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CHAPTERFOURTEEN: VALIDATIONSTUDYOFACUTE MYOCARDIALINFARCTION, RESULTS

This chapter summarizes the key findings of the AMI validation study. The seven research questions are listed below, followed by a detailed description of all relevant findings. Where appropriate, the specific measures and methods used to answer question are described.

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QUESTION1: What proportion of cases included in the 1993 AMI study should have been excluded because a cutemy ocardial infarction was incorrectly reported or incorrectly diagnosed?

Ofthe1,005recordsreceived,atotalof3 1requiredexclusion.Eighteenrecordswitha principaldiagnosisofAMIontheoriginaldischargeabstractswereexcludedbecause that diagnosis was never documented by a physician. Hospital coders apparently misinterpretedtheserecordsandmistakenly assignedanunsubstantiateddiagnosis. Thecorrectedprincipaldiagnosesforthese18casesareshowninTable14.1.Four additional records were excluded when they were found to be post -transfer hospitalizations.Thereasonsfortransferwere:infarct extension,coronaryarterybypass grafting(CABG),recurrentanginarequiringintravenousnitroglycerin,androutinepost -infarctcarecomplicatedbypneumonia.

Finally,nineexcludedrecordswerederivedfromasampleof22caseswithsecondary, notprin cipal,diagnosesofAMI. These cases qualified for the 1993 hospital outcomes study because their principal diagnoses (e.g., ventricular tachycardia, ventricular fibrillation, complete atrioventricular block) were presumed to represent AMI complications. Infour of these nine cases, these condary diagnosis of AMI was never documented by a physician, although the principal diagnoses were correct. One additional case with a principal diagnosis of cardiogenic shock contained a brief reference to "probablem" yocardial infarction "but no supportive documentation; it was also excluded. The last four cases represented postoperative AMI sthatoccurred after revascularization for arterial thrombosis (444.xx), which was in appropriately included in the 1993 list of acceptable principal diagnoses.

Notethatthis validationstudydidnotaddressthenumberofAMlcasesthatshouldhavebeenincluded butweremissedbecauseofunderreportingbyhospitals.

Alloftheremaining974recordscarriedaphysiciandiagnosisofAMlandweretherefore fullyabstracted. However, manyofthese recordsfailed to meet more rigorous criteria for the diagnosis of AMI. The international diagnostic criteria used in the World Health Organization's Monitoring of Trends and Determinants in Cardiovas cular Disease (MONICA) project ² and adapted by the Corpus Christi Heart Project were reviewed, along with the comparable ARIC criteria ³ used in the Cardiovas cular Health Study. These criteria were adapted to the AMI validation data in the following manner:

- 1. Chestpainwasdefinedasasensationof"pain,...tightening,pressure,discomfort, angina,...heaviness, crushing, squeezing, burning..." lo cated in the chest or epigastricareawithorwithoutradiationtothearms,jaw,throat,orneck.Thepain hadtohaveoccurredwithin24hoursofpresentation.
- Positive enzymes were defined as a creatine kinase isoenzyme (CK -MB) at least twice theu pper limit of normal or "positive" (if the exact value was not reported), or a CK-MB greater than or equal to 10% of the total CK.CK -MB sreported to be "weakly positive" were coded as "positive."
- 3. Borderline enzymes were defined as a total CK at least twice the upper limit of normal, atotallactatedehydrogenase (LDH) at least twice the upper limit of normal, atotallactatedehydrogenase (LDH) at least twice the upper limit of normal (if no CK was obtained), a CK MB between 5% and 9% of the total CK, or a CK MB between once and twice the upper limit of normal.
- 4. Normalenzym esweredefinedasaCK,CK -MB,andLDHthatdidnotmeetanyof thecriteriaspecifiedin(2)and(3).
- 5. Patients with positive enzymes and chest pain were automatically classified as "definite" AMIs. Patients with positive enzymes in the absence of chest pain, or borderlineenzymes with chest pain, were classified as possible AMIs.
- 6. Forpatientswithnoenzymes,normalenzymes,orborderlineenzymeswithoutchest pain,thefirstandlastelectrocardiogram(ECG)within24hoursafterpresentation werereviewed:
 - a. IftheECGsshowedan"evolvingdiagnostic"pattern(usingspecificMinnesota codes),thenthecasewasclassifiedasa"definite"AMI.
 - b. If the ECGs showed a "diagnostic" pattern, an "evolving ST -T" pattern, or an equivocal pattern and the enzymes were borderline (in the absence of chest pain), then the case was classified as a "possible" AMI.

² GillumRF,FortmannSP,PrineasRJ,etal.Internationaldiagnosticcriteriaforacutemyocardialinfarction andacutestroke. AmHeartJ 1984;108:150 -158.

³ CohortComponentProcedures,ARICProtocol2,version2.0,1988.

- c. If the ECGs showed a "diagnostic" pattern or an "evolving ST -T" pattern, the enzymes were incomplete, and chest pain was present, then the case was classified as a "possible" AMI.
- d. Allothercombinationsoffindingswereclassifiedas"noAMI."

Usingthisalgorithm,the974abstractedcaseswereclassifiedasshowninTable14.2. The74doubtfulcasesweredistributedacross22hospitals,wi th0to9casesateach hospital. Therewasnodifferenceintheproportionofphysician -diagnosedAMIsthat failed to meet clinical criteria across hospital mortality classes. However, medium -volumehospitalstendedtohaveahigherpercentageofthese doubtfulAMIsthanhigh -volumehospitals(9.3%versus5.9%,p=0.062). DoubtfulAMIshadhighermortalitythan definiteandpossibleAMIs(41.9%versus23.1%,p<0.001),althoughonly9.5%ofthese 74patientspresentedtothehospitalincardiacarrest.

Combiningtheresultsofthese analyses, 31 cases from the original sample of 1,005 (3.1%) were definitely false positives using the inclusion and exclusion criteria from OSHPD's 1993 study of AMI mortality, and an additional 74 cases (7.4%) were suspected to be false positives using modified ARIC criteria. These suspected false positives were discharged with a diagnosis of AMI by alicensed physician, but lacked the necessary combination of chest pain, cardiac enzyme, and ECG findings. Reweighting these figures to the statewide population, OSHPD estimates that 2.2% of the case sincluded in its 1993 AMI mortality study are definitely false positives and an additional 7.2% are suspected false positives.

Toexplorewhethermodificationsoftheselectionrul esmightfurtherreducethefalse positiverate, aspecial analysis was performed of the 22 cases with a reported principal diagnosis other than 410.xx (AMI). These cases were included in the California Hospital Outcomes Project because missequencing of diagnoses was suspected. In other words, AMI was thought to be the underlying reason for a dmission when a patienthada secondary diagnosis of AMI and a principal diagnosis of a known AMI complication, such a scardia carrestor ventricular tachy cardia (see Chapter Three of the 1993 report for a complete list).

Ofthe 22 cases with a secondary diagnosis of AMI, 9 were excluded for the reasons shown in Table 14.1 and 13 (57%) were found to have had a qualifying AMI. The diagnoses for the latter set of pati ents were resequenced with AMI as the principal diagnosis Amongthese 13 cases, 5 had a principal diagnosis of paroxysmal ventricular tachycardia (427.1), 3 had cardiac arrest (427.5), 2 had a cute edema of the lung, unspecified (518.4), 1 had arterialem bolismorth rombosis (444.xx), 1 had hypotension, unspecified (458.9), and 2 had other (785.59) or unspecified (785.50) shock. Sorted by principal diagnosis, the proportion of cases upheld as AMIs ranged from 1 of 5 with arterial embolismorth rombosis and 0 of 2 with complete a trioventricular blockto 3 of 3 with cardiac arrest and 5 of 6 with ventricular tachycardia. In the current report, arterial embolismand thrombosis was removed from the list of acceptable principal diagnoses in

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HsiaDC.AccuracyofMedicarereimbursementforcardiacarrest. JAMA1990;264:59 -62.

Table 3.1. This change would be expected to reduce the statewide (weighted) frequencyofdefinitefalsepositivesto 2.1%.

QUESTION2:Whatisthestatewidereportingaccuracyforimportantriskfactors includedintherisk -adjustmentmodels?

Table 14.3 summarizes the ac curacy of reporting for ICD -9-CM coded risk factors derived from original hospital discharge abstracts, based on a comparison with the same risk factors derived from CMRI's reabstraction of matched records in the validation data set. It includes all of the eclinical risk factors that were in AMIM odel B for cases with no prior admissions. At the bottom of the table, over all measures of coding accuracy for diabetes and hypertension are reported (the serisk factors were categorized in multiple levels for risk modelling, which makes dichotomous measures of coding accuracy hard to interpret). Exceptas indicated, all numbers in this table are weighted to adjust for the oversampling of outlier hospitals and deaths.

Sensitivityandpredictivevaluearemeasure sof validitythatpresumetheexistenceofa "goldstandard." CMRI's reabstracted diagnoses are taken to represent the truth; the diagnoses reported to OSHPD are evaluated against this gold standard. Sensitivity equalsthepercentageofpatientswitha riskfactor,accordingtoCMRI'sreabstracted ICD-9-CMcodes, who were reported to have that risk factor on the abstractoriginally submittedtoOSHPD.Asensitivityof30%meansthatthehospitalcodedonly30%of the cases with that risk factor, when co mparedtothe CMRI goldstandard. Positive predictivevalue(PV+)equalsthepercentageofpatientsreportedtohaveariskfactoron theoriginalabstractwhowereconfirmedbyCMRI'sreabstractionAPV+of30%means tohavetheriskfactorshouldhavebeenreported, thatonly30%ofthecasesreported whencomparedtotheCMRIgoldstandard.TheidealvalueforbothsensitivityandPV+ is100%.LowsensitivityrepresentsundercodingandlowPV+representsovercoding.

Table14.3alsoreportsthespec ificity,negativepredictivevalue,andlikelihoodratiosfor eachriskfactor. These statistics are less useful than the sensitivity and PV+, but are offered for the sake of completeness. Specificity is the percentage of patients without a risk factor, according to CMRI's reabstract, who were accurately identified from the original abstract. Negative predictive value (PV -) is the percentage of patients reported not to have a risk factor on the original abstract who were confirmed through CMRI's reabstraction as not having the risk factor. The ideal value for both specificity and PV - is 100%. The likelihood ratio (LR+) equals the sensitivity divided by 1 - specificity; this measure incorporates both sensitivity (coding accuracy among patients with the risk factor) and specificity (coding accuracy among patients without the risk factor) into a single number. Higher values are better, while a value of 1.0 represents "random" coding.

Finally, the kappastatistic is a measure of reliability. It does not pre sumethe existence of a "gold standard." Instead, it is designed to assess agreement between two independent datasources when neither is clearly superior to the other. Kappais equal to the proportion of records for which both data sources agree on the presence or

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absence of a risk factor, corrected for the level of agreement expected by chance.

Values fall between 0 and 1, where 1.0 is ideal. If a risk factor has a prevalence of 1% in two datas ources, the sed at a source swould be expected to a gree in (0.99 *0.99) + (0.01 * 0.01) = 98.02% of cases. Hence, 98% a greement is no better than one would expect by chance. Because the kappa statistic removes this "chance" effect, it is quite conservative. Note that the kappa statistic does not indicate whether there is undercoding or overcoding, because it presumes that both data sources are equally valid.

Thevalidityandreliabilityofcodingrangefromexcellenttopoor,dependingontherisk factor. These val ues are excellent (sensitivity>80% and $\kappa>0.8$) for infarct site and diabetes, althoughabout60% of patients reported to have "other or unspecified" site actually have documentation suggesting aspecific site. The quality of coding is very good (sensitivit y>60% and $\kappa>0.6$) for congestive heart failure (CHF), chronic renal disease, prior coronary by pass surgery, history of pacemaker, complete a trioventricular block, and shock. Several other risk factors, including epilepsy, other cerebrovas cular disease, pri mary and secondary malignancy, and hypertension, have intermediate sensitivities with kappa statistics between 0.45 and 0.60. Six risk factors (chronic liver disease, late effects of cerebrovas cular disease, hypotension, pulmonary edema, nutritional deficiency, and other valved is ease, are poorly coded, with sensitivities under 40% and kappa statistics less than 0.45. The least reliably coded risk factor, hypotension, has alikelihood ratio of 7.4. This means that patients with hypotension are 7.4 times more likely to have that diagnosis reported than patients without hypotension.

QUESTION3: Are important risk factors coded more thoroughly athospitals with low risk -adjusted mortality than athospitals with high risk -adjusted mortality? If so, does the variation in risk -adjusted mortality diminish when inter -hospital differences in risk factor coding are removed?

Table 14.4 shows the sensitivities and kappa statistics for all risk factors present in at least 5% of cases (n=49) according to CMRI srea bstracts, stratified by hospital mortality and hospital volume. The "n" next to each risk factor name represents the number of cases with that risk factor, according to CMRI sreabstracts. Diabetes and hypertension include both complicated and uncomplicated according to ted cases; these results are similar to those based on separate subcategories. Probability values are based on Fisher's 2 -tailed exact test. A relatively high pvalue cut off (p<0.10) is recommended because of the small sample size and exploratory nature of the study.

