value | moment/source                      | data<br>moment | model<br>moment |
|--------------------------------------|-------|------------------------------------|----------------|-----------------|
| <b>Distributions</b>                 |       |                                    |                |                 |
| $\mathbb{E}(\log L_h)$               | -15   | avg $K_{in,h}/K_{req,h}$           | 1.036          | 0.902           |
| $V(\log L_h)$                        | 1.5   | $V(\log \text{output}_h)$          | 1.528          | 1.385           |
| $P(N_h = 0)$                         | 0.112 | $P(\text{non-farm income}_h = 0)$  | 0.112          | 0.117           |
| $V(\log N_h \mid N_h > 0)$           | 2.103 | $V(\log \text{non-farm income}_h)$ | 2.103          | 1.924           |
| <b>Parameters</b>                    |       |                                    |                |                 |
| $\sigma$ (EoS across foods)          | 0.75  | estimated                          | —              | —               |
| $\gamma$ (EoS between food & manuf.) | 1     | —                                  | —              | —               |



## CALIBRATION

| parameter                            | value | moment/source                      | data<br>moment | model<br>moment |
|--------------------------------------|-------|------------------------------------|----------------|-----------------|
| <b>Distributions</b>                 |       |                                    |                |                 |
| $\mathbb{E}(\log L_h)$               | -15   | avg $K_{in,h}/K_{req,h}$           | 1.036          | 0.902           |
| $V(\log L_h)$                        | 1.5   | $V(\log \text{output}_h)$          | 1.528          | 1.385           |
| $P(N_h = 0)$                         | 0.112 | $P(\text{non-farm income}_h = 0)$  | 0.112          | 0.117           |
| $V(\log N_h \mid N_h > 0)$           | 2.103 | $V(\log \text{non-farm income}_h)$ | 2.103          | 1.924           |
| <b>Parameters</b>                    |       |                                    |                |                 |
| $\sigma$ (EoS across foods)          | 0.75  | estimated                          | —              | —               |
| $\gamma$ (EoS between food & manuf.) | 1     | —                                  | —              | —               |
| $d$ (agricultural trade cost)        | 1.75  | avg share sold                     | 0.159          | 0.203           |

## CALIBRATION

| parameter                            | value | moment/source                      | data<br>moment | model<br>moment |
|--------------------------------------|-------|------------------------------------|----------------|-----------------|
| <b>Distributions</b>                 |       |                                    |                |                 |
| $\mathbb{E}(\log L_h)$               | -15   | avg $K_{in,h}/K_{req,h}$           | 1.036          | 0.902           |
| $V(\log L_h)$                        | 1.5   | $V(\log \text{output}_h)$          | 1.528          | 1.385           |
| $P(N_h = 0)$                         | 0.112 | $P(\text{non-farm income}_h = 0)$  | 0.112          | 0.117           |
| $V(\log N_h \mid N_h > 0)$           | 2.103 | $V(\log \text{non-farm income}_h)$ | 2.103          | 1.924           |
| <b>Parameters</b>                    |       |                                    |                |                 |
| $\sigma$ (EoS across foods)          | 0.75  | estimated                          | —              | —               |
| $\gamma$ (EoS between food & manuf.) | 1     | —                                  | —              | —               |
| $d$ (agricultural trade cost)        | 1.75  | avg share sold                     | 0.159          | 0.203           |
| $\psi$ (kcal deviation penalty)      | 0.5   | output elasticity of $K_{in}$      | 0.091          | 0.124           |

