BUEC 333: An Introduction to Introduction to Econometrics

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What is Econometrics?

- *Econometrics* is the application of economic theory and statistical methods to analyze economic data.
- It's both a science and an art.
 - There's always a disconnect between economic theory and the real world. Creatively bridging this gap to apply economic theory to real-world data is an a
 - Economic data are rarely "perfect" (experimental) -sometimes requires some finesse to get believable results out of them.
 - Economic theory and statistics are grounded in mathematics. Direct application of either is scientific.

Econometric Questions

- *Magnitudes* and *Tests*
- Does reducing class size improve the quality of education?
 - Common sense tells us that students get more attention in smaller classes, and therefore probably learn more. But how much more? Reducing class size costs money (need more teachers). Does the improved quality of education justify the expense? Here, the *magnitude* of improvement matters.
- What will be the value of the S&P 500 one year from today?
 - Knowing (with certainty) the answer to this question would be worth a lot.
 - A good (accurate) estimate is also valuable.
- Do Immigrants Face Discrimination in the Labour Market?
 - This is a yes/no question. We can use statistics to *test* whether or not they face disparity.

So What?

- Knowing some basic econometrics can be very lucrative.
 - Lots of jobs, some pay very well.
- Knowing your way around some econometric software can too.
- Useful for future study (4th year courses -- you'll see/read quite a bit of empirical work, and maybe do some).
- Even if you end up doing something completely different, it's useful for understanding the world around you.
 - Managers need to understand where estimates come from and what they tell you (and what they don't!)
 - Helpful for understanding real-world statistics (polls, press about empirical studies, "facts" and arguments, etc.)
 - Helpful for understanding finance (alphas, betas, R², etc.)
- Good Policy Uses Econometric Analysis
 - Magnitudes Matter
 - Causation Matters

BUEC 333 Deliverables...

- In this class, you will learn the statistical methods you would use to answer these questions (and many more ...)
- You will learn how to assess whether a particular estimate/analysis is any good.
- You will learn to use a specific statistical software package (EViews)
- You will get some practice working with real data.

Today

- Correlation
- Magnitudes Matter
- Statistical Precision Matters
- One Dimension Isn't Enough
- Regression
- Causation
- Instruments, Natural Experiments and Field Experiments

Some Data

• Long Example: Do immigrants earn less?

- How? Why?

- 2006 Census public-use *microdata*, sample of from Census long forms.
- 3% sample of population of BC, about 100,000 observations of individuals
- Earnings from Wages and Salaries (ie., working for other people) in 2005

What is Microdata? (Excel)

ABOID	AGEGRP	AGEIMM	ATTSCH	BFNMEMB	CFINC	CFINC_AT	CFINEF	CFSIZE	CFSTAT
6	1	99	9	1	13	12	1	3	8
6	5	2	9	1	12	12	1	5	7
6	11	8	1	1	10	10	2	2	1
6	18	10	1	1	99	99	9	1	12
6	8	99	1	1	27	25	1	4	8
6	6	99	2	1	28	26	1	5	7
6	11	99	1	1	99	99	9	1	12
1	2	99	9	2	11	10	1	3	10
6	18	13	1	1	23	23	1	4	1
6	1	99	9	1	4	4	1	5	7
6	11	7	1	1	8	8	1	3	6
6	12	8	1	1	15	14	1	3	1
6	14	8	1	1	14	13	1	2	2
6	12	99	1	1	22	21	1	4	2
6	16	3	1	1	99	99	9	1	12
6	6	99	2	1	28	26	1	4	7
6	14	9	1	1	88	88	1	2	1
6	5	99	9	1	9	9	1	3	10
6	14	8	1	1	10	10	2	3	1
6	11	5	1	1	19	18	2	5	1

What is Microdata?

- But, there are 56,530 rows and 124 columns
- A bit cumbersome
- Need to summarize this information
 - Cross-Tabs
 - Tables of Mean
 - Regressions

Correlation

- When one number is high, do we expect another number to be high (or low)?
- When a person is an immigrant, are their earnings lower?
- Need to map out (at least) immigration status and earnings

The Cross-Tab

Cross-Tabs show the numbers of observations (rows of the data) that have values in different ranges for different variables. They show how the data are clumped, and how the data vary with each other.

	groupWAGES						
Immigrant status	Zero	1-\$20k	\$20к-\$50k	\$50k+	Total		
Temporary Residents Canadian-born Foreign-born Perm Res	527 7,289 8,814	239 6,212 5,132	115 6,520 4,878	72 4,881 2,355	953 24,902 21,179		
Total	16,630	11,583	11,513	7,308	47,034		

