

Propensity Score 分析

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内容

- * **理論編**

- * なぜPSか？
- * PSとは？
- * PSを利用した分析法は？

- * **実践編**

- * ロジスティック回帰分析によるPS推定
- * PSモデルの評価
- * 主なPS分析法

臨床医学分野での PS研究のゴール

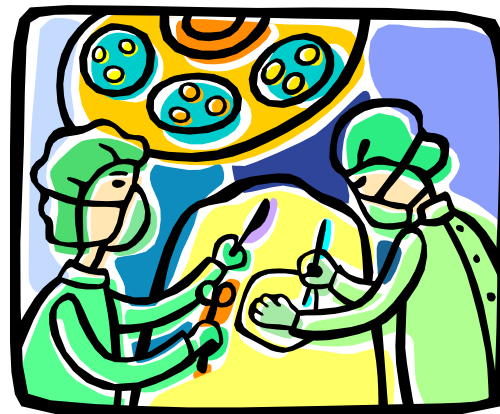
医学的介入の因果効果を検討する
質の高い研究

医学的介入

* 薬剤治療



* 手術



* 手技など

ランダム化比較研究：RCT

医学における比較研究の ゴールドスタンダード

1. ランダム化割り付け
2. (プラセボ)対照
3. 盲検
4. 試験



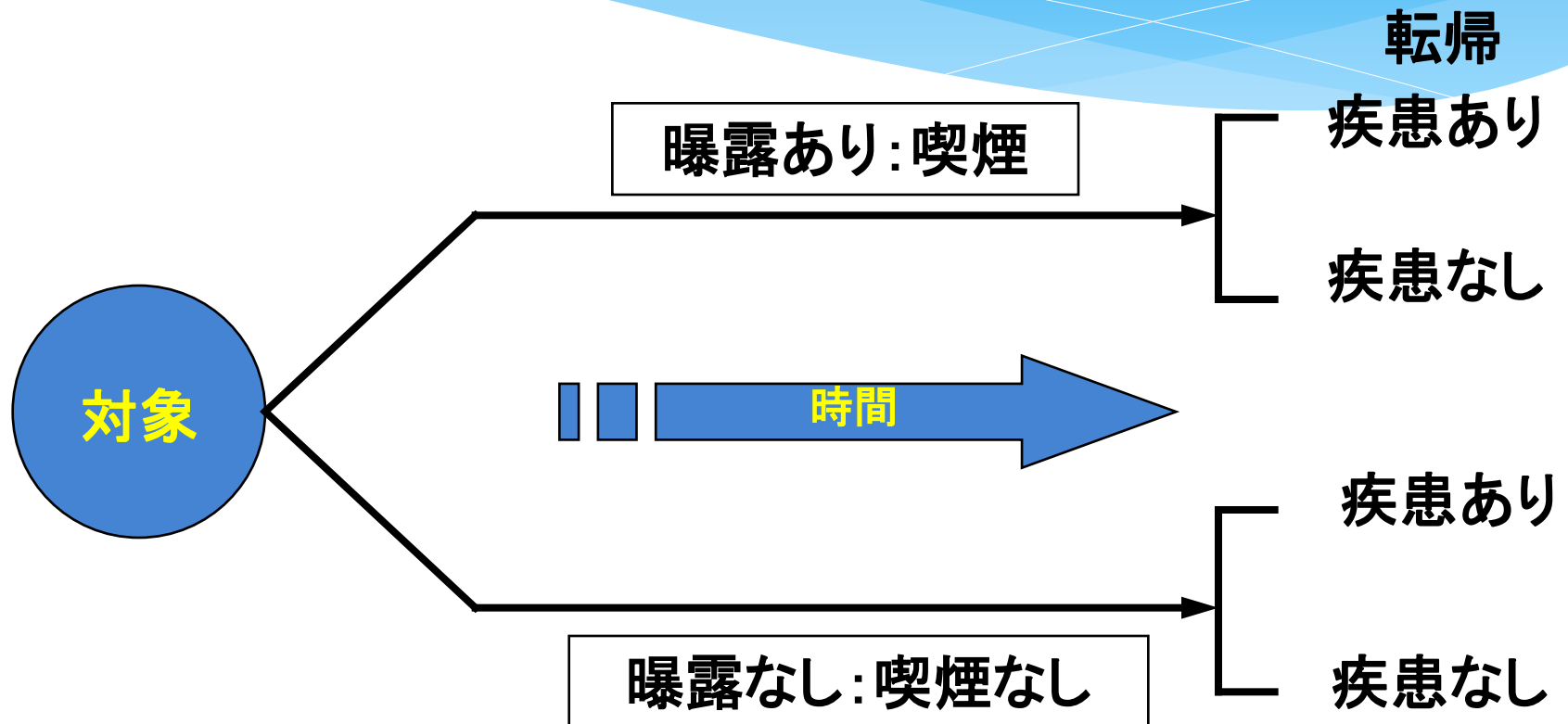
いつもRCT？

無作為化試験がいつも実行可能とは限らない：

- 倫理的(例：喫煙,)
- 計画的に見て実行不可能(時間、費用)
- 治療・手技がすでに日常診療に取り入れられている
(例：インフルエンザワクチン,手術*)

* Br J Surg. 2006 93(4):389

観察研究：コホート研究の構造



治療研究でのコホート研究の問題点

- 治療が非ランダムに割り付けられている
 - 医師は最も治療を必要としている患者を優先（適応がある患者・服薬コンプライアンス）
 - 治療割り付けのメカニズムは測定が困難で、ときにはまったく不可解(local rule)
- 様々な交絡因子 (Confounding) が存在する
- Unknown Confounderの調整ができない

交絡のコントロール

- 研究デザイン
 - 従来のデザイン(コホート、ケースコントロール)
 - 特殊デザイン(ケースクロスオーバーなど)
- データ分析方法
 - マッチング
 - 制限
 - 層別化
 - 多変量調整
 - 傾向スコア (Propensity score: PS) 手法

Klungel, et al. J Clin Epidemiol 2004

PSとは

- * 1983年にローゼンバウムとルービンが導入
- * 重回帰のようなモデルによる調整に代わる手法
- * PS手法は交絡変数の全情報を集約する方法

Rosenbaum & Rubin. *Biometrika* 1983

Propensity Score(PS)

複数の共変量を用いて2群に割り当てられる確率を予測するスコア

$$P(y_i) = \text{Probability (exposure} \mid x_i = \mathbf{X})$$

A propensity score $P(y)$ is the conditional probability of receiving the exposure given a set of observed covariates $\mathbf{X}$.

$$0 < P(y_i) < 1$$

PS手法の手順

PS手法は2段階の手順

PSの推定



分析

PS analysis using



1. PS modeling using Logistic regression
2. Analyzing using PS
 - Matching
 - Regression
 - Weighting
 - Stratification

PS modeling

Logistic regression: treatment+/(dependent variable)
& adjust for risk factors

SW, pe(0.05) pr(0.051):logistic VCM Age Gender

↑
(treatment)

predict PS

Evaluation of model

