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Variability of insulin sensitivity during the first 4 days of critical illness: implications for tight glycemic control

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Abstract

Background: Effective tight glycemic control (TGC) can improve outcomes in critical care patients, but it is difficult to achieve consistently. Insulin sensitivity defines the metabolic balance between insulin concentration and insulin-mediated glucose disposal. Hence, variability of insulin sensitivity can cause variable glycemia. This study quantifies and compares the daily evolution of insulin sensitivity level and variability for critical care patients receiving TGC.

Methods: This is a retrospective analysis of data from the SPRINT TGC study involving patients admitted to a mixed medical-surgical ICU between August 2005 and May 2007. Only patients who commenced TGC within 12 hours of ICU admission and spent at least 24 hours on the SPRINT protocol were included (N = 164). Model-based insulin sensitivity (*SI*) was identified each hour. Absolute level and hour-to-hour percent changes in *SI* were assessed on cohort and per-patient bases. Levels and variability of *SI* were compared over time on 24-hour and 6-hour timescales for the first 4 days of ICU stay.

Results: Cohort and per-patient median *SI* levels increased by 34% and 33% ($p < 0.001$) between days 1 and 2 of ICU stay. Concomitantly, cohort and per-patient *SI* variability decreased by 32% and 36% ($p < 0.001$). For 72% of the cohort, median *SI* on day 2 was higher than on day 1. The day 1–2 results are the only clear, statistically significant trends across both analyses. Analysis of the first 24 hours using 6-hour blocks of *SI* data showed that most of the improvement in insulin sensitivity level and variability seen between days 1 and 2 occurred during the first 12–18 hours of day 1.

Conclusions: Critically ill patients have significantly lower and more variable insulin sensitivity on day 1 than later in their ICU stay and particularly during the first 12 hours. This rapid improvement is likely due to the decline of counter-regulatory hormones as the acute phase of critical illness progresses. Clinically, these results suggest that while using TGC protocols with patients during their first few days of ICU stay, extra care should be afforded. Increased measurement frequency, higher target glycemic bands, conservative insulin dosing, and modulation of carbohydrate nutrition should be considered to minimize safely the outcome glycemic variability and reduce the risk of hypoglycemia.

Keywords: Critical care, Hyperglycemia, Insulin resistance, Mathematical model, Algorithms

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Background

Safe, effective tight glycaemic control (TGC) of critically ill patients can improve outcomes [1-4], but it is difficult to achieve consistently [5-7]. Glycaemic level and variability in TGC are a function of variability in insulin sensitivity, potentially resulting from the level and evolution of the stress response [8], and are independently associated with mortality [9-12].

Insulin sensitivity defines the metabolic balance between insulin concentration and glucose disposal. Insulin-mediated glucose disposal is a dominant pathway to reduce and control glycaemia in critically ill patients. For a fixed insulin concentration, a given percentage change of insulin sensitivity results in a proportional change to glucose disposal and thus glycaemic level, all else equal.

Understanding the variability of insulin sensitivity, over hours and days, is important for safely and effectively managing glycaemic levels with exogenous insulin. Several patient- and treatment-related factors influence insulin sensitivity. Some of the influential and predictable factors (drug therapies and existing patient conditions) are taken into account when developing therapeutic algorithms for insulin treatment.

The objective of this study was to examine the evolution of insulin sensitivity level and variability over the first 4 days of intensive care unit (ICU) stay using data from the SPRINT TGC study [1]. Analyses were performed on two separate timescales, using 24-hour and 6-hour blocks of data. The impact of this insulin sensitivity evolution on glycaemia in the context of TGC protocols is considered.

Methods

Patients

This study is a retrospective analysis of patient data (N = 164 patients, 12,067 hours) from the SPRINT clinical practise change in the Christchurch Hospital ICU [1]. All patients admitted between August 2005 and May 2007 were included where the SPRINT TGC protocol was commenced within 12 hours of ICU admission and continued for at least 24 hours. All patients were treated per protocol, with no specific exclusions. Table 1 presents a summary of cohort details.

The Christchurch Hospital ICU is a 15-bed, closed, mixed medical-surgical unit led by intensive care specialists in a tertiary affiliated teaching hospital. Glycaemic control data were collected from handwritten daily ICU charts and entered into a spreadsheet database. The Upper South Regional Ethics Committee, New Zealand, granted approval for the audit, analysis, and publication of this data.