## CALIBRATION

| parameter                            | value | moment/source                      | data<br>moment | model<br>moment |
|--------------------------------------|-------|------------------------------------|----------------|-----------------|
| <b>Distributions</b>                 |       |                                    |                |                 |
| $\mathbb{E}(\log L_h)$               | -15   | avg $K_{in,h}/K_{req,h}$           | 1.036          | 0.902           |
| $V(\log L_h)$                        | 1.5   | $V(\log \text{output}_h)$          | 1.528          | 1.385           |
| $P(N_h = 0)$                         | 0.112 | $P(\text{non-farm income}_h = 0)$  | 0.112          | 0.117           |
| $V(\log N_h \mid N_h > 0)$           | 2.103 | $V(\log \text{non-farm income}_h)$ | 2.103          | 1.924           |
| <b>Parameters</b>                    |       |                                    |                |                 |
| $\sigma$ (EoS across foods)          | 0.75  | estimated                          | —              | —               |
| $\gamma$ (EoS between food & manuf.) | 1     | —                                  | —              | —               |
| $d$ (agricultural trade cost)        | 1.75  | avg share sold                     | 0.159          | 0.203           |
| $\psi$ (kcal deviation penalty)      | 0.5   | output elasticity of $K_{in}$      | 0.091          | 0.124           |
| <b>Good characteristics</b>          |       |                                    |                |                 |

## CALIBRATION

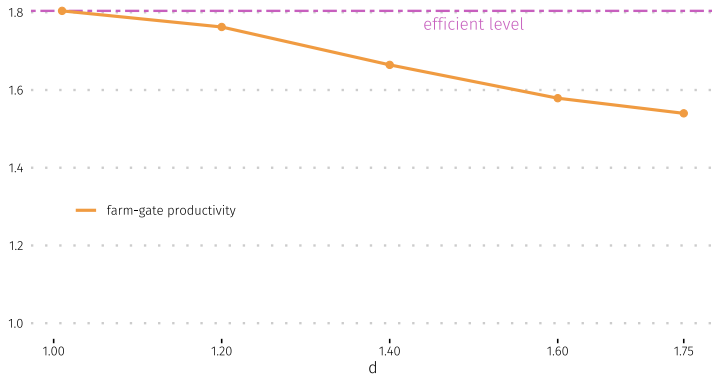
| parameter                            | value | moment/source   | data<br>moment | model<br>moment |
|--------------------------------------|-------|---|----------------|-----------------|
| <b>Distributions</b>                 |       |   |                |                 |
| $\mathbb{E}(\log L_h)$               | -15   | avg $K_{in,h}/K_{req,h}$  | 1.036          | 0.902           |
| $V(\log L_h)$                        | 1.5   | $V(\log \text{output}_h)$                                       | 1.528          | 1.385           |
| $P(N_h = 0)$                         | 0.112 | $P(\text{non-farm income}_h = 0)$                               | 0.112          | 0.117           |
| $V(\log N_h \mid N_h > 0)$           | 2.103 | $V(\log \text{non-farm income}_h)$                              | 2.103          | 1.924           |
| <b>Parameters</b>                    |       |   |                |                 |
| $\sigma$ (EoS across foods)          | 0.75  | estimated   | —              | —               |
| $\gamma$ (EoS between food & manuf.) | 1     | —   | —              | —               |
| $d$ (agricultural trade cost)        | 1.75  | avg share sold  | 0.159          | 0.203           |
| $\psi$ (kcal deviation penalty)      | 0.5   | output elasticity of $K_{in}$                                   | 0.091          | 0.124           |
| <b>Good characteristics</b>          |       |   |                |                 |
| $\varphi_m$ (manuf. taste weight)    | 0.5   | $\frac{\text{aggr. non-farm income}}{\text{aggr. farm output}}$ | 1.539          | 1.632           |