- Hosmer-Lemeshow: fitness
- ROCとAUC: discrimination

Graphics **Statistics** User Window Help

Summaries, tables, and tests ▶

Linear models and related ▶

Binary outcomes ▶

Ordinal outcomes ▶

Categorical outcomes ▶

Count outcomes ▶

Exact statistics ▶

Endogenous covariates ▶

Sample-selection models ▶

Multilevel mixed-effects models ▶

Generalized linear models ▶

Nonparametric analysis ▶

Time series ▶

Multivariate time series ▶

State-space models ▶

Longitudinal/panel data ▶

Survival analysis ▶

Epidemiology and related ▶

Survey data analysis ▶

Multiple imputation ▶

Multivariate analysis ▶

Power and sample size ▶

Resampling ▶

Postestimation ▶

Other ▶

Logistic regression

Logistic regression (reporting odds ratios)

Exact logistic regression

Mixed-effects logistic regression

Panel logistic regression

Probit regression

Probit regression (reporting change in prob.)

Probit regression with endogenous covariates

Probit regression with selection

Bivariate probit regression

Seemingly unrelated bivariate probit regression

Panel probit regression

Complementary log-log regression

Panel complementary log-log regression

GLM for the binomial family

Heteroskedastic probit regression

Skewed logit regression

Grouped data ▶

Postestimation ▶

Goodness-of-fit after logistic/logit/probit

Classification statistics after logistic/logit/probit/ivprobit

ROC curve after logistic/logit/probit/ivprobit

Sensitivity/specificity plot

Number of obs = 1876

LR chi2(6) = 422.18

Prob > chi2 = 0.0000

Pseudo R2 = 0.3050

	> z	[95% Conf. Interval]
.000	.1649678	.2948578
.000	2.181957	4.343028
.000	1.887426	3.794291
.001	2.052831	15.25842
.001	1.00495	1.020626
.003	.8141431	.9585805