The SPRINT protocol

The SPRINT protocol (SPecialised Relative Insulin Nutrition Tables) is a simple, lookup-table system derived

Table 1 Summary details of the study subjects

N	164	
Age (yr)	65 [56-74]	t1.1
Gender (M/F)	102/62	t1.2
APACHE II score	19 [16-25]	t1.3
APACHE II ROD (%)	32 [17-52]	t1.4
Operative/nonoperative	66/98	t1.5
Hospital mortality	25%	t1.6
ICU mortality	18%	t1.7
ICU length of stay (hr)	142 [70-308]	t1.8
Diabetic history: type I/type II	10/22	t1.9
Data are presented as median [interquartile range] where appropriate.		t1.10
		t1.12

from a model-based controller that modulates both insulin and nutritional inputs. The protocol titrates insulin doses and nutrition rates to estimated patient-specific insulin sensitivity for tight glycaemic control in the range 4.0-6.1 mmol/L BG range [1,13,14]. SPRINT has been the standard of care in the Christchurch ICU since August 2005. The requirement for the patients in this study to be on the SPRINT protocol ensured that they had regular and accurate records of blood glucose levels, insulin administered, and nutrition given.

The entry criterion for the SPRINT protocol was two BG measurements >8 mmol/L during normal patient monitoring, or at the discretion of the clinician. Once on the protocol, BG was measured 1- to 2-hourly, with a median measurement interval for this cohort of 1.5 hours. BG measurements were taken by nursing staff using the Arkray Super-GlucoCard II glucometer (Arkray Inc., Japan). Blood samples tested were typically arterial, although when an arterial line was not present, capillary blood was used. Additional File 1 contains a more detailed description of SPRINT and specific, unique differences to other protocols.