- Seems crazy to talk about the earnings of children, and the real elderly, and people who arrived after the 2005 income year
- So, drop
 - persons under 25
 - persons over 69
 - immigrants who arrived in 2005 or 2006

Immigrant status	Zero	1-\$20k	\$20к–\$50k	\$50k+	Total
Temporary Residents	274	162	103	69	608
Canadian-born	3,962	3,125	5,823	4,820	17,730
Foreign-born Perm Res	5,206	3,608	4,573	2,304	15,691
Total	9,442	6,895	10,499	7,193	34,029

Cross-Tabs with Column %s

	groupWAGES				
Immigrant status	Zero	1-\$20k	\$20K-\$50k	\$50k+	Total
Temporary Residents	2.90	2.35	0.98	0.96	1.79
Canadian-born	41.96	45.32	55.46	67.01	52.10
Foreign-born Perm Res	55.14	52.33	43.56	32.03	46.11
Total	100.00	100.00	100.00	100.00	100.00

Cross-Tabs with Row %s

	groupWAGES				
Immigrant status	Zero	1-\$20k	\$20к–\$50k	\$50k+	Total
Temporary Residents	45.07	26.64	16.94	11.35	100.00
Canadian-born	22.35	17.63	32.84	27.19	100.00
Foreign-born Perm Res	33.18	22.99	29.14	14.68	100.00
Total	27.75	20.26	30.85	21.14	100.00

Do Immigrants Earn Less?

- Temporary Residents have way more zeroearners, and foreign-born have somewhat more zero-earners
- Foreign-born permanent residents more lowearners and less high-earners than Canadianborns. The pattern is even stronger for temporary residents.

Integration Takes Time

- Immigrants are heterogeneous
- Some arrived long ago, some very recently
- Can we account for this in a Cross-Tab?

Do Immigrants Earn Less?

Immigrant		group	WAGES		
Cohort	Zero	1-\$20k	\$20K-\$50k	\$50k+	Total
Before 1950 1950s 1960s 1970s 1980s 1990s 2000s Cdn-Born	37.88 40.84 40.87 29.79 29.51 34.77 33.66 22.35	19.70 16.40 14.92 16.16 19.50 25.92 32.07 17.63	19.70 20.77 24.44 29.94 33.49 29.23 27.28 32.84	22.73 21.99 19.76 24.11 17.51 10.07 6.98 27.19	$ \begin{array}{c c} 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00 \end{array} $
Total	45.07	20.04	20.91	21.35	
ισται	27.00	20.11	10.01	21.20	100.00

Table of Means

- Only 7% of recent immigrants have high earnings, compared with 27% of Cdn-born.
- But, 24% of immigrants from the 1970s have high earnings.
- This is getting clunky: how else can we summarize these data?
- A *Table of Means* gives the average value of some variable for values or ranges of another variable

Table of Means

Immigrant Cohort	N(WAGES)	mean(WAGES)
Before 1950	66	27561
1950s	573	30682
1960s	1,260	31760
1970s	2,679	33365
1980s	2,416	27478
1990s	5,412	20283
2000s	2,463	17103
Cdn-Born	17,730	38317
Temporary	608	21737

Table of Means

- Those 2000s arrivals have it bad: \$20,000 lower earnings than Canadian-borns.
- 1990s arrivals have similarly low earnings.
- But, 1970s arrivals have earnings about \$5000 lower than Canadian-born.

Variation

- Used 3% of Greater Vancouver households.
- The mean is not the mean---it is an *estimate* of the mean.
- There's dispersion around the mean. Lots of people have income far from the mean.
- Standard Deviation (SD) characterises how far people are from the mean

- SD=the average squared distance from the mean

Table of Means and SDs

Immigrant Cohort	N(WAGES)	mean(WAGES)	sd(WAGES)
Before 1950	66	27561	38380
1950s	573	30682	58677
1960s	1,260	31760	65581
1970s	2,679	33365	49297
1980s	2,416	27478	34858
1990s	5,412	20283	31889
2000s	2,463	17103	24848
Cdn-Born	17,730	38317	56307
Temporary	608	21737	57811

- Immigrants arriving 1961-1970 and 1971-1980 earn almost as much as Canadian-born people
- Those arriving after seem to earn less and less
- But, are these differences statistically significant---that is, are we sure that they are there?