This analysis shows no consistent differences in risk factor coding across hospital mortality and volume categories. Hospitals with high risk -adjusted mortality code anteriorwallsitewithgreatersensitivity(p=0.030)thanhospitalswith loworintermediate risk-adjusted mortality, but hospitals with intermediate mortality code chronic renal

LandisJR,KochGG.Themeasurementofobserveragr eementforcategoricaldata. Biometrics1977; 33:159-174.

diseasewiththehighestsensitivity(p=0.044) and reliability(p=0.043). A crosshospital volume categories, high -volume hospitals code other valve d isease more reliably (p=0.031), and hypertension(p=0.045) and shock(p=0.056) less reliably, than medium volume hospitals. Only for CHF is a significant tradeoff between sensitivity (undercoding) and positive predictive value (overcoding) seen: high -volume hospitals code CHF with lower sensitivity (p=0.039) but higher predictive value (p=0.015) than medium-volume hospitals.

Overall,65.0%oftheoriginaldischargeabstractshaveatleastonemissingclinicalrisk factorand30.9%haveatleastwomissin griskfactors.Therearenodifferencesacross hospital mortality categories in the percentage of original discharge abstracts with missing clinical risk factors, but this occurrence is more frequent at high -volume hospitals than at medium -volume hospitals (68.8% versus 61.2%, p=0.015). At the hospitallevel, the percentage varies from 45% to 87%.

Conversely, 31.5% of the original discharge abstracts have at least one unsupported clinical risk factor based on CMRI's reabstraction. Coding of unsupported clinical risk factors is more frequentation -mortality hospitals than at intermediate or high -mortality hospitals (36.7% versus 29.2% and 29.0%, p=0.039), but is unrelated to hospital volume. At the hospital level, the percentage varies from 10% at a high-mortality hospital to 74% at a low -mortality hospital.

The aggregate impact of undercoding and overcoding risk factors was evaluated by recalculating risk -adjusted hospital mortality rates, using only risk factors that were identified through CMRI's reabstraction of the records in the validation dataset. Eight models (e.g., four versions of Model A and four versions of Model B) were used in this analysis:

- 1a,1b The1993risk -adjustmentmodelsforAMImortality(amongcaseswithnoprior admissions) were applied to the validation sample, thereby estimating each patient's risk of death using the ICD -9-CM codes reported to OSHPD and the coefficients listed in the 1993 report of the California Hospital Outcomes Project. For patients with no prior admissi ons, these estimates equal the predicted probabilities reported to hospitals in 1993.
- 2a,2b Thesamerisk -adjustmentmodelswereappliedtothesamecases,butthelCD 9-CMcodesreportedtoOSHPDwerereplacedbythosereabstractedbyCMRI. This proced ure generated the predicted probabilities that would have been estimated if an individual hospital had improved its coding practices to match CMRI's standard (assuming that the hospital had too few cases to significantly affect the coefficient estimates in the risk -adjustment model). AlthoughCMRI'sreabstractsmaynotrepresentatruegoldstandard,theyareat leastcodeduniformlyacrosshospitalcategories.
- 3a,3b The1993risk -adjustmentmodelsforAMImortality(amongcaseswithnoprior admissions) were reestimated on the validation sample, using the ICD -9-CM codes reported to OSHPD. Sampling variation explains the difference sbetween

- these models and the comparable models reported in the 1993 report of the California Hospital Outcomes Project.
- 4a,4b Thesamerisk -adjustmentmodelswerereestimatedonthesamecases,butthe ICD-9-CMcodesreportedtoOSHPDwerereplacedbythosereabstractedby CMRI. This procedure generated the predicted probabilities that would have been estimated if all hospita Ishad coded their records as CMRI did . By comparing these models with models 3a and 3b, one can evaluate whether reabstracted diagnoses yield amore powerful, less biased model than diagnoses reported to OSHPD.

Insummary,models1a,1b,2a,and2bused thesameregressioncoefficientsthatwere originallyestimatedin1993usingstatewidedata. Thesecoefficients are quitereliable because they are based on nearly 29,000 AMIs, but they are biased by measurement error because hospitals do not reportris kfactors in a uniform manner. Models 3 a and 4 awere estimated on the 974 cases in the validation sample; models 3 b and 4 b were estimated on the 938 cases with non missing values of all risk factors. The regression coefficients in the semodels are unrelia ble because of the small sample size, but they are also less biased because risk factors are presumably measured more accurately All four models were weighted to compensate for the oversampling of both deaths and cases from extreme - outcome hospitals.

Table 14.5 shows the results of all eight models, comparing risk -adjusted mortality across hospital outcome and volume strata. In this analysis, each stratum should be regarded as a single facility whose patients were drawn randomly from all hospitals in that stratum. The ISR represents the indirectly standardized mortality ratio comparing that subset of hospitals to the statewide experience (e.g., the number of observed deaths divided by the number of expected deaths). An ISR greater than one indicates higher than expected mortality, whereas an ISR less than one indicates lower than expected mortality. Asterisk sedenote the ISR sthat significantly differ from one, at a 95% confidence level (p<0.05).

UsingOSHPDdataand1993OSHPDModelBcoefficients(mo del1b), "better" hospitals had30%fewerdeathsthanexpected(1 -0.7007)while"worse"hospitalshad47%more deathsthanexpected(1.4701 -1). Bothofthese ISRs differ significantly from one, as would be expected because hospitals were sampled based on their Model Brisk adjustedoutcomeclassificationfromOSHPD's1993report. Usingreabstracteddata (model2b).thesamesetof"better"hospitalshad37%fewerdeathsthanexpectedand thesamesetof"worse"hospitalshad18%moredeathsthanexpected. Theconfidence limitsfor"better"hospitalsstilldonotincludeone,but"worse "hospitalsnolongerdiffer significantly from expected. Another way to interpret these numbers is that the differenceinrisk -adjustedmortalitybetween"better"and"wor se"hospitalsdecreasesby 28%ifCMRIdataareusedinplaceofOSHPDdata((0.7694 -0.5519)/0.7694).

Thisdecreasewas **not**foundwhenModelAwasappliedinsteadofModelB.Using OSHPDdataand1993OSHPDModelAcoefficients(model1a),"better"hosp italshad 26%fewerdeathsthanexpectedwhile"worse"hospitalshad45%moredeathsthan expected.Usingreabstracteddata(model2a),thesamesetof"better"hospitalshad

17% fewer deaths than expected but the same set of "worse" hospitals had 55% mor deaths than expected. Hence, the difference in risk and "worse" hospitals is virtually unchanged when CMRI data are used in place of OSHPD data (0.7208 versus 0.7052).

ThelowerhalfofTable14.5showstheresultso fmodelsreestimatedonthevalidation sample.Asdescribedabove,reestimationremovesbiasintheregressioncoefficients butalsodecreasestheirreliability.Becauseallfourpatientswithchronicliverdisease, as reported to OSHPD, died before disc harge, the risk -adjustment model based on OSHPDriskfactorsinitiallycouldnotbereestimated.Whenchronicliverdiseasewas omitted,modelfitsignificantlydeteriorated(c=0.835versus0.840).Thisproblemwas corrected by recoding the value of chronic liver disease for one case, which was randomlyselectedfromthethreecasesthatwereidentifiedbyCMRI,butnotreportedto OSHPD,ashavingthiscondition.

UsingOSHPDdataandreestimatedModelBcoefficients(model3b),"better"hospitals had39 %fewerdeathsthanexpectedwhile"worse"hospitalshad19%moredeathsthan expected. Only the ISR at "better" hospitals differs significantly from one, which is consistentwithamodest((0.7694 -0.5849)/0.7694=24%)decreaseintherisk -adjusted mortality difference between "better" and "worse" hospitals, relative to that obtained usingthe1993regressionmodel.Thisdecreaseisattributabletorandomerrorinthe Using risk-adjustmentmodelsandaphenomenonknownas"regressiontothemean." reabstracteddata(model4b),thesamesetof"better"hospitalshad37%fewerdeaths than expected and the same set of "worse" hospitals had 8% more deaths than expected.Hence.thedifferenceinrisk -adjustedmortalitybetween"better"and"worse" hospitals decreases by 24% if CMRI data are used in place of OSHPD data. Using reestimatedModelAcoefficients(models3aand4a),thisdifferencedecreasesbyonly 12%whenCMRIdataareusedinplaceofOSHPDdata.

These results suggest that unreliable coding (represented by the difference between original OSHPD data and reabstracted CMRI data) explains 24% to 28% of the difference in risk -adjusted mortality between low -mortality and high -mortality outlier hospitals based on Model B, but on 12% of the difference based on Model A. In other words, Model B is somewhat compromised by coding bias but Model A is not. Even with Model B, however, at least 72% of the gap in risk -adjusted mortality persists when coding variation is eliminated.

Severalotherfi ndingsfromthesemodelsareofinterest First, the predicted probabilities of death calculated from OSHPD data are highly correlated with those calculated from reabstracted data at the individual level, regardless whether the 1993 regression models are applied (Spearman r=0.93 with Model A, r=0.91 with Model B) or reestimated

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Regressiontothemean describestheobservationthat when outlier cases (e.g., hospitals) are selected using one measurement (e.g., risk -adjustment model) with a specific statistical threshold (e.g., p<0.05), other measurements on the same cases tend to demonstrateless extremere sults.</p>

(Spearman r=0.86 with Model A, r=0.83 with Model B). The correlations are slightly weaker with Model B because undercoding has more impact on the predictions generated by that mode I. Figures 14.1 and 14.2 show how reabstracting ICD -9-CM codes affects the ISRs of individual hospitals, based on either Model A or Model B. This is shown for illustrative purposes only, as too few patients were sampled to generate statistically signific antresult satthehospital level. However, the effects were generally consistent among the hospital sine ach stratum.

Second, the effect of using reabstracted data on model performance can be described (Table14.6).TheoriginalModelsAandBdevelope din1993hadcstatisticsof0.766 and 0.844, respectively. Reestimating both models on the validation sample using OSHPD risk factors generated similar c statistics of 0.782 and 0.841, respectively. Reestimatingbothmodelsusingreabstractedriskfact orsgeneratedhighercstatisticsof 0.814 and 0.881, respectively. The risk -adjustment models estimated using reabstracteddatahavesignificantlygreaterdiscriminationthanthoseestimatedusing originalOSHPDdata. ⁷Unfortunately,modelsthatexplain moredevianceatthepatient leveldonotnecessarily explain more variation at the hospital level. Alog -loglinear regressionanalysisoftherelationshipbetweenobservedandexpectedweighteddeaths acrossthe30participatinghospitalsrevealedthat usingreabstracteddataincreasesthe explanatory power of Model B (from partial R 2 =0.190 to 0.489) but paradoxically 2 =0.425to0.327). decreasestheexplanatorypowerofModelA(frompartialR

Finally, it is instructive to evaluate how specific regress ioncoefficientschangewhen reabstracteddataareusedinsteadoforiginaldata. Ifreabstracteddatarepresentagold standard, then the regression coefficients based on those data approximate the true valuesandanysignificantchangesreflectcodingb ias.Thefirststepinthisanalysisis toexaminethereestimatedmodels(3aand3b)basedonOSHPDriskfactors,shownin Tables 14.10 and 14.11. The coefficients from the semodels are generally similar to thosepublishedinOSHPD's1993report.Thed ifferencesreflectsamplingvariation;in otherwords, the 974 cases randomly sampled for the validation study differsome what fromthe30,958casesintheoriginalsample.Forexample,CHFwasamajorriskfactor the validation sample. Conversely, in the 1993 models, but had no effect in uncomplicated diabetes had no effect in the 1993 models, but was an important risk factorinthevalidationsample.

Comparing models 3a and 4a (available upon request from OSHPD), complicated diabetes, chronicrenaldis ease, otherinfarctsite, and CHF -ageinteractions were more important risk factors with greater coefficients when reabstracted data were used in stead of original data. Conversely, chronic liver disease and other valve disease were less important risk factors. Comparing models 3b and 4b, the same differences were confirmed but others were identified. Epilepsy and pulmonary edema became less important risk factors using reabstracted data, presumably because CMRI identified patients with milder forms of the eserisk factors. Other cerebrovas cular disease became irrelevant because of better coding of "late effects" of cerebrovas cular accidents (CVA).

14-9

DeLongER,DeLongDM,Clarke -PearsonDL.Comparingtheareasundertwoormorecorrelatedreceiver operatingcharacteristiccurves:Anonparametricapproach. Biometrics1988;44:837 -845.

Conversely, hypotension became a more important risk factor. There were striking changes in the coefficients for demographic factors, such as race and payments ource, possibly due to better adjustment for patients' clinical risk factors.

QUESTION4: How often do the clinical characteristics used as risk factors in ModelBactually represent conditions that developed after admission?

Table 14.7 shows the timing of each ICD -9-CM coded risk factor, including the first recordeddate (relative to the date of emergency room arrival or admission, which ever came first) and whether the condition was documented in an emergency room (ER) or admission note. The results of the set wo approaches a regenerally compatible, but may differ be cause: (1) many patients are admitted one day after the irpresentation to the ER; (2) an admission note may be written one or more days after the date of a dmission; and (3) conditions may develop on the day of presentation (or the following day) which would not be documented in ER and admission notes. The second approach is more compatible with the new reporting mandate, which requires Cali forniahos pital store port, beginning with discharges on 1/1/96, whether each condition was "present at admission."