on pr assumed; Pr(vcm_))
issing values generated)

p ps

byte %8.0g

int %8.0g

Command

Hosmer-Lemeshow

```
. estat gof, group(10) table
```

Logistic model for VCM, goodness-of-fit test

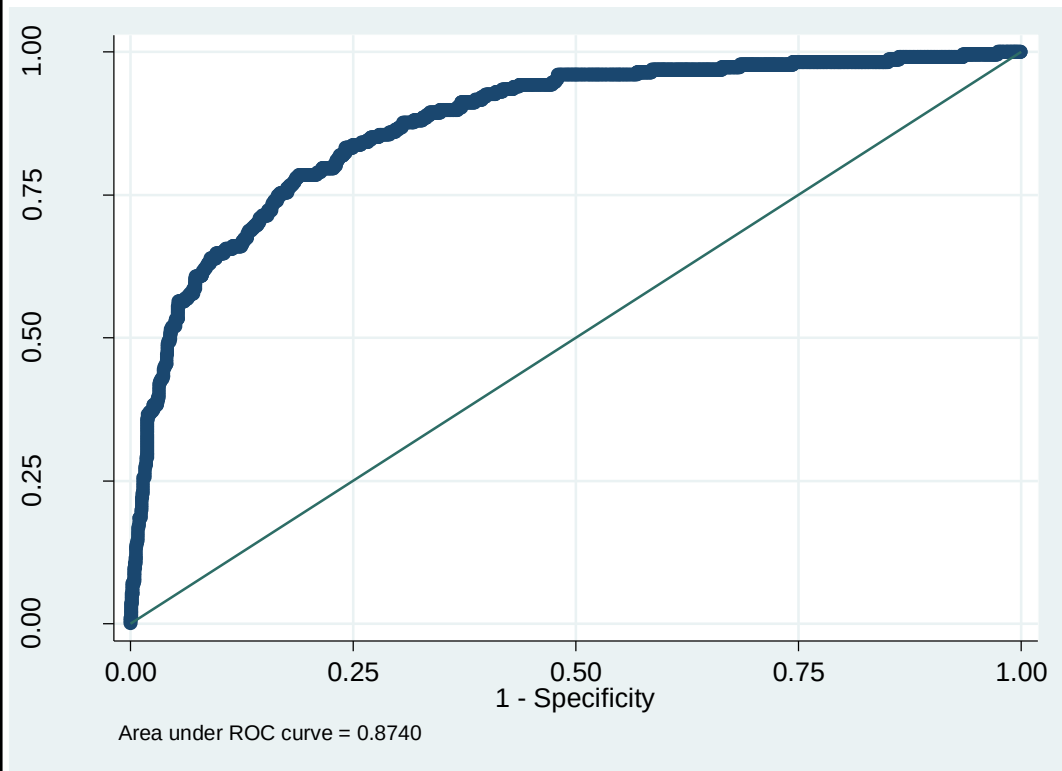
(Table collapsed on quantiles of estimated probabilities)

Group	Prob	Obs_1	Exp_1	Obs_0	Exp_0	Total
1	0.0099	2	1.3	186	186.7	188
2	0.0139	2	2.2	186	185.8	188
3	0.0202	3	3.1	184	183.9	187
4	0.0284	2	4.5	186	183.5	188
5	0.0415	4	6.4	183	180.6	187
6	0.0649	13	9.8	175	178.2	188
7	0.1075	19	15.8	169	172.2	188
8	0.1916	26	26.8	161	160.2	187
9	0.3591	42	48.7	146	139.3	188
10	0.9968	114	108.3	73	78.7	187

```

number of observations =      1876
number of groups      =        10
Hosmer-Lemeshow chi2(8) =        6.64
Prob > chi2           =      0.5760
    
```

