

Identifying Investment Opportunities In International Telecommunications Markets Using Regression Models

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Introduction

► Background

Teledensity

Number of main telephone lines
(copper lines) per 100 inhabitants

1995 Teledensity Data

United States 62.57

Mexico 9.58

Source: *ITU World Telecommunications
Development Report 1998*

Introduction

► Background

Identify countries with teledensity levels that are *lower* than we might expect

Underperforming countries are best candidates for investment

Understand why a country has a low teledensity level

National governments do not want to get left behind

Introduction

► Background

1998 study by Herschel Shosteck Associates

HSA built a simple linear regression model to estimate teledensity

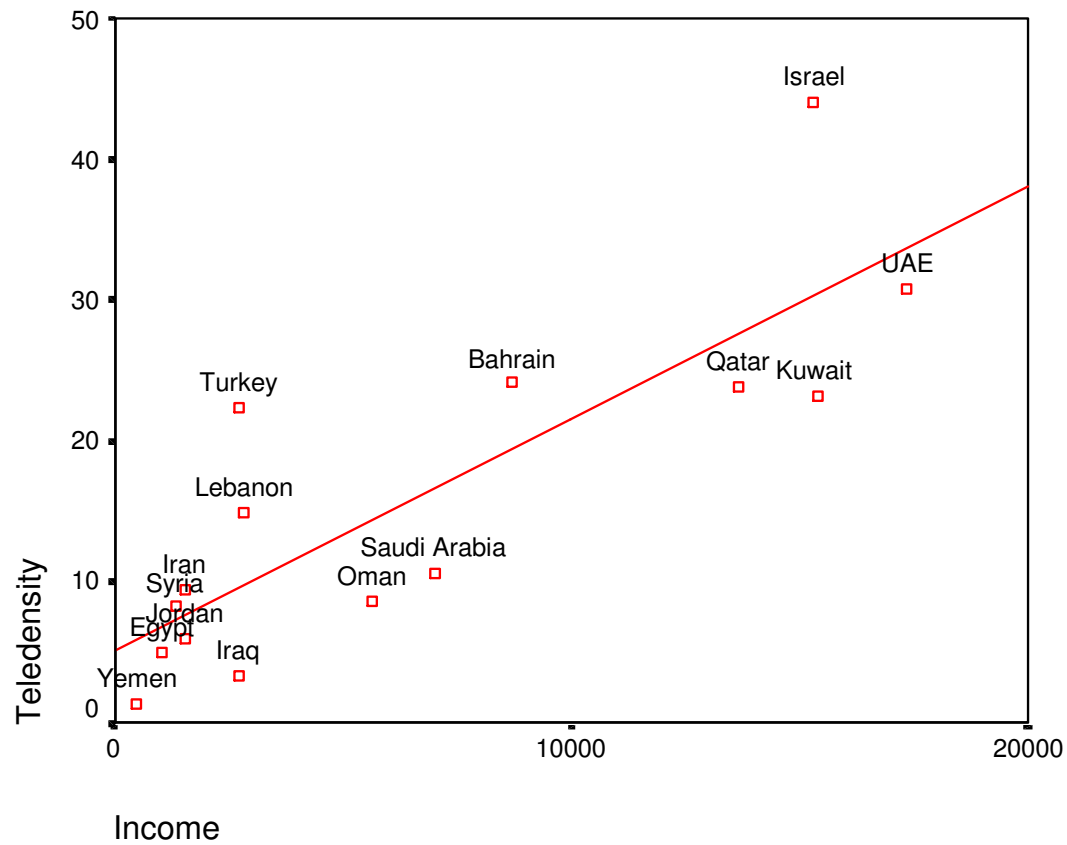
Income (gross domestic product) was the independent variable

15 countries from the Middle East

GDP per capita explains 85% of a countries teledensity level

Introduction

► Background



Introduction

► Background

Nine countries **under** the regression line have actual teledensity levels that are less than their predicted values

Yemen

Egypt

Jordan

Iraq

Oman

Saudi Arabia

Qatar

United Arab Emirates

Kuwait

Represent best opportunities for producing increased telephone utilization

Introduction

► Background

HSA computed the **residual value** for each country

Observed teledensity - Predicted teledensity

Negative residual indicates underperformance with respect to telephone utilization