Precision

- The *precision* of an estimated mean depends on
 - How much data you have (N)
 - How much dispersion there is (*sd*)
- Standard Error (SE)
 - equals the standard deviation divided by the square root of N: SE = SD/VN
- 95% Confidence Interval (CI)
 - gives a range such that we're 95% sure that the true mean lies inside. These are roughly 4 SEs wide (2 above, and 2 below, the estimated value)

Table of Means, SDs and SEs

Immigrant Cohort	N(WAGES)	mean(WAGES)	sd(WAGES)	sem(WAGES)
Before 1950	66	27561	38380	4724
1950s	573	30682	58677	2451
1960s	1,260	31760	65581	1848
1970s	2,679	33365	49297	952
1980s	2,416	27478	34858	709
1990s	5,412	20283	31889	433
2000s	2,463	17103	24848	501
Cdn-Born	17,730	38317	56307	423
Temporary	608	21737	57811	2345

Another Way To Present the Means and Standard Error



So, Do Immigrants Earn Less?

- The 95% confidence interval for the mean earnings of a Canadian-born person aged 25-69 is entirely above that of the mean earnings of every immigrant cohort.
- A *test* considers the probability that some relationship is true.
- Can we test whether or not immigrants earn less? Given the probabilities, it is highly improbable that they earn the same as the Canadian-born. So, we reject the hypothesis that they earn the same.

Age and Arrival Cohort

- Immigrants arriving before 1961 and after 1980 earn less than Canadian-born workers.
- But, those arriving before 1961 are old.
- We know old people have low earnings, because they quit working.
- Those arriving between 1961 and 1980 are in their prime earning years.
- Is this all about age?

Immigrants Age and Earnings

Immigrant	Age Group					
Cohort	25-34	35-44	45-54	55-64	65-69	
Before 1950				31765	13267	
1950s			49199	36359	7537	
1960s		53708	42967	28865	10187	
1970s	37704	41336	33789	33291	6876	
1980s	31648	31057	30124	19169	4008	
1990s	20317	25139	22777	10365	2117	
2000s	18793	19564	15538	8810	2896	
Cdn-Born	31552	43511	47176	35218	10337	
Temporary	16820	26950	38551	13000	1333	

Do Immigrants Earn Less?

- Now that we consider people who are the same age, the disparities look smaller.
- Immigrants aged 25-34 who arrived in the 1970s and 1980 have average earnings higher than their Canadian-born counterparts.
- But, a lot of age-arrival year groups earn less.

Controlling for Age

- When we *control* for a variable *x*, we ask "what is the difference in *y* for a given value of *x*?"
- When you look **within** age categories, there's less disparity.
- Thus, *controlling* for age reduces the amount of disparity we see between immigrants and Canadian-born people

Controlling for Education

• Immigrants are more educated than Canadianborn people. Shouldn't they earn more?

Immigrant						
Cohort	LT HS	HS	Some PS	BA/BSC	Post BA	Total
Before 1950	13.64	16.67	36.36	1.52	31.82	100.00
1950s	16.75	27.92	30.89	4.89	19.55	100.00
1960s	13.67	24.09	33.07	6.36	22.81	100.00
1970s	13.14	22.10	30.20	7.61	26.95	100.00
1980s	17.00	23.91	24.70	8.81	25.57	100.00
1990s	15.22	23.35	19.93	9.29	32.20	100.00
2000s	10.68	17.63	13.89	9.42	48.38	100.00
Cdn-Born	9.21	26.27	32.47	5.36	26.69	100.00
Temporary	4.94	14.99	16.97	8.90	54.20	100.00
Total	11.41	24.37	28.02	6.82	29.38	100.00

Mean Earnings by Education

Immigrant	Highest Level of Schooling								
Cohort	LT HS	HS	Some PS	BA/BSC	Post BA				
Before 1950	10222	11000	43167	12000	26571				
1950s	17343	26142	29176	34143	49405				
1960s	13041	26711	34918	29211	44817				
1970s	15679	24290	34611	32363	48300				
1980s	11915	20463	29960	29356	41335				
1990s	9438	14238	21435	20271	29106				
2000s	8502	12737	15944	13242	21692				
Cdn-Born	20218	33702	35447	40512	52129				
Temporary	11533	13571	18226	12037	27577				

Controlling for Education

- Unlike when we controlled for age, when we control for education, the disparity gets
 bigger.
- So, is the disparity big or small? Should we control for age or education?
- You could make an even bigger table of means, or...we could run a regression.

Regression Analysis

- Tool to look at differences along many dimensions (ie., along age, education and immigration cohort all at the same time)
- Modeling Assumption: the effects of all these things can be summarised by adding up a bunch of individual effects.
- Then, assessing any one of them is easy: its effect on the total is isolated.