ROCとAUC



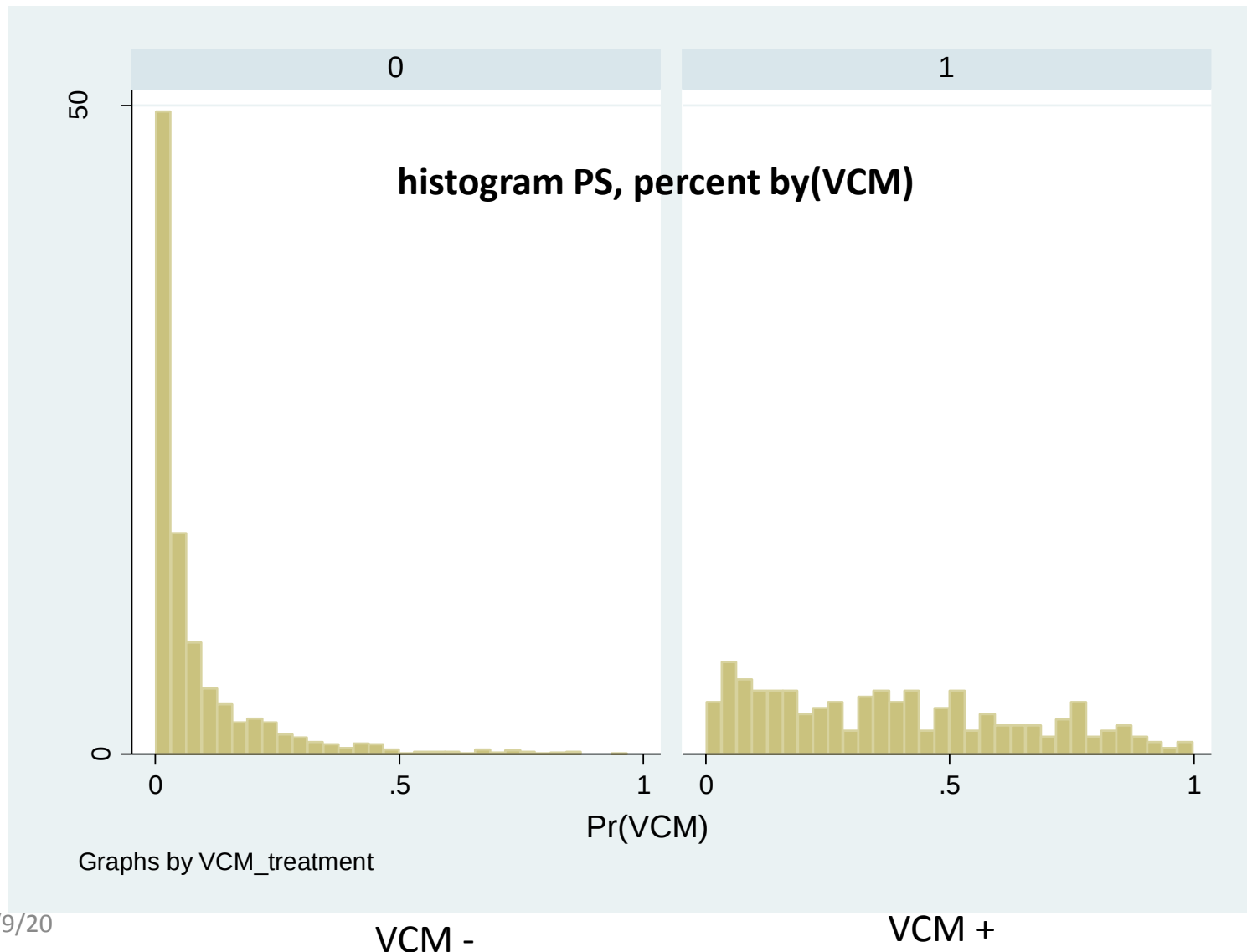
. lroc

Logistic model for VCM

number of observations = 1876

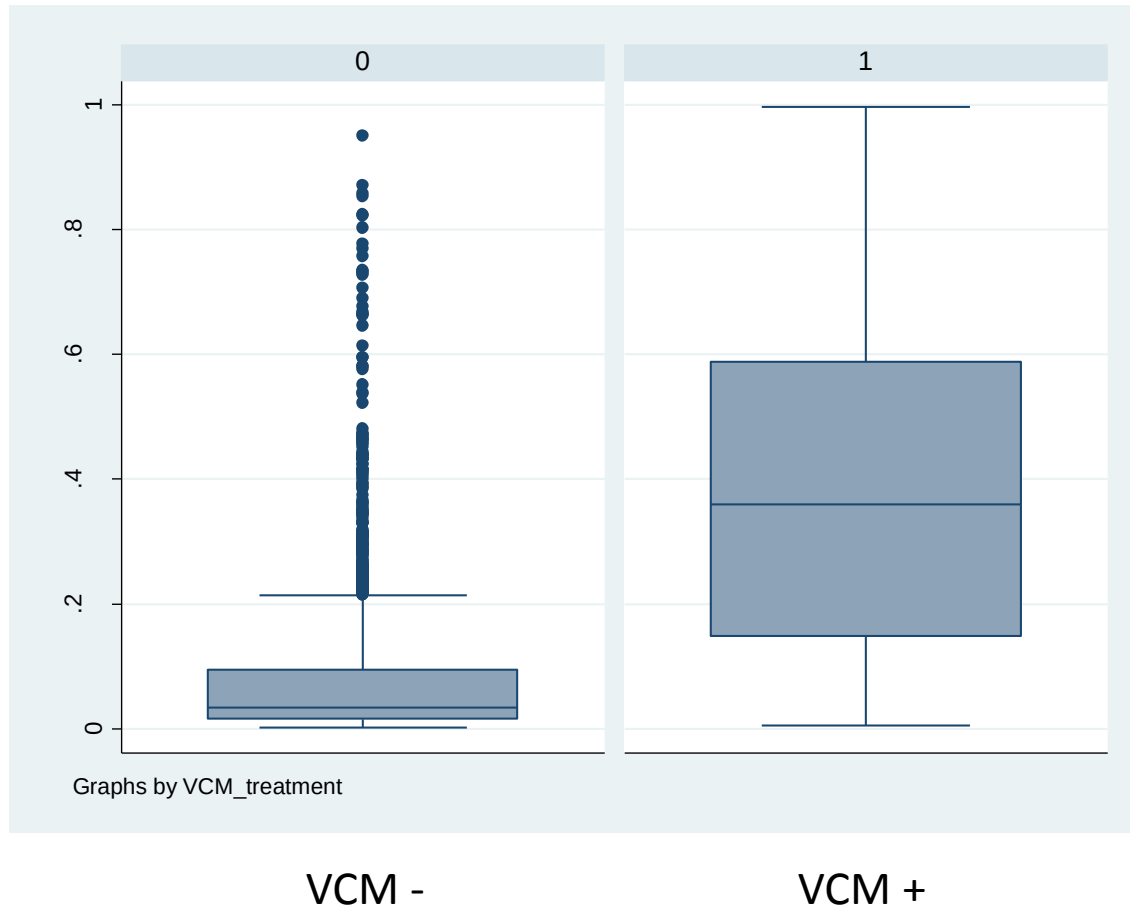
area under ROC curve = 0.8740

Histogram of Propensity Score

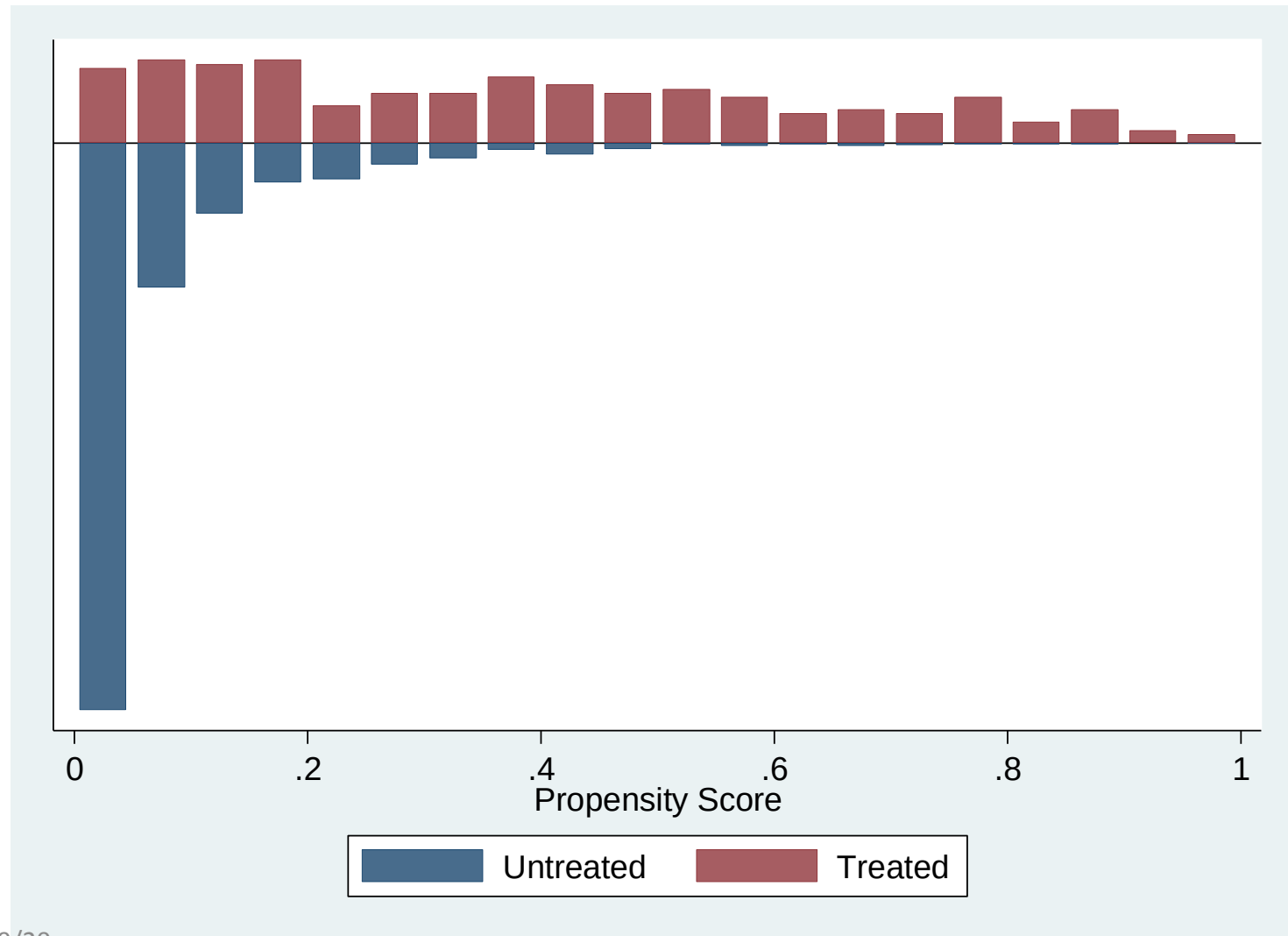


Box plot of PS by VCM

graph box PS, by(VCM)



Stata command: psgraph



Propensity Score (PS) Analysis

1st step

PS Modeling

2nd step

Matching
using PS

Regression
using PS

3rd step

Analyzing
with matched
data

PS **matching** using STATA

PS model

- Summary data: d, sum
- Modeling: Sw,pe()pr():logistic, Model checking: lroc, estat
- Overall PS checking: sum, graph (hist, graph box), psgraph
- Checking each variable: quintile (regress, logistic, ttest, tabulate, ptest)

Matching

- Psmatch2, ptest, psgraph
- teffects psmatch (Ver. 13)

Analyzing

- Regreession (clogit after reshape)
- clogit with other confounders

Matching in Stata

- [psmatch2](#) package created by Edwin Leuven and Barbara Sianesi.
- Matching: psmatch2 implements various types of propensity score matching estimators.
 - one-to-one, k -nearest neighbors, radius, kernel, local linear regression, spline, Mahalanobis