Introduction

► Background

<u>Country</u>	<u>Residual</u>	<u>Country</u>	<u>Residual</u>
Kuwait	-7.6	Syria	0.9
Iraq	-6.2	Iran	1.9
Saudi Arabia	-5.9	Bahrain	4.8
Oman	-5.7	Lebanon	5.2
Yemen	-4.6	Turkey	12.8
Qatar	-3.4	Israel	14.0
UAE	-2.6		
Egypt	-1.8		
Jordan	-1.6		



All nine countries **under** the line have negative residuals

Introduction

► Overview

Extend HSA study in three ways

Model 208 countries using regression

Expand the set of covariates

Telecommunications

Socioeconomic

Political

Conduct a benchmarking exercise

Identify peer group for each country

Three partitioning methods

Aggregate results to rank countries

Variables

► Database

HSA database

29 variables for 208 countries

Telecommunications data

*World Telecommunications
Development Report, ITU, 1998*

Socioeconomic data

*World Development Report
World Bank, 1998*

Political data

*1998 Index of Economic Freedom
Heritage Foundation*

Variables

► Independent variables

With HSA input, identified 12 independent variables that are candidate regressors

Variable	Definition
Income	Gross domestic product per capita
Connection Charge	Cost to install a telephone line
Income Growth	Compounded growth of GDP per capita from 1990 to 1995
Main Line Growth	Compounded growth of the number of main lines from 1990 to 1995
Main Lines per Employee	Number of main lines divided by the number of telecom employees

Variables

► Independent variables

Variable	Definition
Outgoing Telephone Traffic	Minutes of international telephone traffic
Traffic Growth	Compounded growth of outgoing traffic from 1990 to 1995
Percent Residential	Proportion of telephones in residential use
Political Risk Ratio	Measures the economic freedom in a country Ratio of teledensity in the largest city to that in the entire country
Telecom Revenue per Main	Revenue generated by each telephone line
Percent Waiting	Percent of people on a wait list for a telephone

Variables

► Dependent variable

Variable	Definition
Teledensity	Number of main lines per 100 inhabitants of a country Most commonly used measure of the prevalence of telephone lines in a country Reasonable proxy for the size of a telephone market