Analysing Regressions

- Cdn-Born, HS, aged 35-44 has estimated earnings of \$29,168 (top number)
- With a PhD (but still Cdn-Born aged 35-44), it would be \$23,459 higher, equalling \$52,627
- So, you can isolate the effect of having a PhD
- You can also isolate the effect of immigration arrival year

Regression Estimates

Variable	Estimate	Std Err	Variable	Estimate	Std Err
Cdn-born 35-44 HS	37629	742			
Before 1950	-3662	5999	25-34	-9657	766
1950s	952	2101	45-54	1692	737
1960s	-1538	1446	55-64	-7878	827
1970s	-4382	1020	65-69	-27324	1248
1980s	-10952	1057	LT HS	-7989	965
1990s	-18928	760	Some PS	3813	738
2000s	-25217	1059	BA/BSc	5838	1157
Temporary	-20328	2017	Post BA	17300	736

Part of a Regression

Regression are cool because you can look at **part** of a regression.

Consider the part with the immigrant cohort effects.

Holding age and education constant, if you compare a Canadian-born person with a Temporary resident, the latter earns \$20,328 less.

Immigrants from the 1960s earn only \$1538 less, and the confidence interval for that is about [-1500, 4500]: it could be zero dollars less!

Variable	Est	SE
Cdn-born 35-44 HS	37629	742
Before 1950	-3662	5999
1950s	952	2101
1960s	-1538	1446
1970s	-4382	1020
1980s	-10952	1057
1990s	-18928	760
2000s	-25217	1059
Temporary	-20328	2017

Multivariate Analysis

- With tables of proportions or frequencies (ie. Cross-tabs), you can see concentration.
 - Eg., concentration of immigrants at the low end of the income distribution
- You can also see that they are concentrated in the middle of the age distribution and the top of the education distribution
- Thus, *controlling* for these things is essential.

Multivariate Analysis

- Regressions allow you to control for lots of things at once.
- One-way analysis on the basis of age might tell you one thing (eg., earnings seem okay)
- One-way analysis on the basis of education might tell you another (eg., earnings are low)
- Multivariate analysis can resolve this by controlling for both at the same time.

R-squared

- The regression model has just three things: year of arrival, age and education.
- These three things 'explain' 7.5% of the variation in earnings from wages and salaries.
 - Explain, as in *are correlated with*
 - Not explain, as in *causes*
- What explains the other 92.5%?
 - Things that explain earnings that are correlated with arrival, age and education
 - Things that explain earnings that are not correlated with arrival, age and education

Men Earn More---So What

• Men earn more than women. What if we add "male" to the regression?

- 1970s immigrants have the same sex ratio as Canadian-borns

Immigrant	male		
Cohort	0	1	Total
Before 1950	45.45	54.55	100.00
1950s	49.57	50.43	100.00
1960s	52.34	47.66	100.00
1970s	51.92	48.08	100.00
1980s	53.08	46.92	100.00
1990s	54.27	45.73	100.00
2000s	55.94	44.06	100.00
Cdn-Born	50.10	49.90	100.00
Temporary	55.59	44.41	100.00
Total	51.74	48.26	100.00

Comparing Regressions

	Coef	SE	Coef	SE	Coef	SE
Cdn-born 35-44 HS	37629	742	29566	777	48207	1360
Before 1950	-3662	5999	-4387	5915	-5535	10462
1950s	952	2101	778	2072	1538	3807
1960s	-1538	1446	-1218	1426	-1345	2688
1970s	-4382	1020	-4114	1006	-7893	1890
1980s	-10952	1057	-10523	1042	-15131	1975
1990s	-18928	760	-18298	750	-25109	1433
2000s	-25217	1059	-24254	1045	-31833	2042
Temporary	-20328	2017	-19392	1989	-15805	3895
25-34	-9657	766	-9537	755	-14263	1418
45-54	1692	737	1849	726	2777	1362
55-64	-7878	827	-7778	815	-7198	1519
65-69	-27324	1248	-27016	1230	-35258	2312
LT HS	-7989	965	-8057	951	-11041	1789
Some PS	3813	738	3599	728	3929	1366
BA/BSc	5838	1157	6871	1142	6881	2295
Post BA	17300	736	16962	726	20588	1360
male			16172	526		
• .						