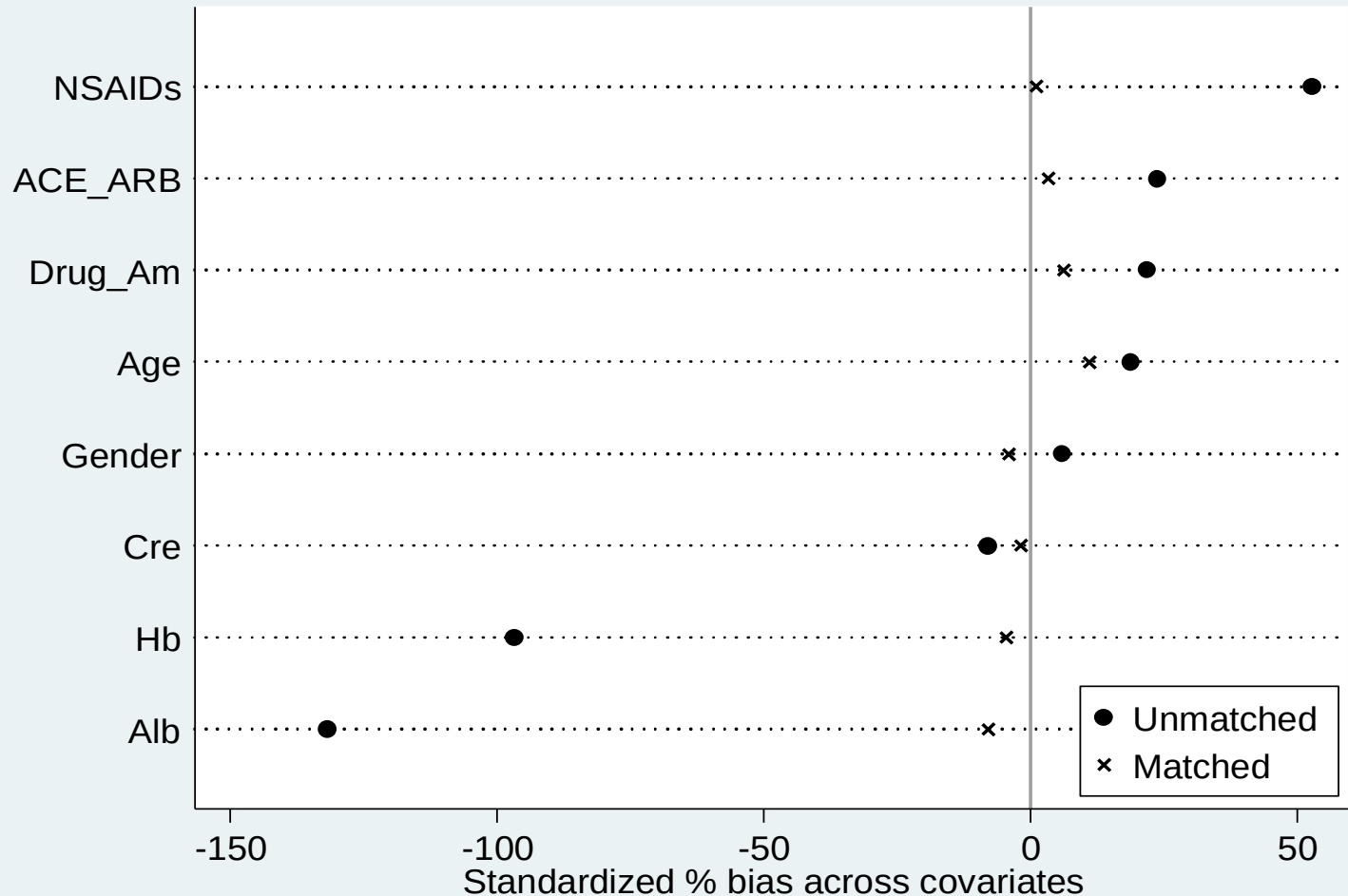
Nearest Neighbor Matching within a caliper

- Typically, one treatment case is matched to several control cases, but one-to-one matching is also common and may be preferred (Glazerman, Levy, and Myers 2003).
 - Caliper: $(SD \text{ of } \log(PS/1-PS)) \times 0.25$

Stata command

- Matching
 - **psmatch2** VCM, pscore(logitPS) caliper (0.4)
noreplacement descending
- Checking the balance after matching
 - **pstest** Gender Age Cre Hb Alb Drug_Am NSAIDs
ACE_ARB, both t(VCM) graph

Checking the balances after matching



Analysis after matching

- Conditional logistic analysis
 - **clogit** aki pair, group(group)
 - clogit, or
- McNemar test

PS regression using STATA

PS model

- Summary data: d, sum
- Modeling: Sw,pe()pr():logistic, Model checking: lroc, estat
- Overall PS checking: sum, graph (hist, graph box), psgraph
- Checking each variable: quintile (regress, logistic, ttest, tabulate, pstest)

Analyzing

- Regression (tabulate, K-M curve, ttest)
- Multivariate regression (Logistic, Cox, regress with other confounders)

1. Crude odds ratio

```
. logistic aki_outcome_doubling VCM, or
```

Logistic regression

Number of obs = 1992

LR chi2(1) = 96.88

Prob > chi2 = 0.0000

Pseudo R2 = 0.0772

Log likelihood = -578.67306

aki_outcome_doubling	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
VCM	<u>6.111996</u>	1.051483	10.52	0.000	4.362604	8.562889
_cons	.0729483	.0068979	-27.69	0.000	.0606076	.0878018