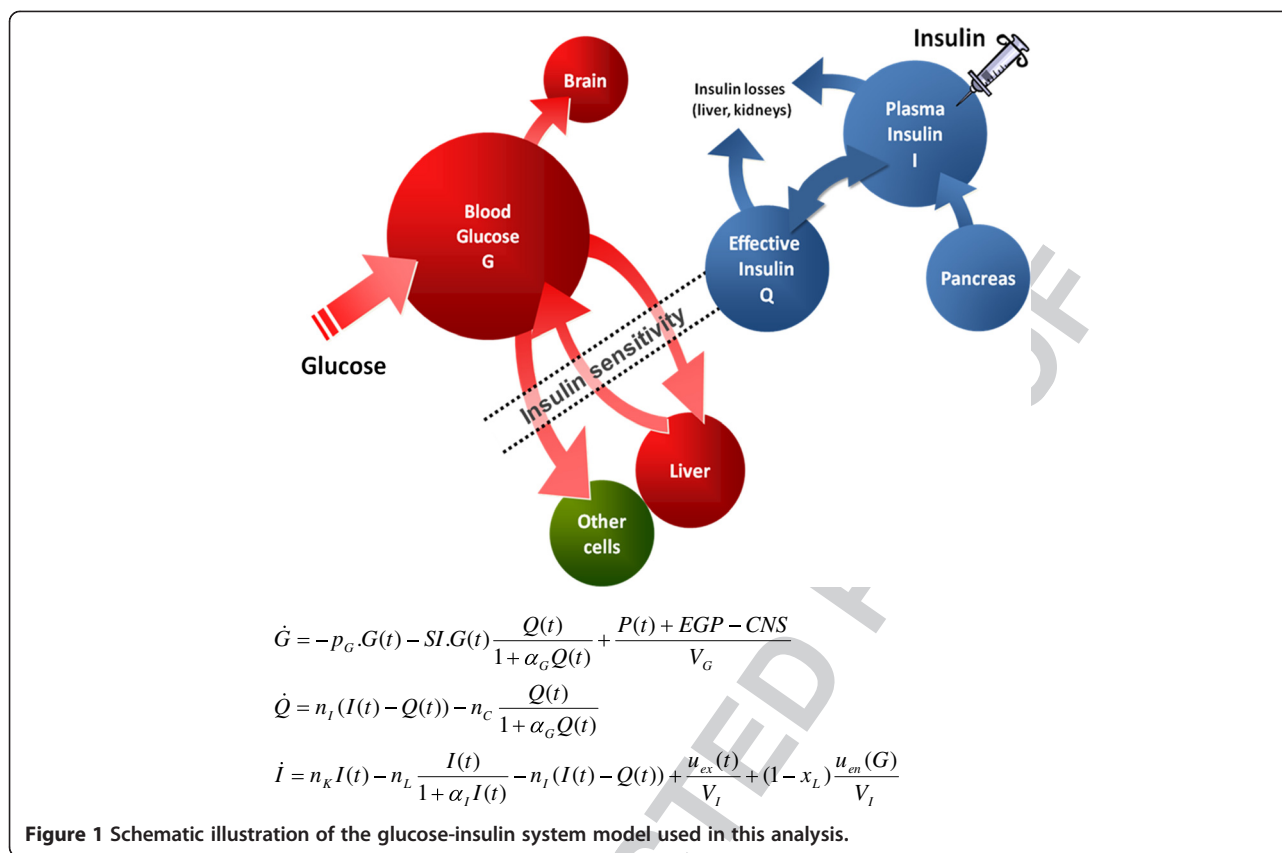
Model-based insulin sensitivity

Model-based methods provide a means of determining physiological parameters that either cannot be measured directly or are impractical to measure with the required frequency. In this study, model-based insulin sensitivity (*SI*) was identified using an integral method [15] with a validated glucose-insulin system model developed for critical care patients [16,17]. The glucose-insulin system model is illustrated schematically in Figure 1 and presented in greater detail in Additional File 2.

The *SI* parameter represents "whole-body" insulin sensitivity. The parameter defines the glycaemic response to exogenous insulin and nutrition, capturing the relative net effect of altered endogenous glucose production, peripheral and hepatic insulin mediated glucose uptake, and endogenous insulin secretion. However, this time-varying

T1

F1



121 insulin sensitivity parameter has been shown to correlate
 122 very well ($r > 0.9$) with the “gold standard” euglycemic
 123 clamp [17] and has been used to guide model-based TGC
 124 in several studies [18-20].

125 A value of SI was identified every hour [15] for each patient
 126 using clinical data and the model implemented in
 127 MATLAB (2011a, Mathworks, Natick, MA). When the
 128 BG measurement interval was greater than 1 hour, linearly
 129 interpolated values were used for identification. Variability
 130 of insulin sensitivity was calculated as the hour-to-hour
 131 percentage change in SI ($\Delta\%SI$), defined below:

$$\Delta\%SI_k = 100 \times \frac{(SI_{k+1} - SI_k)}{SI_k}$$

132 Use of percentage change in SI , rather than absolute
 133 change, normalizes the metric so that patients with very
 134 different absolute levels of SI can be compared fairly.
 135 Equally, for a fixed insulin concentration, a given percent-
 136 age change in insulin sensitivity results in a proportional
 137 change to glucose disposal and thus glycemic level, all else
 138 equal.
 139

140 Analyses

141 SI level and variability are analyzed on overall cohort
 142 and per-patient bases using two separate timescales. The

143 evolution of SI over the first 4 days of ICU stay is analyzed
 144 in 24-hour blocks. Bagshaw [12] reported an association
 145 between hypoglycemia and variability during the
 146 first 24 hours of ICU stay and mortality. We therefore
 147 also analyzed the acute evolution of SI over the first day
 148 using 6-hour blocks.

149 Cohort analysis looks at the hourly values of SI and vari-
 150 ability for the entire cohort grouped together and shows
 151 trends in the overall group behavior. To quantify per-
 152 patient variability, the interquartile range (IQR: 25th-75th
 153 percentile) of $\Delta\%SI$ is examined for each patient within
 154 each timescale. This metric captures the width of the vari-
 155 ability distribution for each patient. Per-patient SI level is
 156 defined by the median value within each timescale.

157 The analyses are linked to time on the SPRINT proto-
 158 col, rather than time in the ICU, to ensure sufficient insu-
 159 lin and nutrition data to accurately identify SI hourly [15].
 160 Hence, day 1 comprises the first 24 hours of SPRINT.
 161 However, because patients were included only if they
 162 commenced SPRINT within 12 hours of ICU admission, a
 163 minimum of half of the day 1 results for each patient
 164 occur during their first 24 hours in the ICU. The median
 165 delay between admission and commencement of SPRINT
 166 for this cohort was 1.9 hours and 81% of the cohort was
 167 on SPRINT within 6 hours. When a patient was taken off
 168 the SPRINT protocol, their SI profile for the last day was

169 included in the analysis only if it contained 6 hours or
 170 more of data.