All974validcasesareincluded;thetotalnumberofreportedpatientswithariskfactor (inthesecondcolumn)isbasedonCMRI'sreabst racts.Theriskfactorsfallintothree groups:

1. Conditionsthatareusuallydiagnosedafteradmission

Conditionsinthisgrouparedocumentedin ERoradmission notes in less than 50% of cases and are first diagnosed at least one day after presentatio ninmore than 50% of cases. Examples include hypotension, other cerebrovascular disease, pulmonary edema, and other valve disease. Shock does not quite meet these criteria, with 46% of occurrences documented in ERoradmission notes and 51% first diagnosed on the day of presentation. All of these risk factors except other valve disease are currently considered Model Bvariables. Other valve disease is an exception because it is often based on the findings of diagnostic tests, such as echocardiography or ventriculography, that are performed after admission. The underlying valve disease was almost certainly present at admission, even if the diagnosis was notestablished until several day slater.

2. Conditionsthatareusuallypresentatadmissionbutma ybediagnosedlater Conditions in this group are documented in ER or admission notes in 50 -80% of cases. Examples include congestive heart failure, chronic liver disease, complete atrioventricular block, epilepsy, secondary malignant neoplasm, nutritiona I deficiency, and skinulcer. Complete atrioventricular block, epilepsy, and skinulcer are currently considered Model Bvariables. Nutritional deficiency is sorare that it did not appear in the updated risk -adjust ment models (Chapter Nine of this volum e); secondary and primary malignant neoplasm swere aggregated. The apparent timing of chronic liver disease presumably reflects delayed diagnosis of a preexisting problem.

3. Conditionsthatusuallyrepresentpreexistingriskfactors

Conditionsinthisg rouparedocumentedin ER oradmission notes in at least 80% of cases. Examples include in farct site, chronic renal disease, diabetes, hypertension, late effects of CVA, prior CABG, primary malignant neoplasm, and history of pacemaker. Note that a signif icant proportion of the sepre existing conditions are first noted on the day after presentation. All of these variables are currently used in Model A.

These findings confirm that the additional variables in Model B may be preexisting conditions. Using currently available OSHPD data, it is impossible to distinguish such cases from those that arose after admission. However, a subset of Model B variables that are especially likely to have been diagnosed after admission has been identified (e.g., group 1 abo ve).

The impact of mislabeling conditions diagnosed after admission as risk factors was evaluated by recalculating expected and risk -adjusted hospital mortality rates, using only risk factors that were documented in the ER or admission notes. Four model swere employed:

- 5a,5b The1993risk -adjustmentmodelsforAMImortality(amongcaseswithnoprior admissions)wereappliedtothevalidationsample,keepingthecoefficientsfixed butusingthelCD -9-CMcodesreabstractedbyCMRI.Thesemodelsdifferf rom models2aand2binthatonlyconditionsdocumentedintheERoradmission noteswereusedtocoderiskfactors.
- 6a,6b The same risk -adjustment models for AMI mortality were reestimated on the validation sample, using the ICD -9-CM codes reabstracted by CMRI. These models differ from models 4 a and 4 b in that only conditions documented in the ER or admission notes were used to code risk factors.

Table 14.8 demonstrates the impact of mislabeling as risk factors those conditions diagnosedafteradmission ,acrosshospitalmortalityandvolumecategories. Using the regression coefficients published in 1993 (models 5 a and 5 b), disregarding conditions diagnosed after admission has little effect on expected mortality rates from Model A and therefore little effect on the ISRs. It has a dramatic effect on expected mortality rates from Model B, which drop from 15.6 % to 12.3 % at low -mortality hospitals, from 14.6 % to 11.2 % at intermediate hospitals, and from 14.6 % to 10.7 % at high -mortality hospitals. As a result, disregarding conditions diagnosed after a dmission increases the difference in risk-adjusted mortality between low and high -mortality hospitals by 49% ((0.8232 - 0.5519)/0.5519).

Afterhospitalsbeginsubmittingdatawith"presentatadmission"indicators ,effectivewith dischargesonJanuary1,1996,OSHPDwillbeabletoestimatemodelsthatadjustonly forpreexistingconditionsanddisregardconditionsdiagnosedafteradmission.Models 6aand6binTables14.10and14.11demonstratethepotentialimpa ctofthischange. DisregardingconditionsdiagnosedafteradmissionwhencodingModelAriskfactors increasestheimportanceofCHF(fromanoddsratio(OR)of0.85to1.42)andweakens

theinteractionsbetweenCHFandinfarctsite,buthaslittleeffec tonotherassociations. DisregardingtheseconditionswhendefiningModelBriskfactorshasasimilareffecton CHF, but also increases the importance of complete atrioventricular block (from OR=1.12to2.14)anddecreasestheimportanceofcomplicated diabetes(fromOR=4.15 to1.74),hypotension(fromOR=1.64to1.00),pulmonaryedema(fromOR=1.59to1.13), andshock(fromOR=22.6to10.0).Thesedifferencesexemplifythesignificantbiasin AMIModelBduetoOSHPD'sinabilitytodistinguishconditio nspresentatadmission fromthosediagnosedlater.

Adjustingonlyforpre -existingconditionscompromisesthediscriminatorypowerofModel B(fromc=0.879to0.815)morethanthatofModelA(fromc=0.814to0.786). At the hospital level, adjusting only for pre -existing conditions similarly compromises the explanatorypowerofbothModelA(frompartialR 2 =0.394to0.287)andModelB(from partialR 2 =0.470to0.310)inlog -loglinearregressions. Using either approach, ModelB remains more powerfulthan ModelA even when conditions diagnosed after admission are disregarded.

Using these models with reestimated coefficients, the expected mortality rate at low mortality hospitals increases (from 15.6% to 17.4% in model 6a; from 15.6% to 16.8% in Model 6b) whereas that at high -mortality hospitals decreases slightly (from 15.9% to 15.2% in Model 6a; from 16.1% to 15.4% in model 6b) when conditions diagnosed after admissionared is regarded. As a result, removing the bias due to mislabeling of these conditions increases the difference in risk -adjusted mortality between low and high mortality hospitals by 25% in Model A and by 20% in Model B. In otherwords, counting conditions diagnosed after admission as risk factors leads one to underestimate the true difference in risk -adjusted mortality, even when the regression coefficients are reestimated.

QUESTION5:Howdotherisk -adjustmentmodelschangewhenadditionalclinical variablesareusedasriskfactors?

As part of CMRI's review of records in the AMI va lidation study, many clinical data elements were abstracted. This process involved reviewing all components of the medicalrecord, including emergency roomnotes, histories and physical examinations, laboratory results, radiology reports, echocardiography reports, and operative notes. Basedon review of the clinical literature and discussions with the AMIC linical Advisory Panel, the following variables (with the alternative specifications listed) were evaluated as potential risk factors for in -hospital death within 30 days after an AMI.

1. Historicalfindings:

- a. PriorAMI,numberofpriorAMIs,priorAMIwithin6months
- b. Peripheralvasculardisease(PVD),PVDwithpriorrevascularization
- c. Priorstroke, priorstroke within 12 months, prior transient is chemicattack
- d. PriorCABG,priorpercutaneoustransluminalcoronaryangioplasty(PTCA),prior CABGorPTCA
- e. Asthma
- f. Chronicobstructivepulmonarydisease

- g. Knownorsuspectedaorticaneurysm
- h. Cardiacarrest, cardiopulmonary resuscitation, defibr illation within 24 hours
- i. Suddendeath,cardiacarrest(ever)
- j. Atrialfibrillationorflutter
- k. Congestiveheartfailure(CHF)
- I. Currentsmoker, eversmoker
- m. Pericarditis
- n. Cocaineuse
- o. Permanentpacemakerorautomaticdefibrillatorinplace
- p. Durationofchestpain, absence of chestpain

2. Physicalfindingsatpresentation:

- a. Systolicheartmurmur(any,gradelllorlouder)
- b. Pulmonaryrales(any,morethanhalfwayup)
- c. Heartrate
- d. Systolicbloodpressureatpresentation, diastolicblo odpressure
- e. Respiratoryrate
- f. Shock,cerebralhypoperfusion,peripheralcyanosis,MilitaryAntishock(MAST) trousersorpressorstosupportbloodpressure
- g. Bilateralperipheralorpresacraledema
- h. S₃orsummationgallop

3. Laboratoryvaluesatpr esentation:

- a. FirstCKvalue,firstCKvalueindexedtoupperlimitofnormal
- b. Hematocrit.anemia
- c. Serumcreatinine,bloodureanitrogen(BUN)
- d. Plateletcount, thrombocytopenia

4. Radiographicfindingsatpresentation:

- a. CHF,cardiomegaly,pulmon aryedema,pulmonaryvascularcongestion
- b. Pleuraleffusion(unilateral,bilateral)
- c. Pulmonaryinfiltrate(unilateral,bilateral)

5. Electrocardiographic finding sat presentation:

- a. Atrialfibrillationorflutter
- b. QRSwidening(e.g.,bundlebranchb lock)
- c. Ventricularhypertrophy(left,right,biventricular)
- d. Leftaxisdeviation, rightaxisdeviation

6. Miscellaneous:

- a. Donotresuscitate(DNR)order, DNR onday of admission, DNR at admission
- b. Leftventricularejectionorshorteningfraction
- c. Non-AMIbyARICcriteria

Bivariatechisquaretableswereusedtodeterminethespecificationofeachclinicalrisk factorthatbestdiscriminatesbetweenlow -riskandhigh -riskpatients. This specification was then tested in weighted and unweighted mu litivariate logistic regression models that included all Model Brisk factors, based on original OSHPD data. Each potential clinical

riskfactorwastestedindividuallyinthesemodels.Riskfactorsthatsignificantly(p<0.10) increased the discrimination (cstatistic) of the weighted models or had a significant (p<0.10). Waldchisquare statistic in the unweighted models were retained for further analysis. 8

This procedure identified seven promising clinical risk factors, which were forced simultaneously i nto an expanded version of Model B (with risk factors coded from originalOSHPDdata). Alloftheothercandidaterisk factors were then tested one final time, using automatic variable selection procedures (backward elimination, forward selection, and ste pwise selection with pto enter and exit of 0.10) on the unweighted validations ample. This final step was designed to ensure that no potentially useful risk factors were discarded. A total of nine clinical risk factors were identified. At the recommendation of the AMIC linical Advisory Panel, the senine risk factors were divided into five "core" variables and four "secondary" variables. The secondary variables either had marginal statistical (e.g., 0.03 < p < 0.10) or clinical significance (e.g., systolic murmur), or became in significant when reabstracted ICD -9-CM codes were used in stead of original OSHPD data (e.g., history of stroke).

Thedefinitionsofthecoreclinical variables are as follows:

- Systolic blood pressure at presentation (in mm Hg). F or statistical reasons, this
 variable was recoded to zero if a patient had a cardia correspiratory arrest within 24
 hours before presentation.
- 2. Heartrateatpresentation(inbeatsperminute). Forstatistical reasons, this variable was recoded to zero if a patient had a cardia correspiratory arrest within 24 hours before presentation.
- 3. Cardiacarrestwithin24hoursbeforepresentation. This variable was recoded to one if a patient had a heartrate or blood pressure equal to zero at presentation. All

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Bothofthesecriteriaofferuniqueadvantagesanddisadvantages.Usingunweightedmodelstoevaluate theimportanceofclinicalriskfactorsallowstheresearchertoapplyvariableselectionrulesbasedon traditional chi square and likelihood ratio statistics, which are particularly appropriate with maximum likelihoodestimationtechniques(e.g.,logisticregression).However,theparameterestimatesandodds ratiosfromunweightedmodelsarebiasedbecauseoftheoversamplingofpatientswhodiedandpatients atselectedhospitals.Usingweightedmodelstoevaluateclinicalriskfactorsobviatesthisproblem,but thepvalues,chisquarestatistics,andstandarderrorsfromsuchmodelsareuninterpretablebecausethey dependenthescaleoftheweightingfactor(e.g.,multiplyingallweightsby10changesthepvalues).A solutiontothisproblemistousechangesinthereceiveroperatingcharacteristiccurvetoidentifyrisk factorsthatsignificantlyimprovediscrimination.

- discrepancies between this variable and related variables on pre -hospital cardiopulmonaryresuscitationanddefibrillationwerereconciled.
- 4. Shockatpresentation(basedontheuseofMASTtrousersorpressors,orclinical evidenceofbothcerebr alandperipheralhypoperfusion).
- 5. Do-not-resuscitateorderonthedayofpresentationtotheemergencyroomorthe dayofadmission.

The definitions of these condary clinical variables are as follows:

- 6. The ratio of the first CK value (within 24 hou rs of presentation) to the hospital's gender-specificupper limit of normal.
- 7. Pulmonaryrales(regardlessofextent)onthepatient'sfirstphysicalexamination.
- 8. Systolicheartmurmur, gradelllorlouderonthepatient's first physical examination.
- 9. Anypriorhistoryofstroke.

This final set of core and secondary clinical risk factors was then added to multivariate risk-adjustment models in the following manner:

- 7a,7b Models3aand3b,whichreestimatedOSHPD's1993risk -adjustmentmodels (forAMIcaseswithnoprioradmissions)onthevalidationsampleusingthe ICD-9-CMcodes reported to OSHPD, were augmented with the five core clinical risk factors.
- 8a,8b Models7aand7bwerefurtheraugmentedwiththefoursecondaryclinical riskfactor s.
- 9a,9b Models4aand4b,whichreestimatedOSHPD's1993risk -adjustmentmodels onthevalidationsampleusingthelCD -9-CMcodesreabstractedbyCMRI, wereaugmentedwiththefivecoreclinicalriskfactors.
- 10a,10b Models9aand9bwerefurtheraug mentedwiththefoursecondaryclinical riskfactors.
- 11a,11b Models6aand6b,whichreestimatedOSHPD's1993risk -adjustmentmodels onthevalidationsampleusingonlyriskfactorsthatweredocumentedinthe ERoradmissionnotes(accordingtoCMRI'sr eabstractedICD -9-CMcodes), wereaugmentedwiththefivecoreclinicalriskfactors.
- 12a,12b Models11aand11bwerefurtheraugmentedwiththefoursecondaryclinical riskfactors.