Methodology

▶ Value approach

Identify countries that are **underperformers**

Assess levels of the covariates
linked to performance

Use **cross-sectional** data

Actual performance is below
estimated potential performance

Methodology

▶ Naïve estimation approach

Construct a **single** regression model

Relate teledensity to 12 covariates
for all 208 countries lumped together

Model may not be **robust** for a
group of similar countries

Ignores particularities of a
group that affect performance

Methodology

▶ Customized regression models

Compare a country against appropriate peer groups

Elaborate criteria may lead to very small peer groups

Including too many variates for a peer group **over fits** the model

Methodology

▶ Constructing peer groups: Three methods

1. Assemble all countries in one group

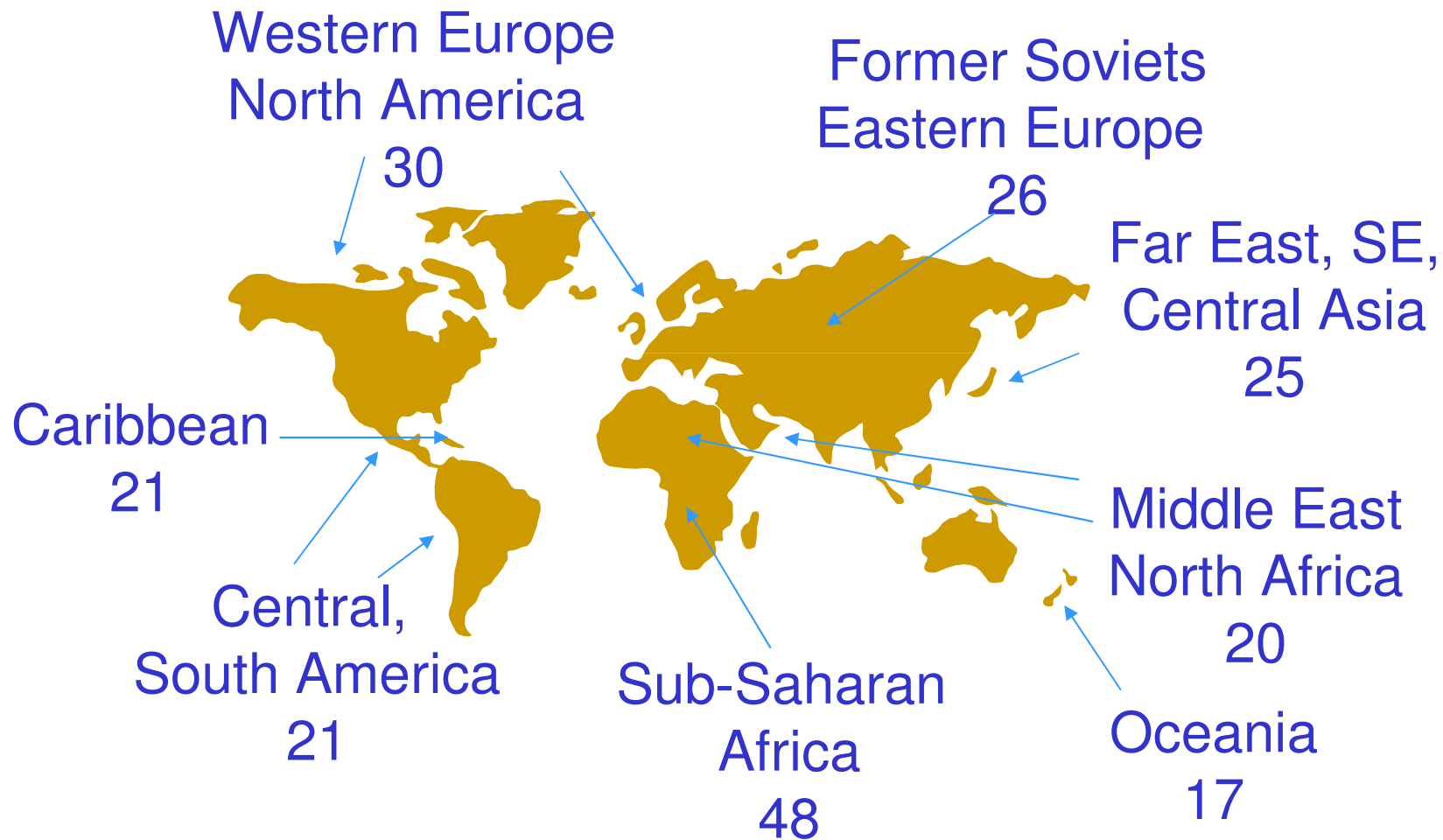
2. Partition on the basis of geography

Countries in a geographic group likely to have similar economic, social, and political dimensions

