Introduction to Stata 19

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Why Stata?New in Stata 19









- Accurate
- Easy to use
- Stata is a complete, integrated software package that provides all your data science needs—data manipulation, visualization, statistics, and automated reporting.



Why Stata

- Master your data
- Broad suite of statistical features
- Publication-quality graphics
- Automated reporting
- Truly reproducible research
- PyStata Python integration
- Trusted and Reliable
- Continuously updated
- Easy to use, to automate, and to extend
- Advanced programming
- Community-contributed features
- Real documentation and World-class technical support
- Cross-platform compatible
- Widely used and Vibrant community



New in Stata 19

What's new in Stata 19

New features in Stata 19

- Machine learning via H2O:
 Ensemble decision trees
- Conditional average treatment effects (CATE)
- High-dimensional fixed effects (HDFE)
- Bayesian variable selection for linear model
- Interval-censored multiple-event Cox model
- Bayesian quantile regression
- Panel-data vector autoregressive (VAR) model

- Correlated random-effects (CRE) model
- Bayesian bootstrap
- Control-function linear and probit models
- SVAR models via instrumental variables
- Instrumental-variables local-projection IRFs
- Latent class model-comparison statistics
- Sayesian asymmetric Laplace model
- Inference robust to weak instruments

- Meta-analysis for correlations
- Mundlak specification test
- Ob-file Editor: Autocompletion, templates, and more
- Graphics: Bar graph Cls, heat maps, and more
- Tables: Easier tabulations, exporting, and more
- Multiple datasets: Modify a set of frames
- Stata in French
- More



Introduction to Machine Learning (ML)



Introduction to Machine Learning (ML)

Modelling as "learning"

More generally, suppose that we observe a quantitative response Y and p different predictors, X_1, X_2, \ldots, X_p . We assume that there is some relationship between Y and $X = (X_1, X_2, \ldots, X_p)$, which can be written in the very general form



ML Techniques in Stata

- Several Stata command can be used in ML:
- rforest
- > svm
- boost (only for Windows)
- > sctree, srtree
- subset
- ≻ mlp2
- > Python related commands: r_ml_stata_cv and c_ml_stata_cv

Machine learning via H2O: Ensemble decision trees

- With the new h2oml suite in Stata 19, use machine learning via H2O to uncover insights from data when traditional statistical models fall short
- Use ensemble decision trees—gradient boosting machine (GBM) and random forest—to perform classification or regression.
- With the access to H2O's machine learning algorithms from within Stata, you can now harness the power of high-performance predictive models without leaving your familiar Stata environment
- With tools such as Shapley additive explanations (SHAP) values, partial dependence plots (PDPs), and variable importance rankings, GBM and random forest provide powerful predictions while maintaining explainability no tradeoffs needed



Machine learning via H2O: Ensemble decision trees

- The GBM algorithm works in a sequential manner, where each model in the sequence is trained to correct the mistakes of the previous model. This is achieved by focusing on the pseudoresiduals (the difference between the actual and predicted values) from the previous model
- Random Forest is a type of machine learning algorithm that helps us make predictions based on lots of decision trees. A decision tree is like a flowchart where you ask questions to split data into smaller groups, helping you predict an outcome (like yes or no, or a number)



Machine learning via H2O: Ensemble decision trees

[H2OML] h2oml — Introduction to commands for Stata integration with H2O machine learning (View complete PDF manual entry)

H2O.ai

Description

This entry describes commands for performing predictive analysis using H2O machine learning methods, specifically ensemble decision tree methods, in Stata. H2O is a scalable and distributed machine learning and predictive analytics platform that allows you to perform data analysis and machine learning. It provides parallelized implementations of many widely used supervised and unsupervised machine learning methods. For more details, see [H2OML] H2O setup, [P] H2O intro, and https://www.stata.com/h2o/h2o18/h2o_intro.html#what-is-h2o. For a software-free introduction to machine learning, see [H2OML] Intro.