Table14.9showshowaddingclinicalriskfactorsaffectstheperforma ncecharacteristics ofbothModelAandModelB,dependingwhetherICD -9-CMcodedriskfactorswere

basedonoriginaldataorreabstracteddata. The statistics given formodels 3a, 3b, 4a, and 4b differs lightly from those listed in Table 14.6, because missing values for race and the first CK limited the sample size to 925 for all of the models shown in Table 14.9.

Addingclinicalriskfactorsclearlyimprovesmodeldiscrimination,regardlesswhetherthe "base"modelusesoriginallyreportedICD -9-CMcode s,reabstractedICD -9-CMcodes, or reabstracted codes from the ER or admission records. Not surprisingly, the magnitudeofthisimprovementissmallerforModelB(e.g.,fromc=0.879toc=0.898with reabstractedcodes) than for ModelA(e.g.,fromc=0.814 to 0.864 with reabstracted codes). The core clinical variables contribute much more than the secondary clinical variables, although the latter factors further improve the discrimination of most models. Although the magnitude of improvement from adding or eclinical variables appears to be smaller when reabstracted ICD -9-CM codes are used in the "base" model in stead of original codes, limiting the analysis to codes reabstracted from the ER or admission notes actually increases the magnitude of improvement.

Atthehospitallevel,addingclinicalriskfactorsalsoimprovestheexplanatorypowerof risk-adjustmentmodels. For example, coreclinical riskfactors improve the proportion of variance in observed weighted deaths attributable to the risk -adjustment model from 0.394 to 0.494 with Model A (using reabstracted codes) and from 0.470 to 0.539 with Model B (using reabstracted codes). Secondary clinical variables provide little or no incremental benefit. Similar improvements are noted when coreclinical variables are added to models that adjust only for risk factors reabstracted from ER or admission notes.

Tables14.10and14.11showtheparameterestimatesandoddsratiosforeachofthe additionalclinicalriskfactorsinModelsAandB,respectively. Giventhebestpossible basemodel(e.g.,ICD -9-CMcodesreabstractedfromERoradmissionnotes),ado -notresuscitateorderincreasestheoddsofdeath8.3(ModelA)or9.9(ModelB)times.A recent cardiopulmonary arrest increases the odds of death 14 .5 (Model A) or 19.6 (ModelB)times. Theoddsofdeathincrease 1.16 (ModelA) or 1.17 (ModelB) times witheach10beatperminuteincreaseintheheartrate, and increase 1.14 times with each10mmHgdecreaseinthesystolicbloodpressure.Shockat admissionincreases the odds of death 4.2 (Model A) or 3.3 (Model B) times. The odds ratios for the secondaryclinicalvariablesaregenerallysmaller:1.4or1.5forrales;1.5or2.5fora loudsystolicmurmur; 1.1 or 1.2 for a history of stroke; and 1. 1foreachmultipleofthe upper limit of normal in the first CK. All of these values are relatively stable across models, except that a history of stroke is significant only when original ICD -9-CMcodes wereusedtodefine"lateeffects"ofaCVA.Confi denceintervalsfortheseoddsratios canbecalculatedusingstandarderrorsavailableuponrequestfromOSHPD.

Tables14.10and14.11alsoshowhowaddingclinicalriskfactorsaffectstheparameter estimates and odds ratios for the risk factors that we reincluded in the 1993 models. Most of the sevalues change relatively little, which indicates that the odds ratios are not confounded by clinical risk factors. The major exceptions are as follows:

1. The unfavorable effect of CHF, which was only seen u sing ICD -9-CM codes reabstracted from ER or admission notes, disappears after adjustment for core

- clinical risk factors (e.g., vital signs, cardiopulmonary arrest, DNR). The latter variables represent the clinical manifestations of poor cardiac function.
- 2. Theunfavorableeffectofotherinfarctsitemarkedlydiminishesafteradjustmentfor coreclinicalriskfactors.
- 3. Theprotectiveeffectofbeingblackandtheunfavorableeffectofbeinguninsured, accordingtomodelB, essentially disappearafter adjustment for core clinical risk factors. In other words, the apparent racial and socioeconomic effects are largely explained by clinical differences.
- 4. SeveraloftheadditionalriskfactorsbasedonICD -9-CMcodesinModelBlosetheir unfavorableef fectorevenbecomeprotectiveafteradjustmentforcoreclinicalrisk factors.Forexample,hypotensionbecomesirrelevantwhenthemodeladjustsfor theactualvalueofthesystolicbloodpressure.

QUESTION 6: Do hospitals with significantly higher or lower than expected mortalityappearclosertoaverageafteradjustingforadditionalclinicalvariables? How do the risk -adjusted mortality rates and p values for individual hospitals change when additional clinical variables are used as risk factors?

Table14.12demonstratestheimpactofaddingclinicalriskfactorstotherisk -adjustment modelsbasedonICD -9-CMdata,acrosshospitalmortalityandvolumecategories.In general, neither core clinical variables nor secondary clinical variables systema tically change expected mortality rates for these groups of hospitals. Starting with a reestimatedversionofModelA.basedonthelCD -9-CMcodesreportedtoOSHPD.the additionofbothcoreandsecondaryclinicalriskfactorsreducesthedifferenceinr iskadjustedmortalitybetweenlow -mortalityandhigh -mortalityhospitalsby10%((0.5273 0.4743)/0.5273) Startingwithasimilarlyreestimated version of ModelB, the addition of trast,theaddition bothsetsofclinicalriskfactorsreducesthisdifferenceby20%.Bycon ofclinicalriskfactorstoareestimatedversionofModelAbasedonreabstractedICD CMdatahasaminimaleffectonthedifferenceinrisk -adjustedmortalitybetweenlow mortalityandhigh -mortalityhospitals.Theadditionofclin icalriskfactorstoasimilarly reestimatedversionofModelBreducesthisdifferenceby21%ifconditionsdiagnosed afteradmissionareusedincodingriskfactors, and by 14% if they are not.

Figures14.3and14.4showhowaddingclinicalriskfactors affectsthelSRsofindividual hospitals,basedoneitherModelAorModelB.Thesefiguresdemonstratetheimpactof addingclinicalvariablestothebestmodelsbasedonICD -9-CMdata(e.g.,models6a and6b,whichincludeonlyconditionspresentatad mission).Althoughtoofewpatients were sampled from each hospital to generate statistically significant results at the hospitallevel,noneofthelow -mortalityorhigh -mortalityoutliersshowdramaticchanges inrisk -adjustedmortalitywhenclinicalva riablesareaddedtothemodel.

QUESTION7: Dohospitals with low risk -adjusted mortality demonstrate better processes of carethanhospitals with high risk -adjusted mortality?

Throughliteraturereviewanddiscussionwithclinicaladvisors, certainme dicationsand invasiveprocedureswereidentifiedasprocessmeasuresthatmightbeassociatedwith lowermortalityamongAMlpatients.ltwashypothesizedthatlow -mortalityhospitalsuse aspirin, heparin, thrombolytics, and beta blockers more often than high -mortality hospitals, controlling for hospital volume. It was also hypothesized that low -mortality hospitalsperformcoronaryangiography, revascularization procedures (i.e., PTCA and CABG), and pulmonary artery (Swan -Ganz) catheterization for hemody namic monitoring more often than high -mortality hospitals. These differences should be magnified by evaluatingthepromptnessoftherapy -particularlyrevascularizationwithin24hours, which has recently been shown to improve outcomes for certain AMIpa tients. Finally.it washypothesizedthatlow -mortalityhospitalshavemoreefficientemergencyroomsthan high-mortalityhospitals,sotheyshouldhaveshortertimesfrompresentationtothefirst ECG,thefirstCKdetermination,andadmission.

The res ults are summarized in Tables 14.13 and 14.14. The former table provides unweighted statistics and includes pvalues to assess performance differences across hospital categories. The estimates in the latter table are weighted to reflect the statewide population of AMIs, but do not allow assessment of statistical significance. The contraindication sused in HCFA's Cooperative Cardiovascular Project ¹⁰(CCP) were applied, although a few minor contraindications could not be matched. For example, CCP distinguishes he morrhagic from no -he morrhagics trokes and identifies patients with prolonged prothrom bin times and he parinal lergies. However, the sed is crepancies should not affect the general results. For a spirinand throm bolysis, are vised version of the CCP criteria was created by dropping marginal or relative contraindications, such as a gegreater than 80 years, strokemore than 6 months prior to admission, and he matocrit less than 30%. The foot notes to Table 14.13 list the secont rain dications in detail.

Allstatisticallysignificantdifferencesacrosshospitaloutcomeorvolumecategoriesare indicatedwithasterisks. Highvolumehospitalsprescribeaspirintoahigherpercentage of AMI patients than medium -volume hospitals, but as pirinusedoes not differ hospital mortality categories. However, low -mortalityhospitals start aspirin within 6 hours of presentation more often than intermediate or high -mortality hospitals. Thrombolyticuseisassociatedwithneitherhospitalvolumenorhospitalmorta litv.This result is unaffected by whether the CCP list of contraindications or the revised list is used.Low -mortalityandhigh -mortalityhospitalsalsodonotdifferintheuseofaspirin andheparinasearlyadjunctivetherapywiththrombolytics.Low -mortalityhospitalsdo, however,administerheparintoahigherpercentageofAMlpatientsthanhigh -mortality hospitals.

AMIpatientsadmittedtolow -volumehospitalsarelesslikelytoundergoPTCA,butare justaslikelytoundergoCABG,aspatientsa dmittedtohigh -volumehospitals.Patients admitted to high -mortality hospitals are somewhat less likely to undergo CABG, but

GrinesCL,BrowneKF,MarcoJ,RothbaumD,StoneGW,O'KeefeJ,etal.Acomparisonofimmediate angioplastywiththrombolytictherapyforacutemyocardialinfarction. NEnglJMed 1993;328:673 -679.

EllerbeckEF,JencksSF,RadfordMJ,KresowikT F,CraigAS,GoldJA,KrumholzHM,VogelRA.Quality ofcareforMedicarepatientswithacutemyocardialinfarctions. *JAMA*1995;273:1509 -1514.

almost as likely to undergo PTCA, compared with those admitted to low hospitals. Revascularization (CABGorPTCA) with in 24 hours of presentation is about twice as frequentially as in a mortality as in high mortality hospitals. Coronary angiography and pulmonary artery (Swan -Ganz) catheterization are also performed more frequently at low-mortality than at high mortality hospitals experience better outcomes than high mortality hospitals experience better outcomes than high mortality hospitals even among their patients who do not undergore vascularization. Therefore, the observed differences in risk -adjusted outcomes cannot be attributed solely to differential rates of revascularization.

Finally,therewerenosystematicdifferencesinthemeasurableefficiencyofemergency servicesbetweenlow -mortalityandhigh -mortalityhospitals.Medium -volume hospitals experiencedlessdelaytothefirstECGandthefirstCKdeterminationthanhigh -volume hospitals, although this analysis may be confounded by differences in the clinical presentation of AMIs.