Eight geographic regions

Methodology

► Constructing peer groups: Geography



Methodology

▶ Constructing peer groups: Three methods

3. Form clusters using *k*-means algorithm

Reduced 12 correlated variables
to four orthogonal factors

Income

Industry characteristics

National growth

Telecommunication cost

Five clusters emerged

Methodology

► Constructing peer groups: Three methods

3. Form clusters using *k*-means algorithm

	Cluster				
	1	2	3	4	5
Number of Countries	16	30	58	47	57

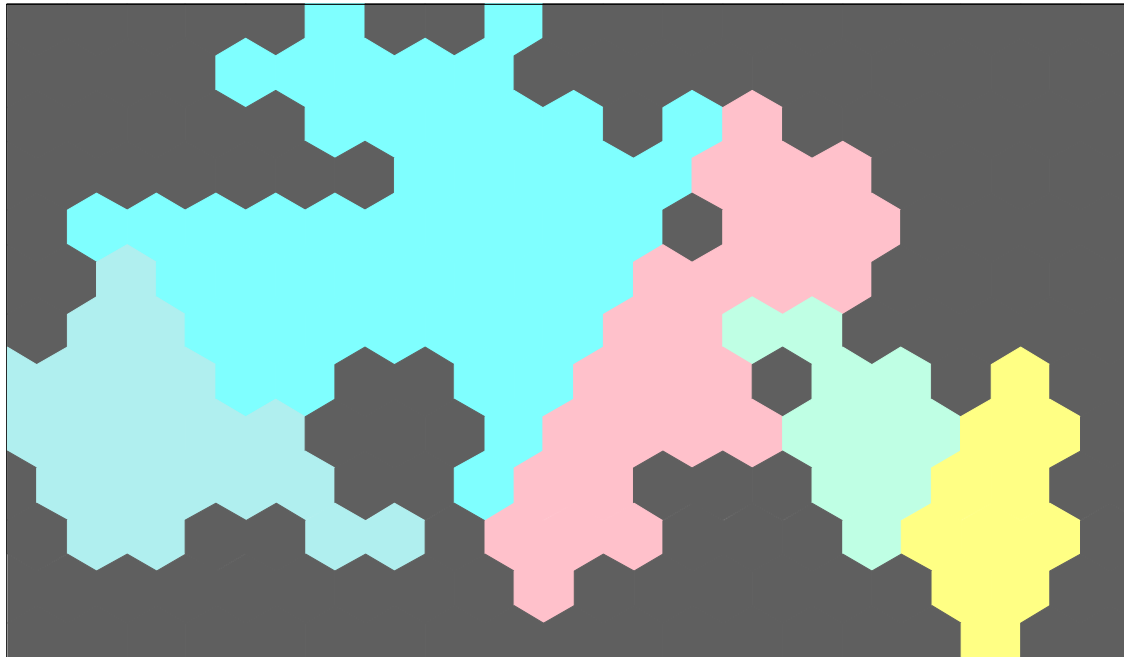
Countries in Cluster 1

Argentina	Croatia	Japan	Slovenia
Bangladesh	Finland	Mexico	Macedonia
Brazil	Greece	Paraguay	Uruguay
Columbia	Iran	Peru	Yemen

Methodology

- ▶ Constructing peer groups: Three methods

Number of clusters verified with [Viscovery](#)



Methodology

▶ Selecting covariates

Three regression models

1. **Simple model** of teledensity on income
2. **Quadratic model** with income and income squared
3. **Stepwise model** using all 12 covariates