Supervised learning

h2oml gbm	gradient boosting machine
h2oml gbregress	gradient boosting regression
h2oml gbbinclass	gradient boosting binary classification
h2oml gbmulticlass	gradient boosting multiclass classification
h2oml rf	random forest
h2oml rfregress	random forest regression
h2oml rfbinclass	random forest binary classification
h2oml rfmulticlass	random forest multiclass classification

Estimation results and postestimation frame

h2omlest	catalog	H20 (estima	tion results	
h2omlpostestframe	specify	fram	e for	postestimation	analysis



Conditional average treatment effects (CATEs)

- The new cate command goes beyond estimating overall treatment effects in your analysis of causal effects to estimating individualized and groupspecific ones. Compare different interventions and policies. Explore treatment-effects heterogeneity
- **cate** helps us answer questions such as the following:
- 1. Are the treatment effects heterogeneous?
- 2. How do the treatment effects vary with some variables?
- 3. Do the treatment effects vary across prespecified groups?
- 4. Are there unknown groups in the data for which treatment effects differ?
- 5. Which is best among possible treatment-assignment rules?



Conditional average treatment effects (CATEs)

- cate estimates conditional average treatment effects (CATEs), which are average treatment effects (ATEs) conditional on a set of variables for which the treatment effects may vary. Estimating CATEs allows us to study treatment-effect heterogeneity and evaluate treatmentassignment policies
- cate provides three different CATE estimates: individualized average treatment effects (IATEs), group average treatment effects (GATEs), and sorted group average treatment effects (GATESs)

[CAUSAL] cate — Conditional average treatment-effects estimation (View complete PDF manual entry)

Syntax

Partialing-out estimator

cate po (ovar catevarlist) (tvar) [if] [in] [, options]

Augmented inverse-probability weighting estimator

cate aipw (ovar catevarlist) (tvar) [if] [in] [, options]

ovar is a continuous outcome of interest.

catevarlist specifies the covariates of the CATE model -- the conditioning variables for the treatment effects. catevarlist may contain factor variables; see fvvarlist.

tvar must be a binary variable representing the treatment levels.



Conditional average treatment effects (CATEs)

- ***We open the assets3 dataset. We define a global macro, catecovars, that stores the names of the variables on which we will condition
- webuse assets3
- global catecovars age educ i.(incomecat pension married twoearn ira ownhome)
- **estimate the effect of 401(k) eligibility (e401k) on net financial assets (asset) to answer the following question
- cate po (assets \$catecovars) (e401k)
- ***use categraph histogram to draw a histogram of the predicted IATEs and see their distribution
- categraph histogram
- **To test whether the treatment effects are heterogeneous, use estat heterogeneity
- estat heterogeneity
- ***We use categraph iateplot to visually examine the change in the IATE function across educ levels
- categraph iateplot educ





High-dimensional fixed effects (HDFE)

- Linear models with high-dimensional categorical variables
- Absorb multiple high-dimensional categorical
- Absorb not just one but multiple high-dimensional categorical variables in your linear, fixed-effects linear, and instrumental-variables linear models using option absorb() with commands areg, xtreg, fe, and ivregress 2sls
- Similar to **reghdfe** Stata user-written command

Linear regression with multiple fixed effects. Also supports individual FEs with group-level outcomes reghdfe — Syntax Least-square regressions (no fixed effects): reghdfe depvar [indepvars] [if] [in] [weight] [, options] Fixed effects regressions: reghdfe depvar [indepvars] [if] [in] [weight], absorb(absvars) [options] Fixed effects regressions with group-level outcomes and individual FEs:

Title

reghdfe depvar [indepvars] [if] [in] [weight], <u>absorb(absvars indvar) group(groupvar) individual(indvar) [options]</u>

High-dimensional fixed effects (HDFE)

- *Import panle data
- webuse nlswork
- xtset id year
- ****Run linear regression without clustering
- reg ln_w tenure age
- ****Run linear regression with double clustering
- areg ln_w tenure age, absorb (race msp)
- **Run panel regression with double clustering
- xtreg ln_w tenure age, fe absorb (race msp)