## CALIBRATION

| parameter                                   | value | moment/source   | data<br>moment | model<br>moment |
|---|-------|---|----------------|-----------------|
| <b>Distributions</b>                        |       |   |                |                 |
| $\mathbb{E}(\log L_h)$                      | -15   | avg $K_{in,h}/K_{req,h}$  | 1.036          | 0.902           |
| $V(\log L_h)$                               | 1.5   | $V(\log \text{output}_h)$                                       | 1.528          | 1.385           |
| $P(N_h = 0)$                                | 0.112 | $P(\text{non-farm income}_h = 0)$                               | 0.112          | 0.117           |
| $V(\log N_h \mid N_h > 0)$                  | 2.103 | $V(\log \text{non-farm income}_h)$                              | 2.103          | 1.924           |
| <b>Parameters</b>                           |       |   |                |                 |
| $\sigma$ (EoS across foods)                 | 0.75  | estimated   | —              | —               |
| $\gamma$ (EoS between food & manuf.)        | 1     | —   | —              | —               |
| $d$ (agricultural trade cost)               | 1.75  | avg share sold  | 0.159          | 0.203           |
| $\psi$ (kcal deviation penalty)             | 0.5   | output elasticity of $K_{in}$                                   | 0.091          | 0.124           |
| <b>Good characteristics</b>                 |       |   |                |                 |
| $\varphi_m$ (manuf. taste weight)           | 0.5   | $\frac{\text{aggr. non-farm income}}{\text{aggr. farm output}}$ | 1.539          | 1.632           |
| $\bar{p}_{\text{tobacco}}/p_{\text{maize}}$ | 5.4   | aggr. tobacco output share                                      | 0.091          | 0.094           |

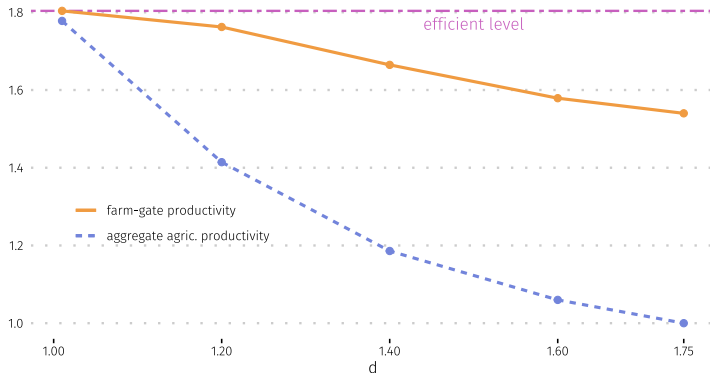
## TRADE COSTS ↓ → AGGREGATE PRODUCTIVITY ↑

- Compare “farm-gate” production to final consumption (“aggregate productivity”)
  - farm production only accounts for product choice changes



## TRADE COSTS $\downarrow \rightarrow$ AGGREGATE PRODUCTIVITY $\uparrow$

- Compare “farm-gate” production to final consumption (“aggregate productivity”)
  - farm production only accounts for product choice changes
  - final consumption also accounts for mechanical losses from  $d$

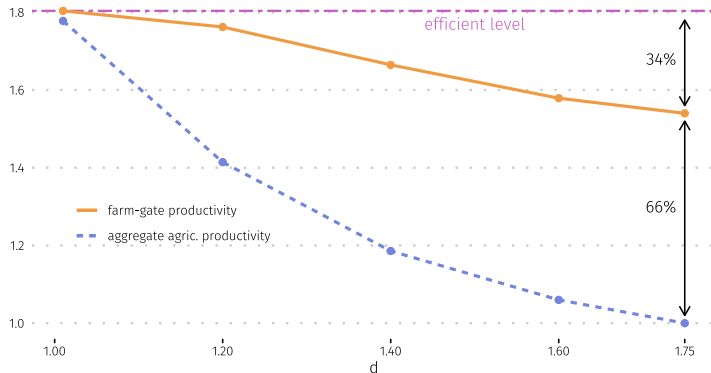


$d \rightarrow 1$ :

aggr. productivity  $\uparrow$  78%

## TRADE COSTS $\downarrow \rightarrow$ AGGREGATE PRODUCTIVITY $\uparrow$

- Compare “farm-gate” production to final consumption (“aggregate productivity”)
  - farm production only accounts for product choice changes
  - final consumption also accounts for mechanical losses from  $d$