171 *SI* levels and variability are non-Gaussian and thus
 172 compared using cumulative distribution functions
 173 (CDFs) and nonparametric statistics. Distributed data
 174 are generally compared using the Wilcoxon rank-sum
 175 test (Mann–Whitney *U* test), except for *SI* variability
 176 results. *SI* variability is compared using the Kolmogorov-
 177 Smirnov test, because it has more power to detect differ-
 178 ences in the shape of distributions than the rank-sum
 179 test when median values are similar. $P < 0.05$ are consid-
 180 ered statistically significant.

181 Results

182 Twenty-four hour analyses

183 Insulin sensitivity level

F2 184 Figure 2 presents the cumulative distribution functions
 185 (CDFs) of hourly *SI* for each day by cohort (left panel) and
 T2 186 median daily *SI* per-patient (right panel). Table 2 presents
 187 the increase in median insulin sensitivity and associated
 188 *p* values between successive days. Both per-patient and cohort
 189 analyses suggest that insulin sensitivity levels start
 190 low, but increase over time in the ICU. There is a particu-
 191 larly significant increase between days 1 and 2 ($p < 0.001$).
 192 On subsequent days the increase continues but to a lesser
 193 degree. Per-patient comparisons between days 2, 3, and 4
 194 are not statistically significant.

T3 195 The results of Figure 2 and Table 2 are further reflected
 196 in Table 3, which shows that daily median insulin sensitiv-
 197 ity increases for a large proportion of the cohort between
 198 days 1 and 2 with lesser proportions on subsequent days.
 199 Table 3 is a matrix where the value in a cell represents the

Table 2 Increasing cohort and per-patient median insulin sensitivity over time (24-hr blocks)

Level analysis	Cohort analysis		Per-patient analysis	
	% Increase at median	<i>p</i> value	% Increase at median	<i>p</i> value
Days 1-2	34	<0.0001	33	0.0004
Days 2-3	16	<0.0001	21	0.2559
Days 3-4	6	0.0013	4	0.6306

P values calculated using Wilcoxon rank-sum test.

proportion of patients for whom daily median insulin sensitivity is greater on the day of the associated column than the day of the associated row. For example, 72% of patients show an increase in median *SI* between days 1 and 2, and 54% when comparing days 2 and 3.

Insulin sensitivity variability

SI variability decreases over time in the ICU, parallel to increases in absolute *SI* level. Figure 3 and Table 4 present the CDFs and tabulated results for cohort and per-patient analyses of the hour-to-hour percentage changes in *SI* ($\Delta\%SI$). The cohort aggregate distributions of $\Delta\%SI$ by day are shown in the left panel of Figure 3. The right panel presents the CDFs for the per-patient IQRs by day.

As with insulin sensitivity level, the largest increase in *SI* variability is between days 1 and 2. The decrease between days 2, 3, and 4 is statistically significant for both cohort and per-patient analyses, but the change is much less than over the first day and may not be clinically significant.