Correlated random-effects (CRE) model

- Want coefficient estimates of time-invariant covariates in your panel-data model? Fit a random-effects model. Want to allow for correlation between covariates and unobserved panel-level effects
- We study the effect on wages of time-varying variables, such as age or tenure. At the same time, we are interested in the effects of time-invariant variables, such as race
- An FE model will omit any variable that remains constant across time and thus cannot fully answer our research question
- An RE model may yield inconsistent estimates because of the possible correlation between individual time-invariant heterogeneity and the regressors age and tenure



Correlated random-effects (CRE) model

- *Import panel data
- webuse nlswork
- ***********Run CRE regression
- xtreg ln_wage tenure age i.race, cre vce(cluster idcode)



Marginal Cox proportional hazards model for interval-censored multiple-event data

- Need to analyze event times from multiple types of events such as the onset of diabetes and hypertension? Don't know the exact event times?
- The new stmgintcox command analyzes such interval-censored multiple-event data and account for possible correlation between event times across the different events
- Stata 18 expanded the functionality of stintcox to support time-varying covariates (TVCs)
- The new stmgintcox command fits a marginal proportional hazards model to interval-censored multiple-event data. You can use this command with either singleor multiple-record-per-event data, and it supports TVCs for all events or specific ones
- It also offers flexible ways to specify models with event-specific covariates



Marginal Cox proportional hazards model for interval-censored multiple-event data

- Interval-censored multiple-event data (or, more precisely, interval-censored eventtimes data on multiple events) commonly arise in longitudinal studies because each study subject may experience several types of events and those events are not observed directly but are known to occur within some time interval
- For example, an epidemiologist studying chronic diseases might collect data on patients with multiple conditions, such as heart disease and metabolic disease, during different doctor visits.
- In these studies, researchers are often interested in evaluating the effects of certain factors on the event times
- However, analyzing interval-censored multiple-event data is challenging because none of the event times are exactly observed and the dependence structure between different event times is often unknown



Marginal Cox proportional hazards model for interval-censored multiple-event data

- ****Import data
- webuse aric
- list id event Itime rtime bmi diabp if id==91 | id==92, sepby(id) noobs
- ******Run marginal proportional hazards model to interval-censored multiple-event data
- stmgintcox age i.male i.community i.race bmi glucose sysbp diabp, id(id) event(event) interval(ltime rtime)



Panel-data vector autoregressive model

- With the new xtvar command, you can now fit a panel-data vector autoregressive (VAR) model to analyze the trajectories of related variables when you observe multiple units or panels over time
- VAR models have long been a staple of multivariate time-series analysis, but these models require relatively long series
- We can apply the same tools to panel data, using observations across panels to compensate for the shorter span typical of these data
- We can evaluate the model using moment- and model-selection criteria and Granger causality tests
- We can interpret results using impulse–response functions (IRFs)



Panel-data vector autoregressive model

Stata example:

****Import data

- webuse swedishgov.dta
- ***We specify the lags(2) option with the xtvar command to include two lags of each dependent variable in each equation
- xtvar grants revenues expenditures, lags(2)
- *** Perform a Granger causality test to see whether grants Gran
- vargranger
- ***obtain IRFs after fitting a panel-data VAR model
- irf create baseline, set(swedish_govt_irfs)
- irf graph irf, byopt(yrescale)





Inference robust to weak instruments

- Do you have weak instruments in your instrumental-variables (IV) regression?
- Use the new estat weakrobust command to perform reliable inference on endogenous regressors
- IV methods allow researchers to estimate causal relationships even when some explanatory variables are endogenous. IV methods exploit other variables instrumental variables—that are correlated with the endogenous variables but do not themselves suffer from endogeneity.
- A well-known problem with IV methods in practice is that when instruments are only weakly correlated with the endogenous regressors, inference can become unreliable even in relatively large samples