	Coef	SE	Coef	SE
Cdn-born 35-44 HS	37629	742	29566	777
Before 1950	-3662	5999	-4387	5915
1950s	952	2101	778	2072
1960s	-1538	1446	-1218	1 <i>4</i> 26
1970s	-4382	1020	-4114	1006
1980s	-10952	1057	-10523	1042
1990s	-18928	760	-18298	750
2000s	-25217	1059	-24254	1045
Temporary	-20328	2017	-19392	1989

- Check out just the top left panel: on the left there's a regression without *male*, on the right a regression with *male*.
- Adding *male* doesn't do anything
- This is because male, though important to earnings, is not very correlated with immigrant arrival cohort.

Just Men's Earnings

This is the rightmost panel: a regression with just men in it.

The same pattern emerges: immigrants from earlier cohorts have higher earnings, and immigrants from recent cohorts have very low earnings.

	Coef	SE
Cdn-born 35-44 HS	48207	1360
Before 1950	-5535	10462
1950s	1538	3807
1960s	-1345	2688
1970s	-7893	1890
1980s	-15131	1975
1990s	-25109	1433
2000s	-31833	2042
Temporary	-15805	3895
25-34	-14263	1418
45-54	2777	1362
55-64	-7198	1519
65-69	-35258	2312
LT HS	-11041	1789
Some PS	3929	1366
BA/BSc	6881	2295
Post BA	20588	1360
male		
just men	yes	

Endogeneity

- Adding Male to the regression changed nothing important (because it wasn't correlated with cohort)
 - But let's stick with just men.
- What about visible minority status? It is correlated.

Immigrant	VM		
Cohort	0	1	Total
Before 1950 1950s 1960s 1970s 1980s 1990s 2000s Cdn-Born Temporary	94.44 86.21 69.38 35.74 18.34 14.61 17.57 90.71 37.78	5.56 13.79 30.62 64.26 81.66 85.39 82.43 9.29 62.22	$ \begin{array}{c} 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00\\ 100.00 \end{array} $
Total	62.70	37.30	100.00

	Coef	SE	Coef	SE
White Cdn-born 35-44 HS	48207	1360	49041	1364
Before 1950	-5535	10462	-5361	10448
1950s	1538	3807	2343	3804
1960s	-1345	2688	998	2708
1970s	-7893	1890	-2463	2060
1980s	-15131	1975	-8222	2233
1990s	-25109	1433	-17912	1799
2000s	-31833	2042	-25021	2286
Temporary	-15805	3895	-11181	3952
25-34	-14263	1418	-13452	1421
45-54	2777	1362	2625	1360
55-64	-7198	1519	-7576	1519
65-69	-35258	2312	-35378	2309
LT HS	-11041	1789	-10752	1787
Some PS	3929	1366	3540	1365
BA/BSc	6881	2295	7352	2293
Post BA	20588	1360	20674	1358
vm			-9523	1444

Endogeneity

- Those numbers changed a lot!
- Visible minority status is correlated with time of arrival: immigrants in later cohorts were more likely to be visible minorities.
- Visible minorities have lower earnings.
- Thus, if you leave visible minority out of the regression, you think you are measuring a cohort effect, but it is really a visible minority effect.
- This is a form of *endogeneity*. If a coefficient suffers from endogeneity, then its estimated value is not informative of the effect of the variable of interest.

Correlation and Causation

- Two things make correlation different from causation.
 - missing variables: things (eg, earnings and arrival cohort) may be correlated, but that correlation may be driven by left-out or unobserved stuff (eg, country-of-birth).
 - *reverse causation*: things can be correlated, but you don't know which causes which.
 - Eg, if Chinese-born people fare well in B.C., then they may choose to move here. This means that their high wages *cause* their clustering here.

Causation and Endogeneity

 If two variables are correlated but are not causally related, the econometric problem is one of endogeneity (just as above).

Policy Needs Isolated Effects

- If we want to know *whether or not* recent cohorts fare badly, a table will do fine.
- If we want to know why they fare badly, we need to control out stuff
 - Eg., part of their ill-fortune is driven by their ethnic origin, not when they came.

Let's Educate Immigrants

Education pays off no matter how you slice it, controlling for age or not.

Should we invest in educating immigrants as a matter of policy?

	Coef	SE	Coef	SE
LT HS	-11041	1789	-10752	1787
Some PS	3929	1366	3540	1365
BA/BSc	6881	2295	7352	2293
Post BA	20588	1360	20674	1358
vm			yes	
just men	yes		yes	

Policy Needs Causal Effects

- Does the table tell what happens if we get more educated immigrants?
- A lot of missing variables here:
 - Quality of education; relevance of education; place of education.
 - Are people who get educated smarter than people who don't? If so, then educating a given person won't raise their wage as much as the difference in earnings might suggest.