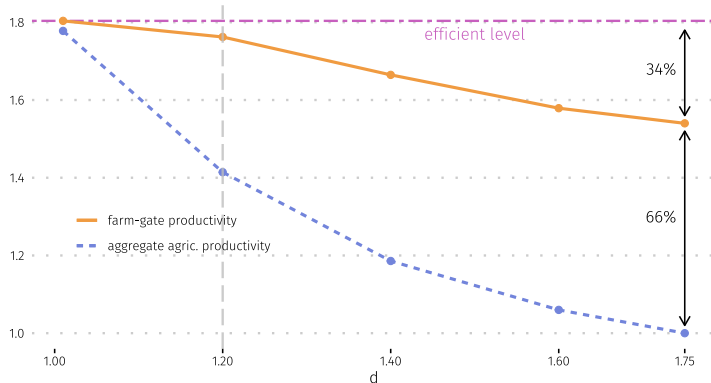
$d \rightarrow 1$ :

aggr. productivity  $\uparrow$  78% ( $\frac{1}{3}$  due to product choice)



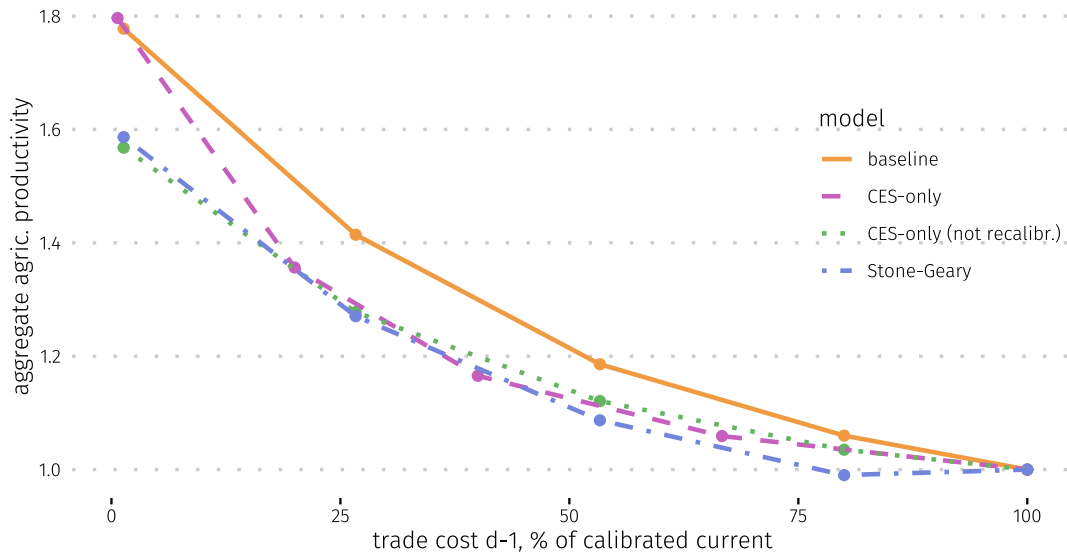
## TRADE COSTS $\downarrow \rightarrow$ AGGREGATE PRODUCTIVITY $\uparrow$

- Compare “farm-gate” production to final consumption (“aggregate productivity”)
  - farm production only accounts for product choice changes
  - final consumption also accounts for mechanical losses from  $d$



- $d \rightarrow 1$ : aggr. productivity  $\uparrow 78\%$  ( $\frac{1}{3}$  due to product choice)
- $d \downarrow$  s.t. avg share sold 16%  $\rightarrow$  50%: aggr. productivity  $\uparrow 42\%$  ( $\frac{1}{2}$  due to product choice)