Inference robust to weak instruments

- ****Import data
- webuse hsng
- ***Instrumental variables 2SLS regression
- ivregress 2sls rent pcturban (hsngval = i.region), vce(robust)
- ***If we suspect that our instruments for hsngval are weak, we can perform a test on the coefficient of hsngval that is robust to weak instruments
- estat weakrobust
- We can also use the user-written command ivreg2



Meta-analysis for correlations

- Traditionally, MA focuses on two-sample binary or continuous data, where the outcome of interest is measured across two groups often labeled as the treatment and control groups
- For example, an MA may compare the efficacy of a new drug versus a placebo or the impact of two different educational interventions on student performance
- Sometimes, we may want to investigate the strength and direction of relationships between variables across multiple studies
- This is where the MA of correlations comes into play



Meta-analysis for correlations

- webuse pupiliq
- meta set stdmdiff se
- meta summarize
- meta forestplot



Do-file Editor enhancements

- The Do-file Editor now provides an even better and more customizable environment for coding in Stata with additional autocompletion, templates, improved code folding, and more
- Autocompletion of variable names, macros, and stored results: if you pause briefly as you type, suggestions of variable names from data in memory, macros, and stored results will appear in addition to the command names and existing words that appeared previously
- Templates: By popular request, you can now save time and ensure consistency when you create new documents
- Current word and selection highlighting: The Do-file Editor will now highlight all caseinsensitive occurrences of the current word under the cursor
- Show whitespace and tabs
- Navigator panel



New graphics features

- Stata 19 supports a new two-way plottype heatmap to create a heat map
- Stata 19 also has two other new two-way plottypes, rpcap and rpspike, which plot a value and range, such as high, low, and opening daily stock prices. To connect the high and low values, rpcap uses a capped spike, and rpspike uses a spike. Both use a marker for the opening value.
- Additionally, new features have been added to graph bar, graph dot, and graph box



New graphics features

- Stata example:
- ****Import data
- webuse cstemp
- ******Twoway heat map
- twoway heatmap temp month year, ylabel(1(1)12, valuelabel) ccuts(45(5)95)
- ****Import data
- sysuse sp500
- ****Plot high low open
- twoway rpcap high low open date in 1/15
- twoway rpspike high low open date in 1/15







- Building customizable and reproducible tables is now even easier in Stata 19
- The putdocx and putexcel table commands for creating and customizing tables in a single command now allows you to add a title and notes and export tables.
- It is also easier to create tables from a collection
- anova, oneway, and tabulate now work seamlessly with collect. And collect now allows you to query layout specifications and remove results from a collection
- Also refer to tabstat2excel and outreg2 Stata user-written commands



- Highlights
- New in putdocx
- Include bookmarks in paragraphs and tables
- Include alternative text to be read by voice software for images
- Include Scalable Vector Graphics (.svg) images
- New in putexcel
- > Freeze a worksheet at a specific row or column
- > Insert a page break at a specific row or column
- Insert a header and footer into a worksheet
- Include hyperlinks in cells
- Create a named cell range
- https://www.stata.com/new-in-stata/new-reporting-features/



- ******Export into word document
- ****Import data
- webuse nhanes2l, clear
- putdocx begin
- quietly: dtable age weight bpsystol tcresult tgresult i.sex, by(diabetes, tests) title(Table 1) continuous(age weight, test(none)) factor(sex, test(none)) nformat(%6.1f mean sd)
- collect style putdocx, layout(autofitcontents)
- putdocx collect
- putdocx save bwtreport, replace



- *********Export into Excel document
- ****Import data
- webuse census, clear
- foreach x of varlist pop death marriage {
- bysort region: egen avg_`x' = mean(`x')

```
•
```

- export excel region avg_* pop death marriage state using report2.xlsx, firstrow(variables) replace
- putexcel set report2.xlsx, modify
- putexcel sheetset, split(1, 4)
- putexcel save