Table14.1:Casesexcludedfromthevalidationsamp

Reason	Number
PrincipaldiagnosismiscodedasAMI,nodocumentationofAMI	
Unstableangina,411.1	7
Cardiacdysrhythmia,427.x	3
CoronaryocclusionwithoutAMI,411.81	3
Otheranduns pecifiedanginapectoris,413.9	1
Congestiveheartfailure,428.0	1
Malignantessentialhypertension,401.0	1
Hypertensiveheartandrenaldisease,404.93	1
Vascularmyelopathies,336.1	1
PrincipaldiagnosisofAMI,post -transferhos pitalization	4
SecondarydiagnosismiscodedasAMI,nodocumentationofAMI	4
SecondarydiagnosismiscodedasAMI, questionable documentation of AMI	1
SecondarydiagnosisofAMI,postoperativeAMI	4
TOTAL	31

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Table14.2:AMIs casesclassifiedaccordingtoclinicalcriteria

Classification	Definite	Possible	NoAMI
Chestpainpresent, positive enzymes	662		
Chestpainpresent, borderlineenzymes		93	
Chestpainpresent,normalenzymes			
EvolvingdiagnosticECGpattern	4		
AllotherECGpatterns			35
Chestpainpresent,incompleteenzymes			
EvolvingdiagnosticECGpattern	0		
DiagnosticorevolvingST -TECGpattern		5	
Equivocal,other,absent,oruncodableECGs			4
Nochestpain,positiveenzymes		115	
Nochestpain,borderlineenzymes			
EvolvingdiagnosticECGpattern	2		
Diagnostic, evolving ST -T, or equivocal ECG pattern		18	
Other,absent,oruncodableECGs			7
Nochestpain,normalenzymes			
EvolvingdiagnosticECGpattern	1		24
AllotherECGpatterns			
Nochestpain,incompleteenzymes			
EvolvingdiagnosticECGpattern Allot herECGpatterns	0		4
TOTALS	669	231	74
	68.7%	23.7%	7.6%

Table14.3:Sensitivityandspecificityofriskfactorreporting

RiskFactor	Number ofcases ¹	Sensitivity (%) ²	Specificity (%) ²	PV+ (%) ²	PV- (%) ²	LR+ ²	Карра
Anteriorwall	328	84	94	87	93	15.2	0.80
CHF	397	72	95	90	84	14.1	0.65
Chronicliver	15	8	100	82	99	636.0	0.31
Chronicrenal	63	72	99	78	99	78.0	0.61
CompleteAVblock	43	62	100	91	98	203.3	0.79
Diabetes, complicated	74	55	97	48	98	17.4	0.50
Diabetes, uncomplicated	176	66	97	82	92	19.0	0.63
Epilepsy	41	37	99	67	98	70.5	0.66
Hypertension	494	60	93	91	66	8.4	0.51
Hypotension	218	25	97	68	82	7.4	0.20
Inferior wall	286	83	94	88	92	14.8	0.81
LateeffectsofCVA	45	20	99	49	97	26.0	0.44
Othercerebrovascular disease	61	45	99	81	96	53.8	0.56
PriorCABG	101	72	98	85	97	44.6	0.73
Pulmonaryedema	82	31	98	60	95	18.5	0.43
Secondarymalignant neoplasm	8	82	100	79	100	418.1	0.46
Shock	111	64	99	82	97	55.6	0.71
Othersiteofinfarction	51	84	95	38	99	15.7	0.61
Othervalvedisease	178	21	99	83	85	23.2	0.31
Primarymalignant neoplasm	6	28	100	87	100	1225. 9	0.60
Nutritionaldeficiency	5	9	100	22	99	41.6	0.40
Historyofpacemaker	30	73	100	95	100	1022. 0	0.77
Skinulcer	20	48	100	76	99	230.6	0.60
Diabetes(any)	250	82	98	94	94	50.7	0.88
Hypertension(any)	554	64	95	95	65	12.3	0.54

Thiscolumnindicatesthenumberofcaseswiththeriskfactor,accordingtoCMRI'sreabstractionoftherecordsin theAMIvalidationdataset.

Thefiguresinthesecolumnsareweightedtocompe nsatefortheoversamplingofoutlierhospitalsanddeaths;the weightedestimatesapproximatethetruevalueoftheseparametersamongallAMlpatientsadmittedtoCalifornia hospitals(exceptlow -volumehospitals).

Table14.4:Sensitivityandpredictiv evalueofriskfactorreporting,byhospitalvolume andoutcomecategory

Riskfactor(cases) 1	Hospita	almortality	category		<u>Hosp</u>	oitalvolume	category
Measure	Better	Neither	Worse	pvalue	High	Medium	pvalue
Antoriomyall/n 220)							
Anteriorwall(n=328)	0.4	00	02	0.020*	O.F.	00	0.627
Sensitivity,%	84	82	93	0.030*	85	88	0.627
Adjustedsensitivity ²	80	83	92	0.770	83	86	0.004
PV+,% ³	85	88	86	0.776	85	87	0.634
Карра	0.76	0.79	0.84	0.218	0.78	0.81	0.449
CHF(n=397)							
Sensitivity,%	69	74	65	0.329	64	74	0.039*
Adjustedsensitivity	66	70	66		66	71	
PV+,%	83	89	90	0.291	93	83	0.015*
Карра	0.60	0.69	0.64	0.388	0.64	0.65	0.738
Chronicrenal(n=63)							
Sensitivity,%	40	80	52	0.044*	59	50	0.614
Adjustedsensitivity	36	75	47		54	41	
PV+,%	67	80	80	0.749	81	71	0.503
Kappa	0.47	0.79	0.61	0.043*	0.66	0.56	0.360
Hypotension(n=218)							
Sensitivity,%	17	27	14	0.125	22	17	0.396
Adjustedsensitivity	11	18	12	0.123	15	14	0.550
PV+,%	48	65	50	0.425	65	45	0.110
Kappa	0.16	0.29	0.14	0.423	0.23	0.15	0.110
	0.10	0.29	0.14	0.100	0.23	0.13	0.211
Inferiorwall(n=286)	0.4	00	00	0.000	00	00	0.447
Sensitivity,%	84	83	86	0.898	86	82	0.417
Adjustedsensitivity	85	83	88	0.533	88	84	0.504
PV+,%	87	88	92	0.577	90	88	0.564
Kappa	0.80	0.79	0.85	0.413	0.84	0.79	0.221
Othercerebrovascular(n=61)							
Sensitivity,%	47	50	44	0.947	47	48	1.000
Adjustedsensitivity	42	42	46		42	49	
PV+,%	69	81	70	0.721	83	63	0.264
Карра	0.54	0.59	0.52	0.871	0.58	0.52	0.661
PriorCABG(n=101)							
Sensitivity,%	64	74	75	0.602	64	77	0.193
Adjustedsensitivity	62	69	76		63	74	
PV+,%	81	83	81	1.000	79	84	0.590
Kappa	0.68	0.76	0.76	0.619	0.68	0.78	0.156
Pulmonaryedema(n=82)		-	-	-		-	
Sensitivity,%	54	31	45	0.226	38	49	0.374
Adjustedsensitivity	36	36	45	0.220	36	41	0.017
PV+,%	44	67	56	0.345	59	47	0.461
Kappa	0.43	0.39	0.47	0.833	0.42	0.44	0.461
Ναμμα	0.43	0.38	0.47	0.033	0.42	0.44	0.008

Table 14.4: Sensitivity and predictive value of risk factor reporting, by hospital volume and outcome category, continued

Riskfactor(cases) 1	<u>Hospita</u>	almortality	<u>category</u>	<u>Hospitalvolumecategory</u>			
Measure	Better	Neither	Worse	pvalue	High	Medium	pvalue
Shock(n=111)							
Sensitivity,%	67	67	56	0.589	57	71	0.166
Adjustedsensitivity	48	56	54		48	61	
PV+,%	97	83	79	0.068*	81	93	0.173
Kappa	0.76	0.71	0.63	0.425	0.63	0.78	0.056*
Othersite(n=51)							
Sensitivity,%	83	88	73	0.548	81	80	1.000
Adjustedsensitivity	89	85	76		81	81	
PV+,%	50	44	67	0.258	61	43	0.173
Карра	0.61	0.56	0.67	0.609	0.67	0.54	0.198
Othervalvedis(n=178)							
Sensitivity,%	27	19	23	0.599	30	18	0.076*
Adjustedsensitivity	20	22	18		26	15	
PV+,%	82	83	74	0.830	86	71	0.194
Карра	0.34	0.27	0.29	0.759	0.39	0.22	0.031*
Diabetes(any,n=250)							
Sensitivity,%	92	82	88	0.203	87	88	0.852
Adjustedsensitivity	91	82	87		86	88	
PV+,%	95	95	97	0.780	97	95	0.537
Карра	0.91	0.84	0.90	0.281	0.89	0.88	0.928
Hypertension(any,n=554)							
Sensitivity,%	64	63	60	0.781	57	68	0.011*
Adjustedsensitivity	64	61	59		59	67	
PV+,%	94	94	95	0.960	97	92	0.114
Kappa	0.53	0.54	0.55	0.957	0.50	0.59	0.045*

^{*} Statisticallysignificantatp<0.10

Onlyriskfactorspresentinatleast5%ofcases(n=49)areshowninthistable.

AdjustedsensitivitiesreflectChoi'scorrectionforsampleselectio nbias. Thereisnoprocedureforevaluatingthe statisticalsignificanceofdifferencesintheadjustedsensitivityacrosshospitalcategories, because the variance of the adjustedsensitivity is unknown. The results of the adjusted and unadjusted sensitivity analyses are generally similar.

³ Oversamplingofdeathsdoesnotbiasthepositivepredictivevalue,sonoadjustmentisnecessary.

Table14.5:Weightedindirectlystandardizedmortalityratiosbyhospitalmortalityand volumecategory,usingri sk-adjustmentmodelsbasedonICD -9-CMcodeddata ¹

	Hospital	volumecate	egory_				
Risk-adjustmentmodel	Better	Neither	Worse	Difference	High	Medium	Difference
Martala (OOLIDD Iara							
Model1a(OSHPDdata, 1993coeff icients) Model2a(CMRIdata,	0.7441	1.2052	1.4494*	0.7052	1.1114	1.2412	0.1299
1993coefficients)	0.8269	1.4556*	1.5476*	0.7208	1.3547*	1.4159*	0.0612
Model1b(OSHPDdata, 1993coefficients) Model2b(CMRIdata,	0.7007*	1.2015	1.4701*	0.7694	1.0690	1.2692	0.2002
1993coefficients)	0.6270*	1.0334	1.1789	0.5519	0.9627	1.0293	0.0666
Model3a(OSHPDdata, reestimatedcoefficients) Model4a(CMRIdata,	0.6775*	1.0273	1.2012	0.5237	0.9700	1.0390	0.0690
reestimatedcoefficients)	0.6151*	1.0534	1.0736	0.4585	1.0090	0.9893	0.0197
Model3b(OSHPDdata, reestimatedcoefficients) Model4b(CMRIdata,	0.6084*	1.0467	1.1933	0.5849	0.9307	1.0920	0.1613
reestimatedcoefficients)	0.6282*	1.0523	1.0751	0.4469	1.0210	0.9772	0.0438

^{*} Thisindirectlystandardizedmortalityratioisstatisticallysignificantlydifferentfromone,whichrepresentstheaverage statewidemortalityex perienceunderthismodel.

Risk-adjustmentmodels1a,2a,3a,and4aarebasedonModelA,whereasmodels1b,2b,3b,and4barebasedon ModelB.ModelBdiffersfromModelAinthatitincludesrace,expectedprincipalsourceofpayment,sourceand typeofadmission,andclinicalfactorsthatmayrepresenteitherriskfactorsorcomplicationsModels1a,2a,3a,and 4ainclude974cases;models1b,2b,3b,and4binclude938cases.

Table14.6:Performancecharacteristicsofrisk -adjustmentmodelsb asedonICD -9-CM codeddata

	<u></u>	estatistic ¹	Calibra	ationcoeffi	icients ²	PartialR ²
Risk-adjustment model	Estimate	95%confidence interval	Intercept	Linear slope	Quadratic slope	(TypeII)for hospital-level mortality ³
Model1a(O SHPD data,1993OSHPD coefficients)	0.766 ⁴ 0.775 ⁵	0.742 -0.808	0.248	1.022	-0.003	0.388
Model2a(CMRI data,1993OSHPD coefficients)	0.799 ⁵	0.768 -0.830	1.099	1.437	0.029	0.395
Model1b(OSHPD data,1993OSHPD coefficients)	0.844 ⁴ 0.836 ⁵	0.806 -0.866	0.180	0.904	-0.025	0.133
Model2b(CMRI data,1993OSHPD coefficients)	0.869 ⁵	0.842 -0.896	0.055	0.987	-0.023	0.435
Model3a(OSHPD data,reestimated coefficients)	0.782	0.749 -0.815	0.045	1.149	0.050	0.425
Model4a(CMRI data,reestimated coefficients)	0.814	0.783 -0.845	0.000	1.036	0.014	0.327
Model3b(OSHPD data,reestimated coefficients)	0.841	0.810 -0.871	0.000	0.997	-0.001	0.190
Model4b(CMRI data,reestimated coefficients)	0.881	0.855 -0.908	0.009	0.985	-0.007	0.489

Thecstatisticisameasureofdiscrimination,oramodel'sabilitytodistinguishindividualswhohadapooroutcome fromthosewhohadagoodoutcome.ltrepresentstheproportionofall randomlyselectedpairsofobservationswith differentoutcomesinwhichthepatientwhodiedhadahigherexpectedprobabilityofdeaththanthesurvivor.This statisticisequivalenttotheareaunderareceiveroperatingcharacteristiccurve, whichpl otssensitivityversus1 - specificityatvariouscutoffvaluesforthepredictedprobability(seeHanleyJA,McNeilBJ.Themeaninganduseof theareaunderareceiveroperatingcharacteristic(ROC)curve. Radiology1982;143:29 -36).Thecstatistictakes on valuesbetween0and1;highervaluesindicategreaterdiscriminationbutthereisnocutoffthatidentifiesinadequate models.Avalueof0.5canbeobtainedbyrandomselection.

Calibrationcoefficients assess the agreement between predicted prob abilities generated by a logistic model and observed outcomes. A weighted logistic model is to use dtoregress the logit fobserved mortality against the logit and logit squared of predicted mortality across all covariate patterns (see Miller ME, HuiSL, Tierney WM. Validation techniques for logistic regression models. Stat Med 1991;10:1213 -1226). The ideal values of the intercept and the quadratics lopear ezero; the ideal value of the linear slope is one (see Miller ME, Langefeld CD, Tierney WM, Hui SL, McDonald CJ. Validation of probabilistic predictions. Med Decis Making 1993;13:49 -58). Because deaths were oversampled in the validation study, weighted analyses are essential and the statistical significance of the

coefficients cannot be determined. Instead, the reader should use the statistics reported in these columns to compare the calibration of hierarchical models. The same problem precludes use of the Hosmer -Lemeshow goodness-of-fittest(seeHosmerDW,LemeshowS. *AppliedLogisticRegression* .NewYork:JohnWiley&Sons, 1989).