Instrumental Variables

- Education is endogenous--that is we don't really believe that we have included all the correlated missing variables.
- We could use an *instrument* to solve this problem.
 - An *instrument* is a variable the is correlated with the problem variable (education in this case) but not with all those correlated missing variables
 - Bit too tricky to define right now.

Experimental Approaches

- The idea of experiments, in the lab, in the field, or natural experiments, is:
 - hold everything constant while changing something of interest
 - Measure the difference across its values
 - Since nothing else is changing, there are no missing variables.
 - Since the experimenter is changing the values of the variable, there is no reverse causation.

Natural Experiments

- Lab experiments don't work so well for economic questions, eg, do immigrant arrivals affect native-born wages? Need a big lab.
- A *natural* experiment addresses reverse causality: use a change that for sure is not driven by your supposed dependent variable.
 - Eg, the Mariel Boatlift put more than 100,000 Cubans into Florida.
 - This was not due to high wages in Florida; it was due to Castro emptying prisons.
 - It had no effect on native-born white or hispanic wages (David Card 1990)

Field Experiments

- Often, you don't get the natural experiment you wish for.
- A field experiment uses data created by the investigator in such a way that there is little or no missing variables problem.
- Phil Oreopoulous (2009) (in MBC-supported research) sent out 8000 fake resumes to figure out whether and why immigrants have trouble finding work.

Oreopoulous' Study

- Field experiment with 8,000 mock resumes, emailed in response to newspaper and online job postings
- Randomize name, experience, education, and other characteristics, send, and compare callback rate differences
- Offers clear and convincing evidence on why recent immigrants fare poorly in the Canadian labour market

Example: Different experience

2390 Credit Valley Road, Mississauga, ON, L5M 4E6, (905) 901-3811, tara.singh45@yahoo.ca

Tara Singh

Professional Summary

- · Motivated professional with demonstrated analytical abilities and investment research skills.
- Highly-developed planning and analytical skills from 5 years of relevant experience.
- · Reliable, dependable, self motivated, flexible and efficient
- Outstanding knowledge of Microsoft Office.
- Able to work extremely well under pressure.

Experience

KPMG Inc	2006 - 2007	Mumbai, India
0		

 Senior Accountant
 Full cycle of accounting including payroll, accounts payable/receivable, account reconciliations and period end closing, reported to CFO Formulated and graphed monthly quarterly sales analysis spreadsheets to refocus sales activity and achieve a 10% gain in sales. Designed daily cash flow report summarizing inflows and outflows to numerous bank accounts resulting in 5% saving in cost of funds. Drafted the GL procedure

manual, automated the month end reporting process instead of manual, designed and analyzed an efficient spreadsheet for management report.

KPMG Inc.	KPMG Inc.	2005 - 2006	IV
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Accounting Supervisor

 Supervised 10 A/P and A/R; followed up payment to projects and maintained daily accounts payable system; Reorganized A/R system Maintained daily transaction records, processed invoices, tracked expenses, filing etc. Reconciled bank statements, directed all cash activities, prepared tax documents and annual financial statements.

Blue Star Infotech Ltd. 2004 - 2005

Accountant

 Coordinated distribution of invoices and classified transactions. Augmented the A/P and A/R process, posted to A/R and A/P journals, prepared drafting P&L statements and monthly balance sheet consolidation.

Mumbai India

Education

ndian Institute of Management	2000 - 2004	Bangalore,	India
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Bachelor of Science, Economics

Additional Interests and Activities

World Traveller

Travelled to thirty-one countries on five continents.

Big Sisters

Mentor for disadvantaged youth.

Other Activities

Competitive squash player, classical piano player, recreational photographer

2390 Credit Valley Road, Mississauga, ON, L5M 4E6, (905) 901-3811, tara.singh45@yahoo.ca

Tara Singh

Professional Summary

Motivated professional with demonstrated analytical abilities and investment research skills.

2006 - 2007

- Highly-developed planning and analytical skills from 5 years of relevant experience.
- Reliable, dependable, self motivated, flexible and efficient
- Outstanding knowledge of Microsoft Office.
- Able to work extremely well under pressure.

Experience

KPMG Inc.

Toronto, Canada

Senior Accountant

Full cycle of accounting including payroll, accounts payable/receivable, account reconciliations and period
end closing, reported to CFO Formulated and graphed monthly quarterly sales analysis spreasheets to refocus sales activity and activitieve a 10% gain in sales. Designed daily cash flow report summarizing inflows and
outflows to numerous bank accounts resulting in 5% saving in cost of funds. Drafted the GL procedure
manual, automated the month end reporting process instead of manual, designed and analyzed an efficient
spreadsheet for management report.