Methodology

► Summary

14 peer groups

1 worldwide, 8 geographic, 5 clusters

3 regression models

Simple, quadratic, stepwise

Each country is in 9 different regression models

9 residual values

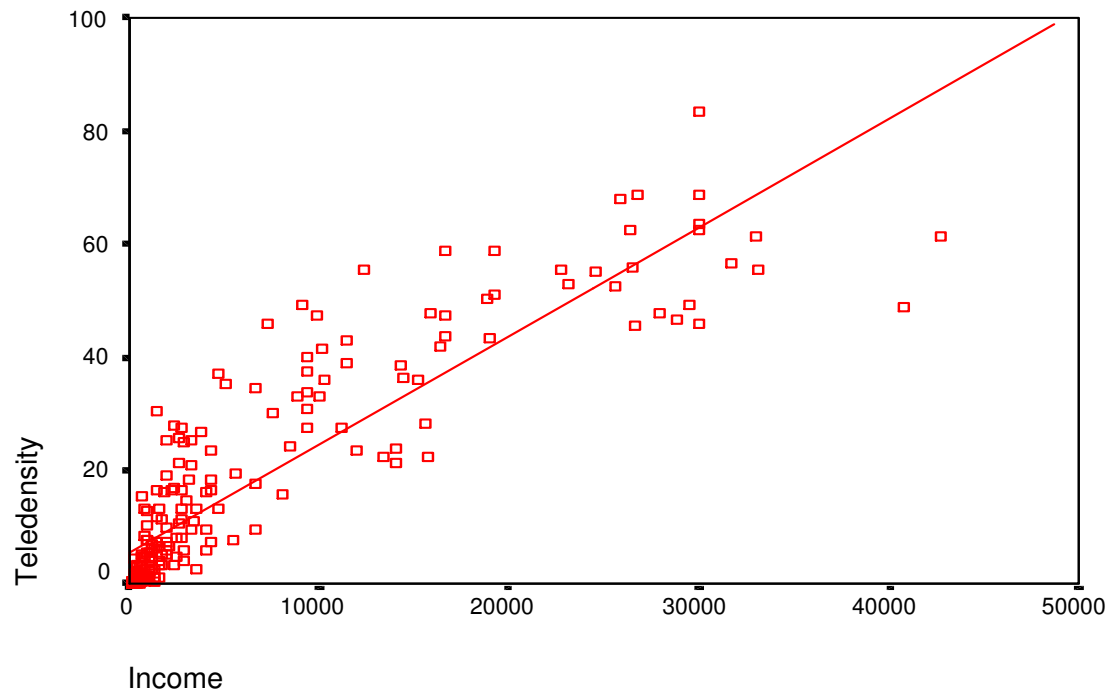
Modeling Results

► Worldwide models

Linear

$$5.59 + 0.00192 \text{ Income}$$

Scatterplot of Income vs. Teledensity



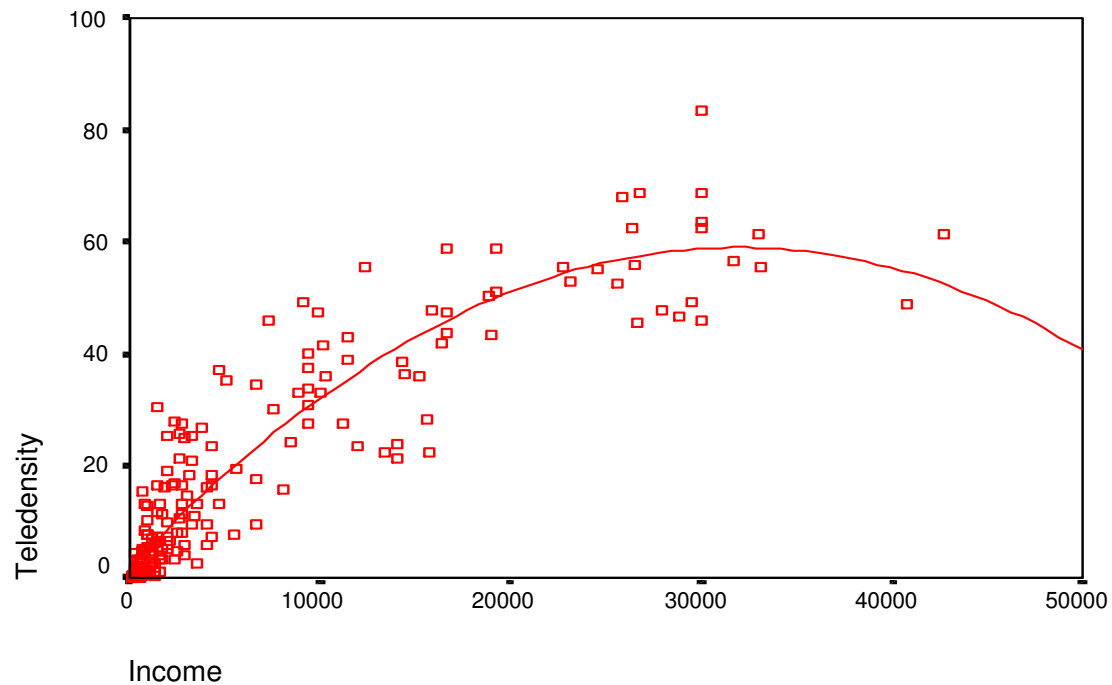
Modeling Results

► Worldwide models

Quadratic

$$1.99 + 0.00358 \text{ Income} - 5.61\text{E-}8 \text{ Income}^2$$

Scatterplot of Income vs. Teledensity



Modeling Results

► Worldwide models

Stepwise

3.00 + 0.00101 Income + 0.03 Traffic
+ 0.08 Main Lines per Employee
+ 0.60 Percent Waiting
+ 0.11 Percent Residential
- 0.25 Main Line Growth
- 2.34 Political Risk

All three models are statistically useful
 p -values < 0.05


Modeling Results

► Worldwide models: Country performance

Standardized residuals

Model	Austria	Brazil	Gabon	Japan
Linear	-1.561	-0.687	-1.090	-3.826
Quadratic	-1.589	-1.160	-1.534	-0.770
Stepwise	-1.337	-1.253	-0.749	-2.952