- ThepartialR ²representsthesquaredTypeIIpartialcorrelationbetweentheobservedandexpectednumbersof weighteddeathsateachhospital,controllingforthetotalnumberofweightedcasesAllvariablesare logarithmically transformedtoreduceheteroscedasticity.ThemethodofKronmal(KronmalRA.Spuriouscorrelationandthefallacy oftheratiostandardrevisited. *JRoyalStatAssoc* 1993;156(3):379 -392)wasusedtoavoidspuriouscorrelationsthat mayap pearwhenratiosareregressed.
- ThesecstatisticswerederivedfromtheoriginalAMIsampleincludedinthe1993reportoftheCaliforniaHospital OutcomesProject.
- ⁵ ThesecstatisticswerederivedbyapplyingOSHPD's1993modelstothevalidations ample.Theydifferfromthec statisticsderivedfromtheoriginalstatewidesampleonlybecauseofsamplingvariation.

Table14.7:Timingofriskfactors

		Dated	ofDocumen	SourceofDocumention	
		ER/admit ¹	Nextday,	>1dayafter ER/admit,	ER/admitnote,
RiskFactor	Total	date,no.(%)	no.(%)	no.(%)	no.(%)
Anteriorwall	328	295(89.9)	31(9.5)	2(0.6)	324(98.8)
CHF	397	267(67.3)	58(14.6)	72(18.1)	271(68.3)
Chronicliver	15	9(60.0)	3(20.0)	3(20.0)	8(53.3)
Chronicrenal	63	53(84.1)	6(9.5)	4(6.3)	59(93.7)
CompleteAVblock	43	28(65.1)	4(9.3)	11(25.6)	24(55.8)
Diabetes, complicated	74	60(81.1)	11(14.9)	3(4.1)	69(93.2)
Diabetes, uncomplicated	176	145(82.4)	24(13.6)	7(4.0)	168(95.5)
Epilepsy	41	27(65.9)	4(9.8)	10(24.4)	28(68.3)
Hypertensionwithout					
CHForrenalfailure	494	434(87.9)	53(10.7)	7(1.4)	473(95.8)
Hypotension	218	103(47.2)	34(15.6)	81(37.2)	98(45.0)
Inferiorwall	286	265(92.7)	17(5.9)	4(1.4)	279(97.6)
LateeffectsofCVA	45	37(82.2)	1(2.2)	7(15.6)	37(82.2)
Other					
cerebrovasculardz	61	27(44.3)	10(16.4)	24(39.3)	28(45.9)
PriorCABG	101	93(92.1)	8(7.9)	0(0.0)	100(99.0)
Pulmonaryedema	82	31(37.8)	17(20.7)	34(41.5)	31(37.8)
Secondarymalignant	_	- (\)		2/2- 2)	- ()
neoplasm	8	5(62.5)	1(12.5)	2(25.0)	6(75.0)
Shock	111	57(51.4)	21(18.9)	33(29.7)	51(46.0)
Othersiteof infarction	51	37(72.5)	10(19.6)	4(7.8)	44(86.3)
Othervalvedisease Primarymalignant	178	48(27.0)	45(25.3)	85(47.8)	33(18.5)
neoplasm	6	5(83.3)	0(0.0)	1(16.7)	5(83.3)
Nutritionaldeficiency	5	3(60.0)	0(0.0)	2(40.0)	3(60.0)
Historyofpacemaker	30	22(73.3)	8(26.7)	0(0.0)	30(100.0)
Skinulcer	20	10(50.0)	4(20.0)	6(30.0)	11(55.0)
Hypertension(any)	554	485(89.0)	60(11.0)	0(0.0)	530(95.7)

 $^{{\}footnotesize \label{thm:patient} 1 \\ \ \ \, \text{The day of presentation is the day of arrival in the emergency room or the day of admission, if the patient was not admitted through the emergency room.}$

Table 14.8: Weighted indirectly standardized mortality ratios by hospital mortality and volume category, using risk -adjustment models that include only risk factors reabstracted from ER or admission notes

-										
	<u>Hc</u>	ospitalmo	rtalitycate	gory	<u>Hospitalvolumecategory</u>					
Risk-adjustmentmodel	Better	Neither	Worse	Difference	High	Medium	Difference			
Model2a(CMRIdata, 1993coefficients) Model5a(CMRIdata fromER/admitnotes,	0.8269	1.4556*	1.5476*	0.7208	1.3547*	1.4159*	0.0612			
1993coefficients)	0.8174	1.4992*	1.5957*	0.7783	1.5069*	1.3481*	0.1588			
Model4a(CMRIdata, reestimatedcoefficients) Model6a(CMRIdata fromER/admitnotes,	0.6151*	1.0534	1.0736	0.4585	1.0090	0.9893	0.0197			
reestimatedcoefficients)	0.5506*	1.0683	1.1236	0.5731	1.0629	0.9535	0.1093			
Model2b(CMRIdata, 1993coefficients) Model5b(CMRIdata fromER/admitnotes,	0.6270*	1.0334	1.1789	0.5519	0.9627	1.0293	0.0666			
1993coefficients)	0.7967	1.3407*	1.6199*	0.8232	1.2665	1.3181	0.0516			
Model4b(CMRIdata, reestimatedcoefficients) Model6b(CMRIdata fromER/admitnotes,	0.6282*	1.0523	1.0751	0.4469	1.0210	0.9772	0.0438			
reestimatedcoefficients)	0.5820*	1.0617	1.1174	0.5355	0.9991	1.0010	0.0019			

^{*} Thisindirectlystandardizedmortalityratioisstatisticallysignificantlydifferentfromone,whichrepresentstheaverage statewidemortalityexperienceunderthismodel.

Risk-adjustmentmodels2a,4a,5a,and6aareba sedonModelA,whereasmodels2b,4b,5b,and6barebasedon ModelB.ModelBdiffersfromModelAinthatitincludesrace,expectedprincipalsourceofpayment,sourceand typeofadmission,andclinicalfactorsthatmayrepresenteitherriskfactors orcomplicationsModels2a,4a,5a,and 6ainclude974cases;models2b,4b,5b,and6binclude938casesbecauseofmissingvalues.

Table 14.9: Performance characteristics of risk -adjustment models with and without additional clinical variables 1

		cstatistic 1	Calibra	tioncoef	ficients ²	PartialR ²
Risk-adjustmentmodel	Estimate	95%confidence interval	Intercept		Quadratic slope	(TypeII)for hospital-level mortality ³
Model3a(OSHPD data,noclinical variables)	0.782	0.749 -0.815	0.066	1.182	0.059	0.476
Model7a(OSHPD data,coreclinical variables)	0.845	0.816 -0.875	0.008	1.132	0.047	0.296
Model8a(OSHPD data,core&secondary clinicalvariables)	0.854	0.825 -0.883	-0.004	1.128	0.047	0.321
Model4a(CMRIdata, noclinicalvariables)	0.814	0.783 -0.845	0.001	0.985	-0.006	0.394
Model9a(CMRIdata, coreclinicalvariables)	0.860	0.831 -0.889	-0.017	1.071	0.027	0.494
Model10a(CMRIdata, core&secondary clinical variables)	0.864	0.835 -0.892	-0.026	1.099	0.038	0.488
Model6a(CMRIdata fromER/admitnotes, noclinicalvariables) Model11a(CMRIdata	0.786	0.753 -0.820	0.018	1.087	0.031	0.287
fromER/admitnotes, coreclinicalvariables)	0.843	0.814 -0.873	-0.002	1.071	0.027	0.406
Model12a(CMRIdata fromER/admitnotes, core&secondary clinicalvariables)	0.845	0.814 -0.875	0.003	1.129	0.046	0.371
Model3b(OSHPD data,noclinical variables)	0.837	0.806 -0.868	-0.001	1.008	0.003	0.170
Model7b(OSHPD data,coreclinical variables)	0.870	0.842 -0.897	-0.013	1.055	0.021	0.249
Model8b(OSHPD data,core&secondary clinicalvariables)	0.877	0.851 -0.904	-0.016	1.060	0.023	0.276

Table14.9:Performancecharacteristicsofrisk -adjustmentmodelswithandwithoutadditional clinicalvariables ¹,continued

-						
	<u>c</u>	statistic ¹	<u>Calibra</u>	tioncoe	fficients ²	PartialR ² (TypeII)
Risk-adjustmentmode I		95%confidence		Linear	Quadratic	forhospital -level
•	Estimate	interval	Intercept	slope	slope	mortality ³
Model4b(CMRIdata, noclinicalvariables)	0.879	0.852 -0.907	0.009	0.986	-0.007	0.470
Model9b(CMRIdata, coreclinicalvariables)	0.898	0.872 -0.923	0.028	0.966	-0.017	0.539
Model10b(CMRIdata, core&secondary clinicalvariables)	0.898	0.873 -0.923	0.016	0.981	-0.009	0.556
Model6b(CMRIdata fromER/admitnotes, noclinicalvariables)	0.815	0.782 -0.848	-0.007	1.063	0.025	0.310
Model11b(CMRIdata fromER/admitnotes, coreclinicalvariables)	0.852	0.822 -0.882	0.002	0.993	-0.003	0.359
Model12b(CMRIdata fromER/admitnotes, core&secondary clinicalvariables)	0.859	0.829 -0.888	-0.013	1.071	0.028	0.357

Thecstatisticisameasureofdiscrimination,oramodel'sabilitytodistinguishindividualswhohadapooroutcomefrom thosewhohadagoodoutcome(seeTable14.6foradditionaldescription).

Calibrationcoefficientsass esstheagreementbetweenpredictedprobabilitiesgeneratedbyalogisticmodelandobserved outcomes Aweightedlogisticmodelistousedtoregressthelogitofobservedmortalityagainstthelogitandlogitsquared of predicted mortality across all co variate patterns (see Table 14.6 for additional description). The ideal values of the interceptandthequadraticslopearezero; the ideal value of the linear slope is one. The readershould use the statistics reported in the secolumn stocompare the call bit at interceptand the secolumn stocompare the call bit at interceptand the secolumn stocompare the call bit at interceptand the second statistic and the second statistic

ThepartialR ²representsthesquaredTypellpartialcorrelationbetweentheobservedandexpectednumbersofweighted deathsateachhospital,controllingforthetotalnum berofweightedcases(seeTable14.6foradditionaldescription).

 $Table\,14.10: A cute\,myocardial\,in farction\,Model\,A, no\,prior\,admission\,(reestimated\,parameter\,estimates\,and odds ratios using validation sample)$

RiskFactor	1993m	odel	Model (OSHPDd noadditionald variable	ata, clinical	Model (OSHPDda clinicalvaria	ta,all	Model((CMRIdatafromE notes,noaddi clinicalvarial	R/admit tional	Model 1 (CMRIdatafromb notes,allclinicaly	ER/admit
	Value	OR	Value	OR	Value	OR	Value	OR	Value	OR
Intercept	-3.0808		-2.9244		-3.3565		-3.1710		-3.2614	
Female	0.1574	1.17	0.2881	1.33	0.0781	1.08	0.0702	1.07	-0.0387	0.96
Age18 -40 ¹	-1.5206	0.22	-1.8021	0.16	-2.4325	0.09	-1.8201	0.16	-2.6201	0.90
Age41 -55 ¹	-1.2456	0.22	-1.0021	0.10	-2.4020	0.03	-1.0201	0.10	-2.0201	0.07
Age56 -65	-0.6150	0.54	-1.1684	0.31	-1.5598	0.21	-1.5194	0.22	-1.8675	0.15
Age76 -85	0.5045	1.66	0.2453	1.28	0.1886	1.21	0.3378	1.40	0.5009	1.65
Age >86	1.0780	2.94	0.5143	1.67	0.7381	2.09	0.5342	1.71	-0.0728	0.93
Anteriorwall	1.4185	4.13	1.2577	3.52	1.2812	3.60	1.9736	7.20	2.1582	8.66
CHF	0.5670	1.76	-0.0064	0.99	-0.3307	0.72	0.4950	1.64	-0.0854	0.92
Chronicliver	1.1743	3.24	4.1570	63.88	3.2771	26.50	2.5741	13.12	2.5091	12.29
Chronicrenal	0.3244	1.38	0.6425	1.90	0.8240	2.28	1.1145	3.05	0.9471	2.58
Diabetes,	0.4658	1.59	0.6665	1.95	0.8769	2.40	0.7387	2.09	0.8004	2.23
complicated										
Diabetes,	0.0383	1.04	0.9775	2.66	1.3025	3.68	0.9280	2.53	1.4391	4.22
uncomplicated										
Hypertension	-0.5779	0.56	-0.5462	0.58	-0.3505	0.70	-0.6605	0.52	-0.5966	0.55
Inferiorwall	1.0944	2.99	1.1931	3.30	1.3340	3.80	1.4757	4.37	1.3893	4.01
Lateeffectsof CVA	0.3648	1.44	1.9226	6.84	1.5169	4.56	2.1009	8.17	2.0370	7.67
PriorCABG	-0.0841	0.92	-0.4600	0.63	0.1975	1.22	-0.4521	0.64	-0.4619	0.63
Secondary malignant neoplasm	0.7533	2.12	-0.0749	0.93	-1.2044	0.30	-1.9750	0.14	-1.9364	0.14
Othersiteof infarction	2.2115	9.13	1.4880	4.43	0.7808	2.18	3.0216	20.52	2.3451	10.43
CHF& Age41 -55	0.7695	2.16	3.0330	20.76	2.9258	18.65	2.8052	16.53	3.8588	47.41
CHF& Age56 -65	0.4149	1.51	-0.1442	0.87	-0.1565	0.86	0.4868	1.63	0.2474	1.28
CHF& Age>86	-0.4371	0.65	0.3411	2.15	0.6922	2.00	0.1189	1.13	1.7001	5.47
CHF& Anteriorwall	-0.2397	0.79	0.7660	2.40	1.0598	2.89	1.2780	3.59	-0.2617	0.77
Female& Age56 -65	0.2138	1.24	0.8742	0.05	-2.5547	0.08	-3.6915	0.02	1.4414	4.23
Inferiorwall& Age41 -55	-0.7049	0.49	-3.0469	1.23	0.4024	1.50	0.3170	1.37	-3.9990	0.02