KPMG Inc.	2005 - 2006	Toronto, Canada
Accounting Supervisor		
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 Supervised 10 A/P and A/R; followed up payment to projects and maintained daily accounts payable system; Reorganized A/R system Maintained daily transaction records, processed invoices, tracked expenses, filing etc. Reconciled bank statements, directed all cash activities, prepared tax documents and annual financial statements.

Mackenzie Financial Corporation 2004 - 2005 Toronto, Canada

Accountant

 Coordinated distribution of invoices and classified transactions. Augmented the A/P and A/R process, posted to A/R and A/P journals, prepared drafting P&L statements and monthly balance sheet consolidation.

Education

Indian Institute of Management	2000 - 2004	Bangalore, India	

Bachelor of Science, Economics

Additional Interests and Activities

World Traveller

- Travelled to thirty-one countries on five continents.
- Big Sisters
- Mentor for disadvantaged youth.
- Other Activities
- Competitive squash player, classical piano player, recreational photographer.

Example: Different name

5354 Russell View Road, Mississauga, ON, L5M 5V8, (647) 477-2873, martin.john15@gmail.com

John Martin

Professional Summary

- Experienced in various business aspects; accounting, customer relations, computer training, sales, marketing, negotiations, presentations, and office operations.
- Highly motivated.
- Experienced multi-tasker.
- Analytical and detail-oriented, problem solver.
- Analytical and detail-onented, problem
 Efficient with time management skills.
- Excellent skills in Visual Basic HTML and Microsoft Office.

Experience

KPMG	Cor	n
	001	μ.

Toronto, Ontario

Financial Analyst

Assisted the Corporate Finance Director with budgeting and forecasting exercises. Identified, explained and
communicated variances for operating plans and latest forecast. Examined the feasibility of business projects
and prepare a plan of action based on financial analysis. Reconciled monthly bank statements entries via
AS400.

ZAC Marketing Inc 2004 - 2006	Toronto, Ontario
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2002 - 2004

2006 to date

Actuarial Analyst

Performed actuarial and statistics analysis of risk to provide the underwriting department with keys contract
valuation metrics. Developed actuarial models used for pricing and/or risk management. Performed
segmentation analysis on the behalf of insurance companies to determine best and worst performing
products/classes and recommend strategies for growing/correcting those areas as appropriate. Examined
expert risk reports on larges individual corporate risks. Improved the decision making process significantly and
the quality of internal statistical and technical reporting documents by creating an Access based program that
offered a wide range analyses of the company's portfolio of reinsurance contracts. Improved the average
technical account reconciliation time by more than fifty percent.

FGF Brands Inc.

Toronto, Ontario

Investment Analyst

Independently performed fundamental research on assigned securities (distribution sectors). Participated in
the decision making process with respect to portfolio management by making buy, sell and hold
recommendations. Analyzed and tracked key data and statistics related to individual stocks and portfolios.
Created and maintained financial models for stock and portfolio characteristics. Created an Excel-based application
that calculates more than 20 financial and operational ratios. Elected employee of the month four times.

Education

University of Waterloo

Waterloo. Ontario

Bachelor of Commerce, Accounting

Additional Interests and Activities

World Traveller: Travelled to thirty-one countries on five continents. Big Brothers: Mentor for disadvantaged youth. Other Activities: Competitive squash player, classical plano player, recreational photographer

1998 - 2002

5354 Russell View Road, Mississauga, ON, L5M 5V8, (647) 477-2873, zhang.long11@gmail.com

Zhang Long

Professional Summary

- Experienced in various business aspects; accounting, customer relations, computer training, sales, marketing, negotiations, presentations, and office operations.
- Highly motivated.
- Experienced multi-tasker.
- Analytical and detail-oriented, problem solver.
- Efficient with time management skills.
- Excellent skills in Visual Basic, HTML and Microsoft Office.

Experience

KPMG Corp.	2006 to date	Toronto, Ontario
Financial Analyst		

 Assisted the Corporate Finance Director with budgeting and forecasting exercises. Identified, explained and communicated variances for operating plans and latest forecast. Examined the feasibility of business projects and prepare a plan of action based on financial analysis. Reconciled monthly bank statements entries via AS400.