All residuals < -1
Consistent *underperformer*



Modeling Results

▶ Geographic models

Three models × Eight geographic groups

Each country appears in three models
that correspond to its group

Results for the Far East, Southeast, and
Central Asia group

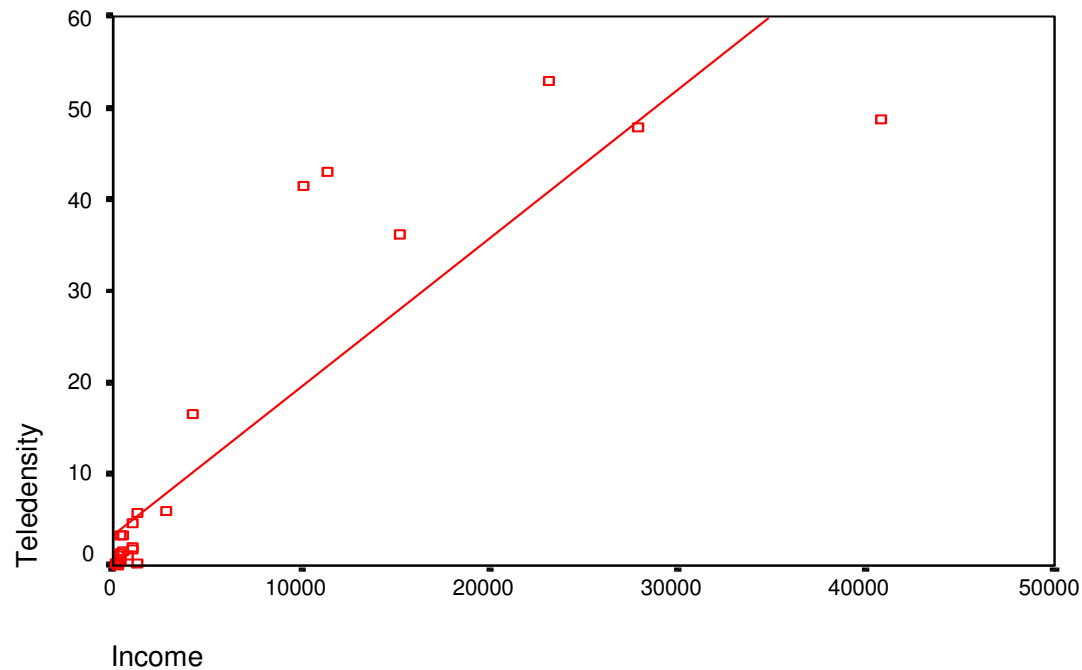
Modeling Results

► Geographic models: Asia

Linear

$$3.40 + 0.00162 \text{ Income}$$

Scatterplot of Income vs. Teledensity



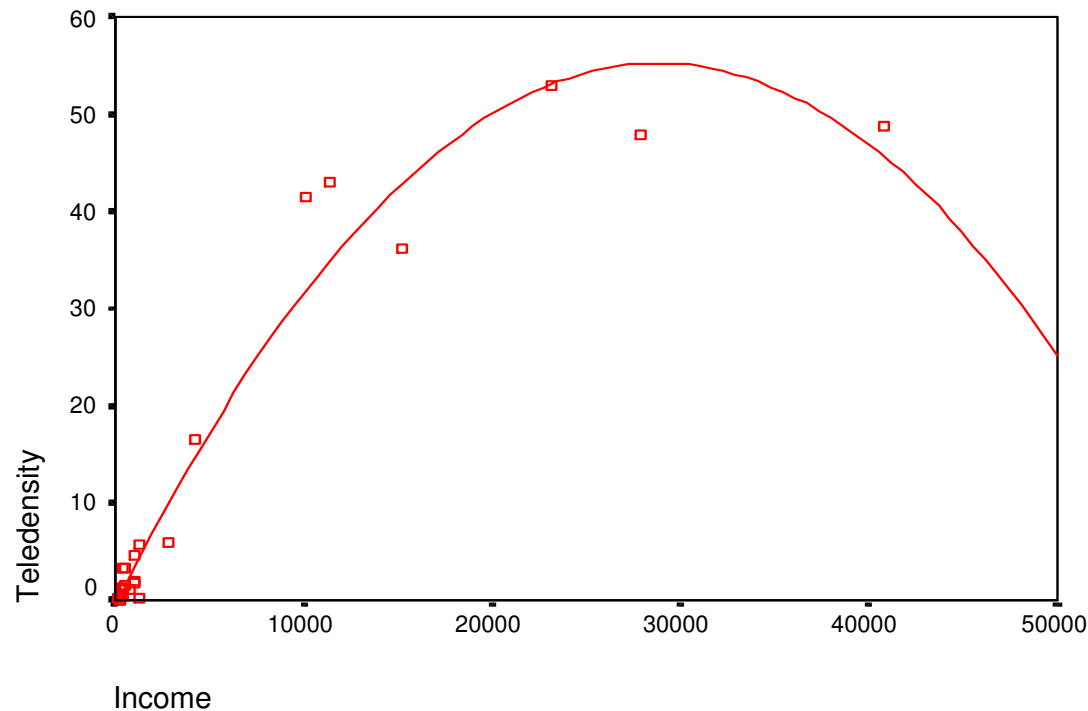
Modeling Results

► Geographic models: Asia

Quadratic

$$-0.32 + 0.00385 \text{ Income} - 6.69\text{E-}8 \text{ Income}^2$$

Scatterplot of Income vs. Teledensity



Modeling Results

► Geographic models: Asia

Stepwise

-2.00 + 0.000557 Income + 0.075 Traffic
+ 0.11 Main Lines per Employee

All three models are statistically useful

p -values < 0.05

Modeling Results

► Geographic models: Asia

Standardized residuals


Countries in 4 different groups



Model	Austria	Brazil	Gabon	Japan
Linear	-1.379	-1.374	-4.260	-2.359
Quadratic	-1.476	-1.291	-3.557	+0.826
Stepwise	-0.980	+0.534	-3.823	-0.509