2. Adjusted odds ratio by Propensity score

```
. logistic aki_outcome_doubling VCM PS , or
```

Logistic regression

Number of obs = 1896

LR chi2(2) = 138.76

Prob > chi2 = 0.0000

Pseudo R2 = 0.1128

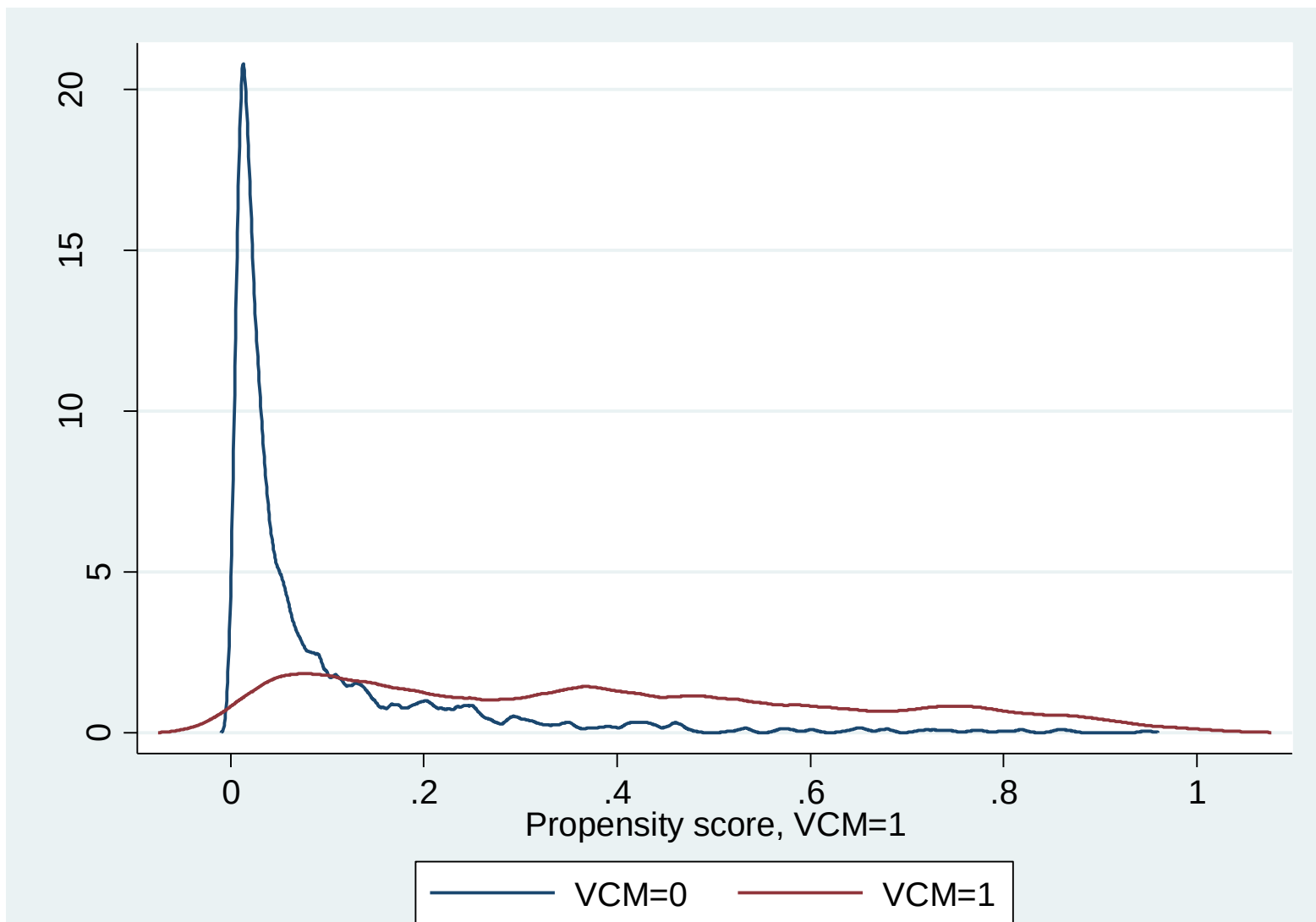
Log likelihood = -545.65981

aki_outcome_doubling	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
VCM	<u>2.495089</u>	.5405971	4.22	0.000	1.631778	3.815146
PS	<u>15.61008</u>	6.096279	7.04	0.000	7.260751	33.56054
_cons	.0570277	.0062395	-26.18	0.000	.0460208	.070667

Propensity scores – command summary

- *psmatch2*
- *pstest*
- *psgraph*
- *pscore*
- *teffects psmatch (stata13)*

Stata command: teffects psoverlap



まとめ

- 観察研究で治療効果を推定する際に有用
(特にイベント数が変数の7倍以下の場合)
- PS手法は2段階
 - 治療を従属変数、共変量を独立変数とするロジスティック回帰によってPSを推定する
 - PSを基に様々な方法で分析(マッチング、層別化、多変量解析、ウェイト付け)
- 共変量のバランスをチェックすることが重要

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