 $\label{thm:continuous} Table\,14.10:\,Acute\,myocardial\,in farction\,Mo\qquad del\,A,\,no\,prior\,admission\,(reestimated\,parameter\,estimates\,and odds ratios using validation sample), continued$

RiskFactor	1993m	odel	ModelS (OSHPDdata additionalclinicalv	a,no	Modelo (OSHPDd allclinicalvaria	ata,	Model6a (CMRIdatafromER/admit notes,noadditional clinicalvariables)		Model 1 (CMRIdatafromE notes,allclinicalva	R/admit
	Value	OR	Value	OR	Value	OR	Value	OR	Value	OR
Inferiorwall& Age56 -65	-0.4120	0.66	0.2101	6.63	2.2513	9.50	0.9041	2.47	0.3344	1.40
CHF&Othersite	-0.8908	0.41	1.8914	0.30	-1.6541	0.19	-1.4412	0.24	1.7111	5.54
Othervalve disease	-0.4078	0.67	-1.2127	0.52	-1.2360	0.29	-1.8137	0.16	-1.0746	0.34
Othersite & Age56 -65	0.4199	1.52	-0.6611		2.2257	9.26			-0.6528	0.52
DNR					2.5887	13.31			2.1172	8.31
Cardiopulmonary arrest					-0.0130	0.99			2.6756	14.52
Systolicblood pressure					0.0177	1.02			-0.0130	0.99
Heartrate					1.5841	4.87			0.0145	1.01
Shock(clinical)					0.0926	1.10			1.4259	4.16
FirstCKindex					0.4627	1.59			0.0883	1.09
Rales					0.3742	1.45			0.4079	1.50
Systolicheart murmur					1.0366	2.82			0.4195	1.52
Historyofstroke (clinical)									0.1531	1.17

Thesetwovariableswerecombinedintooneriskfactor(age18 sample.

 $[\]hbox{-}55 years) for all models estimated on the validation$

 $Table\,14.11: Acute\,my \quad ocardial\,in farction\,Model\,B, no\,prior\,admission\,(reestimated\,parameter\,estimates\,and odds ratios using validation sample)$

	1993m	odel	Mode (OSHPDdatan variabl	oclinical	Mode (OSHPDd clinicalvan	lataall	Mod (CMRIdatafro notes,noclinio	mER /admit	Model (CMRIdata ER/admitno clinicalvaria	afrom etes,all
RiskFactor	Value	OR	Value	OR	Value	OR	Value	OR	Value	OR
Intercept	-3.4275	0.03	-3.1802		-3.4451		-3.1707		-2.8891	
Female	0.1470	1.16	0.3346	1.40	0.2880	1.33	0.0526	1.05	-0.0552	0.95
Age18 -40 ¹	-1.4958	0.22	-1.7079	0.18	-2.7790	0.06	-2.1848	0.11	-3.4902	0.03
Age41 -55 ¹	-1.0211	0.36								
Age56 -65	-0.4704	0.62	-1.1738	0.31	-1.5811	0.21	-1.6018	0.20	-2.1330	0.12
Age76 -85	0.6067	1.83	0.8677	2.38	0.8002	2.23	0.5233	1.69	0.7501	2.12
Age>86	1.1307	3.10	0.7573	2.13	0.6993	2.01	0.7892	2.20	0.1854	1.20
Race:black	-0.0171	0.98	0.5714	1.77	0.8779	2.41	-1.0165	0.36	-0.3440	0.71
Race:Hispanic	0.0854	1.09	0.0861	1.09	0.0021	1.00	-0.6266	0.53	-0.6097	0.54
Race:other nonwhite	-0.0476	0.95	0.4503	1.57	0.1320	1.14	0.2734	1.31	-0.2450	0.78
Type:urgentor elective	-0.3881	0.68	-0.2728	0.76	-0.1324	0.88	0.4774	1.61	0.4634	1.59
Source:ER	0.0200	1.02	-0.3948	0.67	-0.6076	0.54	0.4485	1.57	0.4645	1.59
MediCal	0.3522	1.42	-0.5776	0.56	-1.2217	0.29	-2.8854	0.06	-2.7192	0.07
Medicare	0.1782	1.20	-0.0781	0.92	-0.2970	0.74	-0.6461	0.52	-1.0865	0.34
Uninsured	0.2949	1.34	0.4998	1.65	0.1784	1.20	0.6066	1.83	0.0375	1.04
Anteriorwall	1.2160	3.37	1.3351	3.80	1.5687	4.80	1.7928	6.01	2.0051	7.43
CHF	0.3335	1.40	-0.0624	0.94	-0.3407	0.71	0.5737	1.77	-0.1697	0.84
Chronicliver	1.1069	3.02	4.6002	99.50	3.7791	43.78	2.6778	14.55	2.7178	15.15
Chronicrenal	0.3279	1.39	0.3180	1.37	0.6092	1.84	1.0024	2.72	0.7880	2.20
CompleteAV block	0.5835	1.79	-0.1835	0.83	-0.4758	0.62	0.7729	2.17	0.3434	1.41
Diabetes, complicated	0.3906	1.48	1.0578	2.88	1.1397	3.13	0.5423	1.72	0.6607	1.94
Diabetes, uncomplicated	0.0557	1.06	1.3159	3.73	1.6567	5.24	1.1349	3.11	1.6320	5.11
Epilepsy	1.2591	3.52	0.8924	2.44	0.1912	1.21	0.0108	1.01	-0.6665	0.51
Hypertension	-0.4740	0.62	-0.5441	0.58	-0.4247	0.65	-0.6611	0.52	-0.5943	0.55
Hypotension	0.4911	1.63	0.3195	1.38	-0.2624	0.77	-0.1048	0.90	-0.9531	0.39
Inferiorwall	0.8124	2.25	0.8778	2.41	1.0878	2.97	1.2951	3.65	1.2974	3.66
LateeffectsCVA	0.2137	1.24	2.1433	8.53	1.9352	6.93	2.2282	9.28	1.9812	7.25
Other	0.7112	2.04	1.2398	3.45	1.1137	3.05	-0.3033	0.74	0.0242	1.02
cerebrovascular										
disease	0.0507	0.00	0.0040	0.55	0.04.47	0.00	0.0500	0.50	0.0440	0.50
PriorCABG	-0.0507		-0.6043	0.55	-0.0147	0.99	-0.6590	0.52	-0.6442	0.53
Pulmonary edema	0.9532	2.59	1.3112	3.71	0.9079	2.48	0.1734	1.19	-0.4246	0.65

	1993m	odel	Mode (OSHPDdata variab	noclinical	(OSHPDa	Model8b (OSHPDdataall (CMRIdatafromER /admit notes,noclinicalvariables)		Model12b (CMRIdatafrom ER/admitnotes,all clinicalvariables)		
RiskFactor	Value	OR	Value	OR	Value	OR	Value	OR	Value	OR

 $Table\,14.11: Acut \quad e\,myocardial\,in farction\,Model\,B, no\,prior\,admission\,(reestimated\,parameter\,estimates\,and odds ratios using validation sample), continued$

	1993m	odel	Mode (OSHPDdatar variabl	noclinical	Model (OSHPDdataal variable	llclinical	Model (CMRIdatafrominotes, noclinical)	ER/admit	Model (CMRIdatafron notes,allclinica	nER/admit
RiskFactor	Value	OR	Value	OR	Value	OR	Value	OR	Value	OR
Secondary malignant neoplasm	0.9146	2.50	-1.5436	0.21	-2.2980	0.10	-0.8713	0.42	-0.9963	0.37
Shock	2.5734	13.11	3.1970	24.46	3.3009	27.14	2.3212	10.19	1.6943	5.44
Othersiteof infarction	2.0517	7.78	0.7801	2.18	0.1218	1.13	3.1675	23.75	2.3290	10.27
CHF& Age 41-55	0.4567	1.58	2.8144	16.68	3.4310	30.91	3.3612	28.82	4.9900	146.9
CHF& Age56 -65	0.1514	1.16	-0.8766	0.42	-0.3630	0.70	0.2321	1.26	0.2350	1.26
CHF&Age>86	-0.2368	0.79	0.4723	1.60	-1.0276	0.36	0.5479	1.73	1.2819	3.60
CHF&Anterior wall	-0.2180	0.80	0.3501	1.42	0.1992	1.22	-0.1508	0.86	-0.2832	0.75
Female& Age56 -65	0.1235	1.13	0.8024	2.23	0.9829	2.67	1.1942	3.30	1.6747	5.34
Inferiorwall& Age41 -55	-0.7117	0.49	-2.3718	0.09	-2.1078	0.12	-3.8124	0.02	-4.0970	0.02
Inferiorwall& Age56 -65	-0.4268	0.65	0.6914	2.00	0.5918	1.81	-0.0743	0.93	-0.1354	0.87
CHF&Othersite	-0.8179	0.44	2.6475	14.12	2.7769	16.07	0.7790	2.18	1.8660	6.46
Othervalve disease	-0.3662	0.69	-1.7239	0.18	-1.7306	0.18	-1.1301	0.32	-1.0242	0.36
Othersite& Age56 -65	0.4528	1.57	0.2914	1.34	-0.4806	0.62	-1.9354	0.14	-1.3261	0.27
DNR					2.3244	10.22			2.2970	9.94
Cardiopulmonary arrest					3.1404	23.11			2.9759	19.61
Systolicblood pressure					-0.0112	0.99			-0.0127	0.99
Heartrate					0.0166	1.02			0.0158	1.02
Shock(clinical)					1.4161	4.12			1.2084	3.35
FirstCKindex					0.0939	1.10			0.0964	1.10
Rales					0.2952	1.34			0.3310	1.39
Systolicheart murmur					0.6628	1.94			0.9186	2.51
HistoryofCVA (clinical)					1.0895	2.97			0.1144	1.12

¹ Thesetwovariableswerecombin	-55years)forallmodelsestimatedonthevalidationsample.

Table14.12:Weightedindirectlystandardizedmortalityratiosandconfidenceintervals, byhospitalmortalityandvolumecategory,usingrisk -adjustmentmodels withandwithout additionalclinicalriskfactors

	Hospital	mortality	category		Hospita	lvolumec	ategory
Risk-adjustmentmodel	Better	Neither			High		Difference
Model3a (OSHPDdata,no	0.6757*	1.0292	1.2030*	0.5273	0.9631	1.0437	0.0805
additionalclin icalvariables) Model7a(OSHPDdata, coreclinicalvariables)	0.6832*	1.0347	1.1222	0.4391	0.9976	1.0026	0.0050
Model8a (OSHPDdata, core&secondaryclinical variables)	0.6684*	1.0362	1.1426	0.4743	0.9956	1.0048	0.0092
Model4a(CMRIdata,no additionalclinicalvariables)	0.6430*	1.0518	1.0519	0.4090	1.0192	0.9798	0.0395
Model9a (CMRIdata,core clinicalvariables)	0.6819*	1.0419	1.0544	0.3724	1.0684	0.9345	0.1339
Model10a (CMRIdata,core& secondaryclinicalvariables)	0.6759*	1.0407	1.0791	0.4032	1.0689	0.9341	0.1348
Model6a (CMRIdatafrom ER/admitnotes,no additionalclinicalvariables)	0.5689*	1.0704	1.0871	0.5182	0.9633	1.0435	0.0802
Model11a (CMRIdatafrom ER/admitnotes,core clinicalvariables)	0.6149*	1.0565	1.0799	0.4650	1.0097	0.9896	0.0200
Model12a (CMRIdatafrom ER/admitnotes,core& secondaryclinical variables)	0.5993*	1.0583	1.1074	0.5082	1.0099	0.9894	0.0205
Model3b(OSHPDdata,no additionalclinicalvariables)	0.6137*	1.0447	1.2085	0.5955	0.9251	1.0971	0.1720
Model7b(OSHPDdata, coreclinicalvariables)	0.6282*	1.0519	1.0880	0.4598	0.9414	1.0730	0.1316
Model8b(OSHPDdata , core&secondaryclinical variables)	0.6287*	1.0499	1.1060	0.4773	0.9453	1.0675	0.1222
Model4b(CMRIdata,no additionalclinicalvariables)	0.6386*	1.0487	1.0915	0.4529	1.0179	0.9811	0.0368
Model9b (CMRIdata,coreclinical variables)	0.6720*	1.0457	1.0418	0.3698	1.0366	0.9628	0.0739
Model10b (CMRIdata,core& secondaryclinicalvariables)	0.6823*	1.0436	1.0388	0.3565	1.0509	0.9497	0.1011
Model6b(CMRIdatafrom	0.5875*	1.0603	1.1246	0.5371	0.9927	1.0081	0.0153

							
	<u>Hospital</u>	mortality	category	<u>Hospitalvolumecategory</u>			
Risk-adjustmentmodel	Better	Neither	Worse	Difference	High	Medium	Difference
ER/admitnotes,no additionalclinicalvariables) Model11b(CMRIdatafrom ER/admitnotes,core clinicalvariables) Model12b (CMRIdatafrom ER/admitnotes,core& secondaryclinical variables)	0.6115* 0.6086*			0.4362 0.4645	1.0060 1.0051	0.9935 0.9944	0.0124 0.0107

^{*} Thisindirectlystandardizedmortalityratioisstatisticallysignificantlydifferentfromone,whichrepresentstheaverage statewidemortalityexperienceunderthism odel.