All residuals $\ll -1$

Highly *underperforming* when compared to the 47 countries in Sub-Saharan Africa group



Modeling Results

▶ Cluster models

208 countries placed in five clusters using factor analysis and *k*-means analysis

Three models × Five clusters

Each country appears in three models that correspond to its clusters

Results for Cluster 1 (Brazil, Japan)

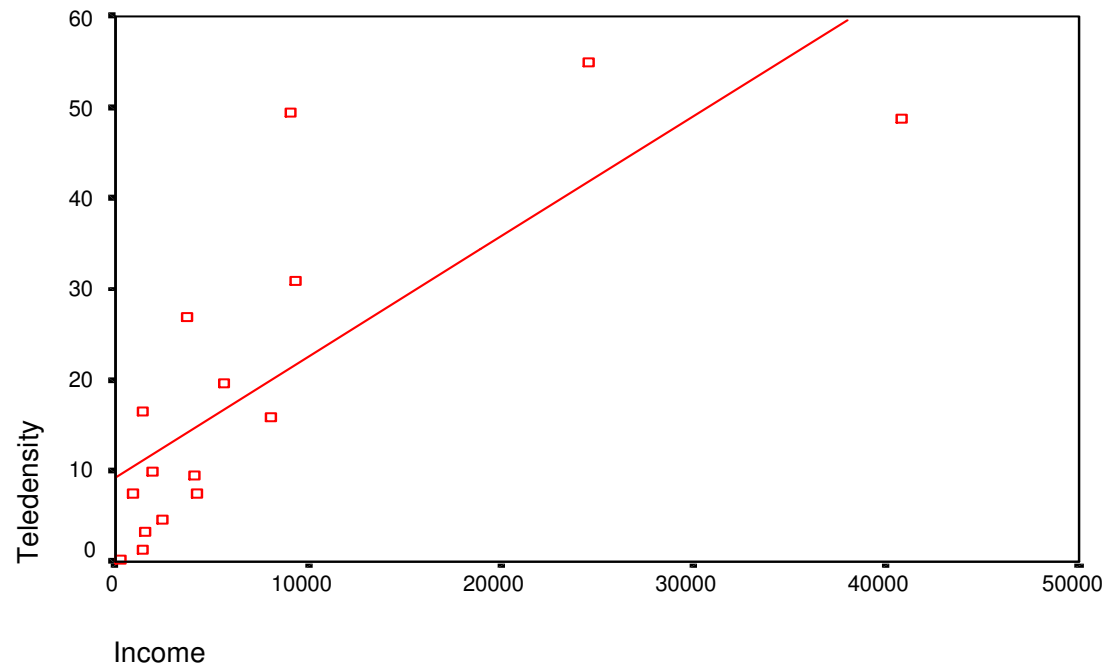
Modeling Results

► Cluster models: Cluster 1

Linear

$$9.30 + 0.00133 \text{ Income}$$

Scatterplot of Income vs. Teledensity



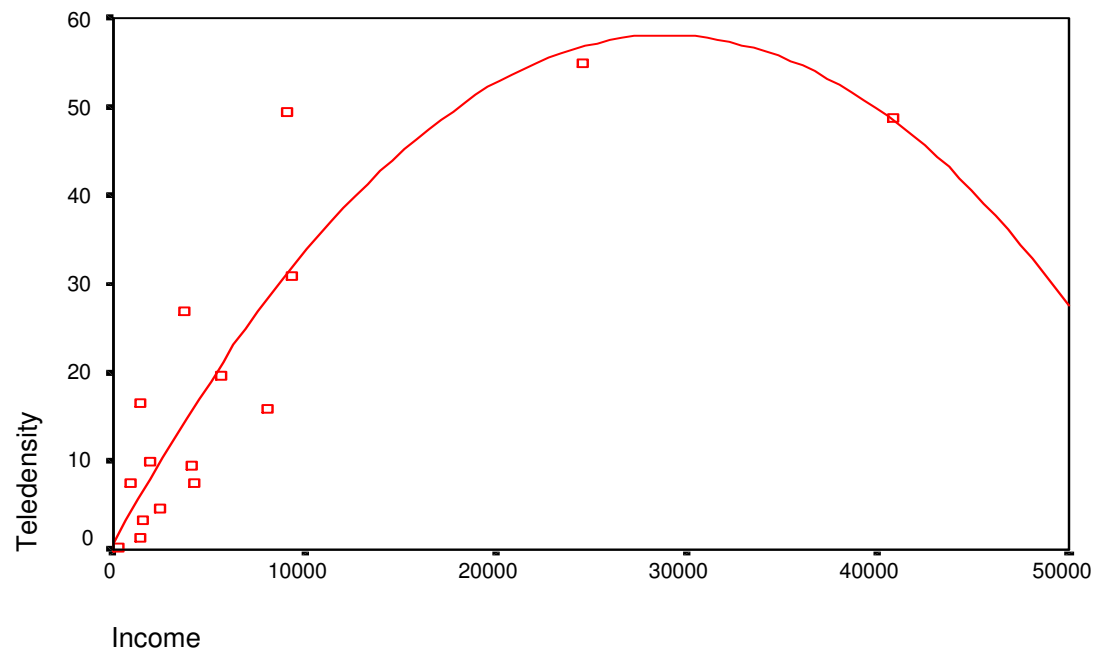
Modeling Results

► Cluster models: Cluster 1

Quadratic

$$0.74 + 0.00397 \text{ Income} - 6.86\text{E-}8 \text{ Income}^2$$

Scatterplot of Income vs. Teledensity



Modeling Results

► Cluster models: Cluster 1

Stepwise

$$2.83 + 0.000955 \text{ Income} + 0.52 \text{ Traffic}$$

All three models are statistically useful

p -values < 0.05

Modeling Results

► Cluster models: Cluster 1

Standardized residuals

Model	Austria	Brazil	Gabon	Japan
Linear	-1.099	-0.647	-4.483	-1.264
Quadratic	-1.113	-1.049	-4.181	+0.043
Stepwise	-1.095	-0.065	-4.097	+0.032

Countries in 3 different clusters

Consistently *underperforms* in worldwide, geographic, and cluster models

All residuals $\ll -1$
Highly *underperforming* when compared to the 57 countries in Cluster 3

Analyzing Results

- ▶ Combine results from different models: First cut

Each country appears in nine models

Focus on 63 countries that have at least one standardized residual less than -1

Average standardized residual

Weighted average standardized residual

Number of standardized residuals less than -1

Analyzing Results

► Combine results from different models: Gabon

Model		(1) Adjusted R ²	(2) Stand. Residual	(1)×(2)
Worldwide	Linear	0.791	-1.090	-0.862
	Quadratic	0.858	-1.534	-1.316