- ********Export Tables into word and Excel document
- ****Import data
- clear
- webuse nhanes2l
- table hlthstat, statistic(frequency) statistic(percent) missing title(Table 1) note("Source: NHANES II") note("Sample includes blank but applicable and missing responses.") export(table.docx)
- table hlthstat, statistic(frequency) statistic(percent) missing title(Table 1) note("Source: NHANES II") note("Sample includes blank but applicable and missing responses.") export(table2.xlsx)



Stata in French

Dénoser Modifier Données Graphique	Statistiques Utilisateur Fenêtre Aide						
	Pérumés tableaux et tests	•					
	inesumes, tableaux et tests	-					
Histoire	Modèles linéaires et modèles annexes	· ·	Rayon de braquage (pieds)	1,	Variables		тļ;
 Filtrer les commandes ici 	Résultats binaires	•	Régression logistique		K Filtrer les variable	es ici	
# Commande	Résultats ordinaux	•	Régression probit		Nom	Étiquette	
Halve and Alternative States	Résultats catégoriels	•	Régression log-log complémentaire		modèle	Marque et modèle	
ll n'y a aucun élément à afficher.	Compter les résultats	•	Régression logistique conditionnelle	ľ	prix	Prix en dollars US	
	Résultats fractionnaires	►	Régression logistique exacte		mpg	Kilométrage (mpg)	
	Modèles linéaires généralisés Modèles de choix		Régression logistique asymétrique		rep78	Registre de réparation 1	
			Modèle probit avec covariables endogènes		hauteur	Hauteur sous plafond (
	Séries chronologiques	•	Péarsonian probit de control function		coffre	Capacité du coffre (pi ³)	
	Séries chronologiques multivariées Modèles autorégressifs spatiaux Données longitudinales/de panel Modèles à effets mixtes multi-niveaux	Modèle probit avec sélection d'échantillons Régression probit hétéroscédastique Probit avec endogénéité, sélection et traitement	Medèle prehit avec sélection d'échaptille		poids	Poids (Ib)	
			Mouele probit avec selection d echantilloñs		longueur	Longueur (po)	
			regression probit neteroscedastique		rayon	Rayon de braquage (pie	
			Probit avec endogeneite, selection et traitement		cylindrée	Cylindrée (po³)	
			GLM pour la famille binomiale		ratio_transmission	Ratio de transmission	
	Analyse de survie	Régression probit bivariée	Régression probit bivariée		Descriftée	A	п.
	Epidémiologie et sciences connexes	•	Régression probit bivariée apparemment sans rapport		Proprietes		*
	Covariables endogènes	•	Régression de panel	Þ.	< >		
	Modèles de sélection d'échantillons	•	Régression à effets mixtes à plusieurs niveaux		 Variables 		
	Inférence causale/Effets du traitement	•	FMM (modèles de mélange fini)		Nom		
	SEM (modélisation par équations structurelles) LCA (analyse des classes latentes) FMM (modèles de mélange fini) IRT (théorie de la réponse aux items) Analyse multivariée Analyse des données d'enquête	•	Régression non paramétrique		Tures		
		 Inférence Modèles i 	Inférence coursele/Effets du traitement	-	Format		
				-	Étiquette de valeur		
			Modèles inférentiels Lasso	•	Remarques	Aucune remarque	
		•	Régression bayésienne	•	▲ Données		
		Post-estimation	•	Cadre	default		
					Nom de fichier	auto_fr.dta	
	Apprentissage automatique H2O				Etiquette	données automob	iles de 1
	Lasso	•			Remarques	1 note	
	Meta-analyse				Observations	74	
	Imputation multiple		¥		Taille	3.1KB	
	Analyse non paramétrique	•			Mémoire	64MB	
	Statistiques exactes	•			Trié par	étranger	
:\Program Files\StataNow19	Rééchantillonnage	•				CAP NU	M INS
	Puissance, précision et taille de l'échantillon						
	Analyse bayésienne	•					
	Moyenne du modèle bayésien	•					
	Post-estimation						





- https://www.stata.com/new-in-stata/
- https://www.stata.com/why-use-stata/





THANKS !

Stata

We remain at your disposal to answer any questions . your Stata-related requests .

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