Risk-adjustmentmodels3a,7a,8a,4a,9a,10a,6a,11a,and12aarebasedonModelA,whereasmodels3b,7b, 8b,4b,9b,10b,6b,11b,and12barebasedonModelB.ModelBdiffersfromModelAinthatitincludesrace, expectedprincipalsource ofpayment,sourceandtypeofadmission,andclinicalfactorsthatmayrepresenteither riskfactorsorcomplications.Tomaximizecomparability,allmodelsinclude925caseswithoutmissingvalues.

Table 14.13: Unweighted process of care characterist ics, by hospital mortality and volumecategory (including pvalues)

Process(eligiblecases) 1	Hos	spitalmor	talitycat	egory	Hospita	lvolumec	ategory
Restriction	Better	Neither	Worse	pvalue ²	High	Medium	pvalue ³
AspirinifCC P-eligible(n=809) 4							
Anytime,%	80	68	77	0.568	81	69	<0.001*
Within6hours,%	35	23	26	0.032*	31	24	0.034*
Meanhourstofirstdose	21.8	25.5	23.1	0.024*	25.4	21.1	0.655
Aspirinifeligible(n=850) 5							
Anytime,%	77	68	75	0.570	79	68	<0.001*
Within6hours,%	33	23	25	0.035*	30	24	0.037*
Meanhours tofirstdose	21.8	25.8	22.9	0.044*	25.6	20.9	0.613
ThrombolyticifCCP -eligible(n=302) ⁶							
Anytime,%	40	57	46	0.475	45	50	0.356
Within2hoursofarrival,%	31	44	31	0.806	30	40	0.115
Meanhourstofirstdose	2.2	2.2	4.9	0.672	4.8	1.7	0.469
Thrombolyticifeligible(n=381) ⁷							
Anytime,%	34	44	42	0.216	39	42	0.532
Within2hoursofarrival,%	26	35	28	0.770	27	33	0.178
Meanhourstofirstdose	2.7	2.1	4.7	0.506	4.5	1.9	0.393
IVheparinifthrombolysed(n=230)							
Within6hours,%	86	81	86	0.893	79	89	0.045*
Aspirinifthrombolysed(n=230)							
Within6hours,%	57	40	52	0.696	55	44	0.089*
HeparinifCCP -eligible(n=861) ⁸	O,	10	02	0.000	00	• • •	0.000
Anytime, %	77	59	67	0.016*	68	68	0.999
Within24hours,%	63	43	55	0.010	50	57	0.056*
Meanhourstofirstdose	17.0	23.0	18.6	0.160	22.4	16.2	0.409
PTCA(n=974)	17.0	20.0	10.0	0.100		10.2	0.100
Anytime,%	16	14	13	0.298	17	12	0.014*
Within24hours,%	7	3	4	0.034*	7	2	<0.001*
PTCAifeligible(n=700) 9	•	Ü	•	0.001	•	_	10.001
Anytime,%	18	15	17	0.681	20	13	0.011*
Within24hours,%	7	4	5	0.230	8	3	0.003*
•	,	7	0	0.230	U	3	0.003
CABG(N=974)	14	12	9	0.100*	12	12	0.921
Anytime,% Within24hours,%	3	1	1	0.100	2	1	0.921
	3		ı	0.040	_	í	0.110
CABGif eligible(n=700) 9 Anytime,%	15	14	11	0.289	14	13	0.825
Within24hours,%	3	14	11	0.289	3	13	0.825 0.083*
	3	ı	ı	0.232	3	1	0.003
PTCAorCABG(n=974)	27	25	22	0.004*	20	22	0.045*
Anytime,% Within24hours,%	27 9	25 4	22 4	0.094* 0.021*	28 9	22 3	0.045* <0.001*
vviuiii124110u15,70	9	4	4	0.021	3	ა	<0.001

Table 14.13: Unweighted process of care characteristics, by hospital mortality and volumecategory (including pvalues), continued

Process(eligiblecases) 1	Hos	spitalmor	talitycat	Hospitalvolumecategory			
Restriction	Better	Neither	Worse	pvalue ²	High	Medium	pvalue ³
PTCAorCABGifeligible(n=700) 9							
Anytime,%	30	28	27	0.417	32	25	0.024*
Within24hours,%	8	5	6	0.266	10	3	<0.001*
Swan-Ganzcatheterization,% (n=974)	22	18	13	0.004*	19	16	0.152
Coronaryangiography,%(n=974)	38	34	25	<0.001*	40	24	<0.001*
Betablockerifeligible,%(n=530)	38	40	51	0.015*	48	39	0.066*
Meanhourstoadmission(n=887)	3.4	3.0	3.2	0.817	3.3	3.1	0.887
MeanhourstofirstECG(n=841)	1.8	0.8	1.3	0.492	1.4	1.1	0.025*
MeanhourstofirstCK(n=879)	2.7	1.9	2.7	0.001*	2.7	2.2	0.003*

Statisticallysignificantatp<0.10.

Alltherapiesareascertainedexclusivelyfromtheindexorinitialhospitalizati on,exceptthatCABGandPTCAare ascertainedfromtheindexhospitalizationoranysubsequenttransferhospitalization.

For dichotomous factors, this p value represents a test of the hypothesis that there is a monotonic relationship between hospital process and risk-adjusted outcomes (i.e., Mantel-Haenzel chi square for trend, df=1). For continuous factors, this pvalue represents a test that the distributions differences hospital categories (i.e., Kruskal Wallisrank sum test).

Fordichotomousfa ctors,thispvaluerepresentsatestofthehypothesisthatthereisanassociationbetweenhospital processandvolume(i.e.,2 -tailedFisher'sexacttest). Forcontinuousfactors, thispvaluerepresentsatest that the distributions differacrosshosp italcategories (i.e., Wilcoxonranksumtest).

Exclusioncriteriaforthisanalysisincludedeathortransferonthedayofpresentation(ifthepatientisobtundedor experiencedacardiacarrestbeforeoratarrival),bleedingdiathesisorcoagulopat hy,aspirinallergy,gastrointestinal or genitourinary bleeding within the prior six months, guaiac positive or bloody stool at admission, warfarin at admission,thrombocytopeniaatadmission(plateletcountbelow100,000),anyhistoryofintracranialneop lasmor neurosurgery,chronicliverdisease,headtraumawithinthepriorsixweeks,serumcreatininegreaterthan3mg/dl, hematocritlessthan30%orhemoglobinlessthan10g/dl,andahistoryofmetastaticcancer.

Exclusioncriteriaforthisanalysi sincludeallofthoselistedaboveexceptthethresholdplateletcountisloweredto 50,000andthethresholdsforserumcreatinineandhematocritareeliminated.

Exclusioncriteriaforthisanalysisincludechestpainforlessthan30minutesormore than6hoursatpresentation, bleedingdiathesisorcoagulopathy, gastrointestinalorgenitourinarybleedingwithinthepriorsixmonths, guaiac positiveorbloodystoolatadmission, warfarinatadmission, anyhistoryofintracranialneoplasmorneurosurg ery, chronidiverdisease, headtraumaormajorsurgerywithinthepriorsixweeks, cardiopulmonaryresuscitationwithin theprior24hours, knownorsuspectedaorticaneurysm, anyhistoryofstroke, uncontrolledhypertension(systolic bloodpressuregrea terthan200mmHgordiastolicbloodpressuregreaterthan120mmHgatpresentation), age greaterthan80years, oranyotherspecifiedcontraindicationorrefusaloftherapy.

Exclusion criteria for this analysis include all of those listed above exce ptless than 30 minutes of chest pain at presentation, stroke more than six months before admission, age greater than 80 years, and other specified contraindications or refusal of the rapy.

Exclusion criteria for this analysis include bleeding diathesis or coagulopathy, guaiac positive or bloody stool at admission, warfarin at admission, thrombocytopenia at admission (platelet count below 100,000), any history of

intracranial neoplasmorneur osurgery, head trauma within the prior six weeks, and he matocrithemoglobin less than 10 g/dl).

lessthan30%(or

- Exclusioncriteria for this analysis include a gegreater than 80 years, a "do not resuscitate" or deron the date of presentation or the date of admission, and death or transferon the day of presentation (if the patient is obtunded or experienced acardia carrest before or a tarrival).
- Exclusioncriteriaforthisanalysisincludeahistoryofasthmaorchronicobstructivepulmonarydisease,diabetes mellitusrequiringinsulinatadmission,congestive heartfailureorpulmonaryedemabythefirstchestradiograph, systolicbloodpressurelessthan100mmHgatpresentation,secondorthirddegreeatrioventricularblockonthefirst orlastECGinthefirst24hours(unlessapermanentpacemakerwasin placeorinsertedduringthishospitalization), shock at any time during the hospitalization, and poor left ventricular function (ejection fraction less than 25%, shortening fraction less than 15%, or severe/very severedy sfunction).

Table14.14:Weighte dprocessofcarecharacteristics,byhospitalmortalityandvolume category¹

Process(eligiblecases) 2	State wide	<u>Hospi</u>	talmortalit	tycategory	Hospitalvolumecategory		
Restriction		Better	Neither	Worse	High	Medium	
Aspirin ifCCP -eligible(n=809) ³	70	00		70	0.4	0.4	
Anytime,%	73	86	71	79	81	64	
Within6hours,%	26	37	25	27	31	20	
Meanhourstofirstdose	24.5	26.7	24.1	24.3	22.0	28.6	
Aspirinifeligible(n=850)	70	00			00	0.4	
Anytime,%	73	83	71	77	80	64	
Within6hours,%	26	35	25	26	30	20	
Meanhourstofirstdose	24.8	26.5	24.6	24.1	22.9	27.9	
ThrombolyticifCCP -eligible(n=302) ³							
Anytime,%	51	37	54	47	40	68	
Within2hoursofarrival,%	39	26	42	32	29	54	
Meanhourstofirstdose	2.5	3.1	2.2	5.2	3.3	1.9	
Thrombolyticifeligible(n=381) ³							
Anytime,%	41	32	42	44	32	56	
Within2hoursofarrival,%	32	23	33	30	24	44	
Meanhourstofirstdose	2.5	3.2	2.2	5.0	3.1	1.9	
IVheparinifthrombolysed(n=230)							
Within6hours,%	81	87	80	86	71	90	
Aspirinifthrombolysed(n=230)							
Within6hours,%	44	56	42	53	54	36	
HeparinifCCP -eligible(n=861) ³							
Anytime,%	63	79	60	70	61	65	
Within24hours,%	46	60	43	58	43	51	
Meanhourstofirstdose	22.5	19.6	23.5	18.6	23.9	20.9	
PTCA(n=974)							
Anytime,%	17	18	17	15	20	12	
Within24hours,%	5	8	4	4	8	1	
PTCAifeligible(n=700) ³							
Anytime,%	18	21	17	18	22	12	
Within24hours,%	5	8	4	4	9	0	
CABG(N=974)							
Anytime,%	12	14	12	10	12	12	
Within24hours,%	1	3	1	1	2	0	
CABGifeligible(n=700) ³	•	Ū	•	•	_	Ü	
Anytime,%	13	15	13	12	14	13	
Within24hours,%	1	2	1	1	2	0	
PTCAorCABG(n=974)		_	•	'	_	J	
Anytime,%	28	28	28	24	31	23	
Within24hours,%	20 6	20 9	20 5	2 4 5	9	23 1	
v v iu iii 124110u13, /0	U	Э	5	5	Э	ı	

Table 14. 14: Weighted process of care characteristics, by hospital mortality and volume category¹, continued

Process(eligiblecases) 2		<u>Hospit</u>	almortality	<u>Hospitalvolumecategory</u>		
Restriction		Better	Neither	Worse	High	Medium
PTCAorCABGifeligible(n=700) ³						
Anytime,%	29	32	29	29	35	22
Within24hours,%	6	9	6	6	11	1
Swan-Ganzcatherterization,% (n=974)	37	45	37	28	41	32
Coronaryangiography,%(n=974)	37	45	37	28	41	32
Betablocker ifeligible,%(n=530) ³	43	41	42	53	44	42
Meanhourstoadmission(n=887)	3.0	3.6	2.9	3.2	3.3	2.7
MeanhourstofirstECG(n=841)	0.9	1.6	0.7	1.4	0.8	0.9
MeanhourstofirstCK(n=879)	2.1	3.0	1.9	2.7	2.4	1.7

Weightingcompensatesfortheoversamplingofoutlierhospitalsanddeaths; theweightedestimatesapproximate thetruevalueoftheseparametersamongallAMIpatientsadmittedtothissubsetofhospitalsstatewide.

 $^{^2 \}quad \text{All the rapies are ascertained excl} \quad \text{usively from the index or initial hospitalization, except that CABG and PTCA are ascertained from the index hospitalization or range usual subsequent transfer hospitalization.}$

³ These exclusion criteria are described in the notesto Table 14.13.