	Stepwise	0.882	-0.749	-0.661
Sub-Saharan Africa	Linear	0.884	-4.260	-3.766
	Quadratic	0.921	-3.557	-3.276
	Stepwise	0.948	-3.823	-3.624
Cluster 3	Linear	0.907	-4.483	-4.066
	Quadratic	0.907	-4.181	-3.792
	Stepwise	0.934	-4.079	-3.827
Total		8.032	-27.774	-25.190

Analyzing Results

- ▶ Combine results from different models: Gabon

Average standardized residual

$$U_{\text{Gabon}} = -27.774/9 = -3.086$$

Weighted average standardized residual

$$W_{\text{Gabon}} = -25.190/8.032 = -3.136$$

Number of residuals less than -1

$$N_{\text{Gabon}} = 8$$

Analyzing Results

- ▶ Combine results from different models: Second cut

Focus on countries that satisfy at least one of the following three criteria

Average standardized residual < -1

Weighted average standardized residual < -1

At least five standardized residuals less than -1

Analyzing Results

► Combine results from different models: Second cut

Country	W_i	U_i	N_i
Gabon	-3.14	-3.09	8
Kuwait	-1.71	-1.77	7
Botswana	-1.55	-1.53	7
French Polynesia	-1.47	-1.50	7
Saudi Arabia	-1.32	-1.42	6
Austria	-1.31	-1.29	8
Qatar	-1.20	-1.26	5
Singapore	-1.18	-1.12	5
Guam	-1.16	-1.15	6
Iraq	-1.15	-1.18	5
Japan	-1.14	-1.20	4
Belgium	-1.11	-1.12	5
Luxembourg	-1.01	-0.87	3
Marshall Islands	-0.99	-1.02	4
Argentina	-0.97	-0.99	6
Oman	-0.86	-1.00	3
Brazil	-0.76	-0.78	5

- First seven countries maintain ordering based on W_i and U_i

- Eleven countries satisfy all three criteria – good candidates for investment

Conclusions

► Summary

Partitioned 208 countries into three groups for benchmarking purposes

Large negative standardized residual indicated underperformance relative to peers

Aggregated results from nine regression models

Identified a candidate list of potential investment opportunities

Combine results with domain-specific knowledge (missing covariates?) to evaluate countries