ZAC Marketing Inc	2004 2006	Toronto Ontario
	2004 - 2000	

Actuarial Analyst

Performed actuarial and statistics analysis of risk to provide the underwriting department with keys contract
valuation metrics. Developed actuarial models used for pricing and/or risk management. Performed
segmentation analysis on the behalf of insurance companies to determine best and worst performing
products/classes and recommend strategies for growing/correcting those areas as appropriate. Examined
expert risk reports on larges individual corporate risks. Improved the decision making process significantly and
the quality of internal statistical and technical reporting documents by creating an Access based program that
offered a vide range analyses of the company's portfolio of reinsurance contracts. Improved the average
technical account reconciliation time by more than fifty percent.

EGE Brands Inc.	2002 - 2004
	2002 - 2004

Investment Analyst

 Independently performed fundamental research on assigned securities (distribution sectors). Participated in the decision making process with respect to portfolio management by making buy, sell and hold recommendations. Analyzed and tracked key data and statistics related to individual stocks and portfolios. Created and maintained financial models for stock and portfolio analysis. Produced performance reports that include analyses of returns, risk, added value and portfolio characteristics. Created an Excel-based application that calculates more than 20 financial and operational ratios. Elected employee of the month four times.

Education

University of Waterloo 1998 - 2002

Waterloo, Ontario

Toronto, Ontario

Bachelor of Commerce, Accounting

Additional Interests and Activities

World Traveller: Travelled to thirty-one countries on five continents. Big Brothers: Mentor for disadvantaged youth. Other Activities: Competitive squash player, classical piano player, recreational photographer

Field Experiment Setup

- Respond only to internet or newspaper postings (e.g. Workopolis, Job Bank, Craigslist,...) May – Nov. 2008
- 4 resumes per posting, sent over 2-3 days:
- 0: Can. name, Can. education, Can. experience
- 1: foreign name, Can. education, Can. Exp.
- 2: foreign name, foreign education, Can. Exp.
- 3: 4: foreign name, foreign educ., some or all foreign exp., respectively

		Callback Rates by Resume Type (Difference Compared to Type 0) [Standard Error of Difference, * indicates sign. Diff. compared to prev. type] {Callback Ratio: Type 0 / Type}					
				Ethni	c Origin		
		English- Canada	India	China	Pakistan	Britain	India/China/ Pakistan
Туре 0	English Name Cdn Educ/Exp	0.158					
Type 1	Foreign Name Cdn Educ Cdn Exp		0.121 (-0.037) [0.019]* {1.31}	0.108 (-0.050) [0.018]*** {1.46}	0.11 (-0.048) [0.016]*** {1.44}	NA	0.113 (-0.045) [0.011]*** {1.40}
Type 2	Foreign Name Foreign Educ Cdn Exp		0.122 (-0.036) [0.022] {1.30}	0.094 (-0.064) [0.020] {1.68}	0.14 (-0.018) [0.027] {1.13}	0.129 (-0.029) [0.019] {1.22}	0.114 (-0.044) [0.014] {1.39}
Type 3	Foreign Name Foreign Educ Mixed Exp		0.075 (-0.083) [0.019]*** {2.11}	0.103 (-0.055) [0.021] {1.53}	0.078 (-0.080) [0.020]*** {2.03}	0.157 (-0.001) [0.023] {1.01}	0.088 (-0.070) [0.013]*** {1.80}
Type 4	Foreign Name Foreign Educ Foreign Exp		0.051 (-0.107) [0.017]** {3.10}	0.053 (-0.105) [0.018]*** {2.98}	0.052 (-0.106) [0.015]** {3.04}	0.141 (-0.017) [0.021] {1.12}	0.052 (-0.106) [0.011]*** {3.04}

Oreopoulous' Results

- Name discrimination likely illegal
 - "Subsection 11 (1) of the Code also establishes that the right of a person under Part I is infringed where a requirement, qualification or factor exists that is not a prohibited ground of discrimination, but that results in the exclusion or restriction of a group of persons who are identified by a prohibited ground of discrimination, unless the requirement or factor is reasonable and genuine in the circumstances, subject to hardship on the employer"
- For those with 4-6 years Can. Experience, minimal differences for foreign vs. Can education
- Lower returns to foreign experience not because of type of job or employer
- Brits do fine

Food For Thought

- One-way tables are not enough:
 - Missing variables may be correlated, and may in fact be driving things
 - Multivariate analysis can fix up missing variables problems, and allow you to get at causation
- Statistical precision matters:
 - if things look different, but are not statistically distinguished (eg, because they are only 1 std err apart), then they may not be different at all.
- Regression analysis is a multivariate tool that lets you think about each variable separately.
- Natural experiments, field experiments and instrument variables help with both missing variables and reverse causation, and thus allow you to get at causation.