## AGGREGATE AGRICULTURAL PRODUCTIVITY ACROSS MODELS

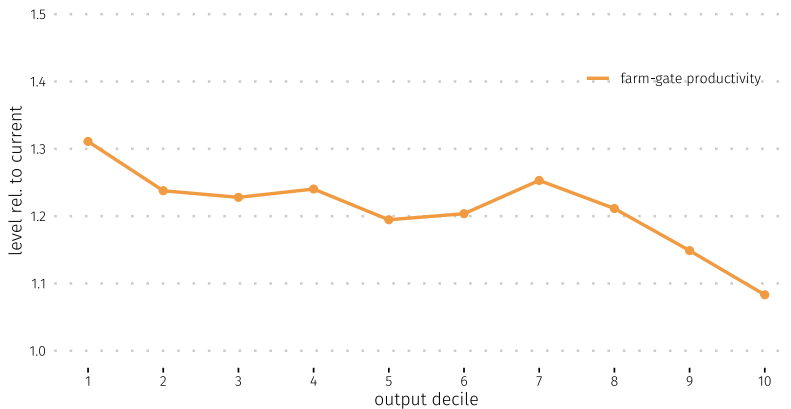


## TRADE COSTS $\downarrow \rightarrow$ HETEROGENEOUS EFFECTS IN FARM SIZE

- $d \downarrow$  s.t. avg share sold 16%  $\rightarrow$  50%:

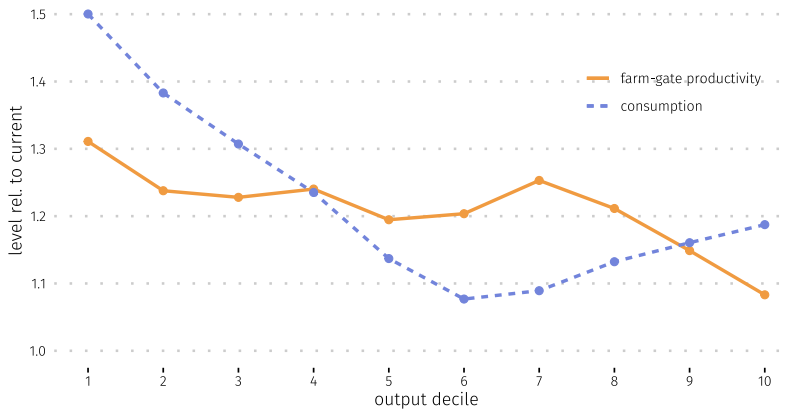
## TRADE COSTS ↓ → HETEROGENEOUS EFFECTS IN FARM SIZE

- $d \downarrow$  s.t. avg share sold 16% → 50%:
  - farm productivity: small ↑ the most, large ↑ the least



## TRADE COSTS ↓ → HETEROGENEOUS EFFECTS IN FARM SIZE

- $d \downarrow$  s.t. avg share sold 16% → 50%:
  - farm productivity: small ↑ the most, large ↑ the least
  - consumption: small ↑ the most, medium ↑ the least



## FARM SIZE AND FOOD CONSUMPTION: STONE-GEARY

Household food consumption vs farm size: Stone-Geary vs baseline model and data

|                        | log kcal intake              |                           |                     | food diversity               |                           |                     |
|------------------------|------------------------------|---------------------------|---------------------|------------------------------|---------------------------|---------------------|
|                        | (1)<br>model:<br>Stone-Geary | (2)<br>model:<br>baseline | (3)<br>data         | (4)<br>model:<br>Stone-Geary | (5)<br>model:<br>baseline | (6)<br>data         |
| log output             | 0.260<br>(0.001)             | 0.124<br>(0.001)          | 0.091***<br>(0.005) | -0.118<br>(0.001)            | 0.428<br>(0.002)          | 0.395***<br>(0.034) |
| log non-farm<br>income | 0.223<br>(0.001)             | 0.084<br>(0.001)          | 0.063***<br>(0.004) | 0.029<br>(0.001)             | 0.396<br>(0.002)          | 0.857***<br>(0.033) |
| N                      | 35,483                       | 33,613                    | 8,674               | 35,483                       | 33,613                    | 8,675               |
| Adj. R <sup>2</sup>    | 0.819                        | 0.393                     | 0.063               | 0.196                        | 0.762                     | 0.131               |

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$