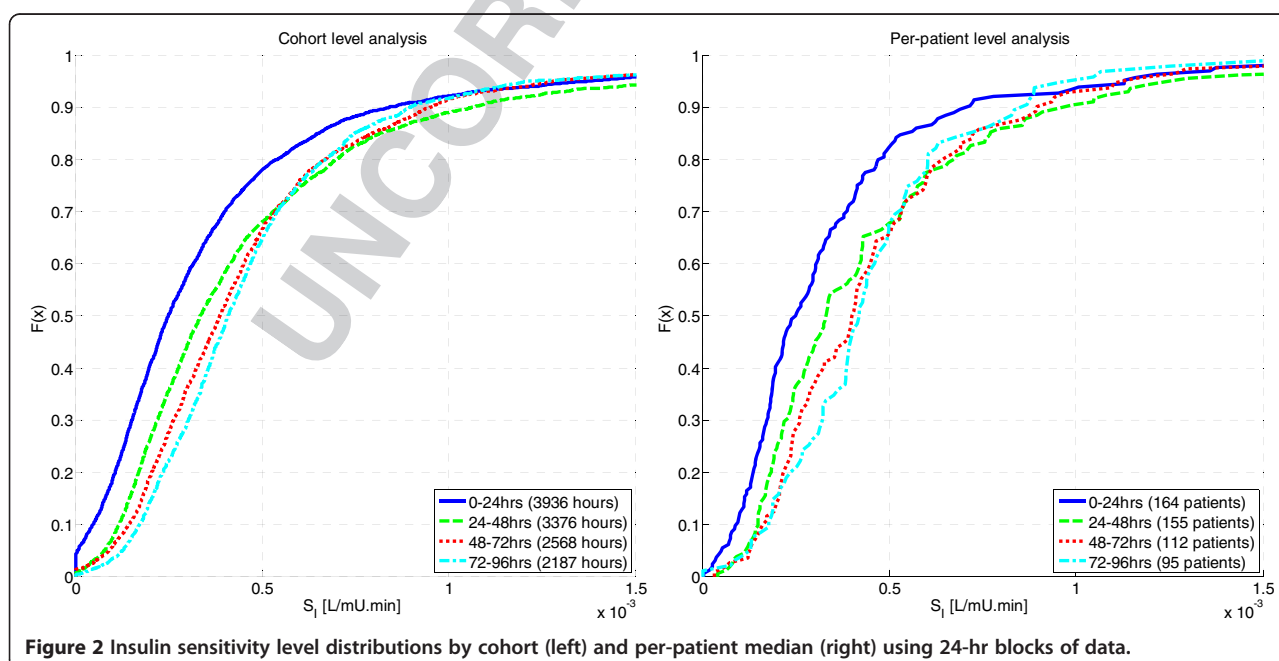


Figure 2 Insulin sensitivity level distributions by cohort (left) and per-patient median (right) using 24-hr blocks of data.

t3.1 **Table 3 Proportion of patients for whom median insulin**
 t3.2 **sensitivity increases between the days indicated in the**
 t3.3 **rows and columns**

	Day 2	Day 3	Day 4
Day 1	0.72	0.74	0.71
Day 2		0.54	0.64
Day 3			0.53

220 **Six-hour analyses**

221 *Insulin sensitivity level*

F4 222 Figure 4 presents the distributions of cohort and per-
 223 patient insulin sensitivity over the first 24 hours in 6-
 224 hour blocks. Also shown for comparison is the day 2 dis-
 225 tribution from Figure 1 (labeled 24–28 hours). It is evi-
 226 dent that the insulin sensitivity level increases over the
 227 first day up to the level of the second day. Hence, the
 228 differences between day 1 and 2 seen in Figure 2 are a
 229 function of the low, but increasing, insulin sensitivity
 230 during the first 12–18 hours.

T5 231 Table 5 lists the differences in median insulin sensitiv-
 232 ity levels from the distributions shown in Figure 4. The
 233 increases in *SI* during the first 18 hours are large and
 234 statistically significant. Subsequent increases are unlikely
 235 to be clinically significant at less than 10%. Of particular
 236 interest is the comparison between 18–24 hours and day
 237 2, which indicates that by 18 hours, the rapid increase in
 238 *SI* is largely complete.

T6 239 Table 6 shows that during the first 18 hours, a large
 240 proportion of the patients have an increase of insulin
 241 sensitivity using the 6-hour timescale. After 18 hours,

Table 4 Reductions in the interquartile range (IQR) and
median per-patient range of hour-to-hour percentage
insulin sensitivity change over time

Variability analysis	Cohort analysis		Per-patient analysis	
	% Reduction of IQR	p-value	% Decrease at median	p value
Days 1-2	32	<0.0001	36	<0.0001
Days 2-3	20	0.0028	18	0.0091
Days 3-4	14	0.0269	17	0.0369

P values calculated using Kolmogorov-Smirnov test for cohort comparisons and Wilcoxon rank-sum test for per-patient comparisons.

the proportion of patients with increasing *SI* is similar to that seen between days 2, 3, and 4 (Table 3) at slightly more than 50%.

Insulin sensitivity variability

As with absolute *SI* level, the majority of the decrease in *SI* variability occurred during the first 18 hours. Figure 5 shows the CDFs of the cohort and per-patient variability metrics. Table 7 shows that only the differences between 0–6 hours and 6–12 hours are statistically significant at the 5% level. The 6–12 vs. 12–18-hour comparison is close to statistical significance, with $p < 0.07$ for both cohort and per-patient analyses.

Discussion

Insulin sensitivity variability

Both cohort and per-patient results suggest that critically ill patients have significantly lower and more variable insulin sensitivity on day 1 than later in their ICU stay.

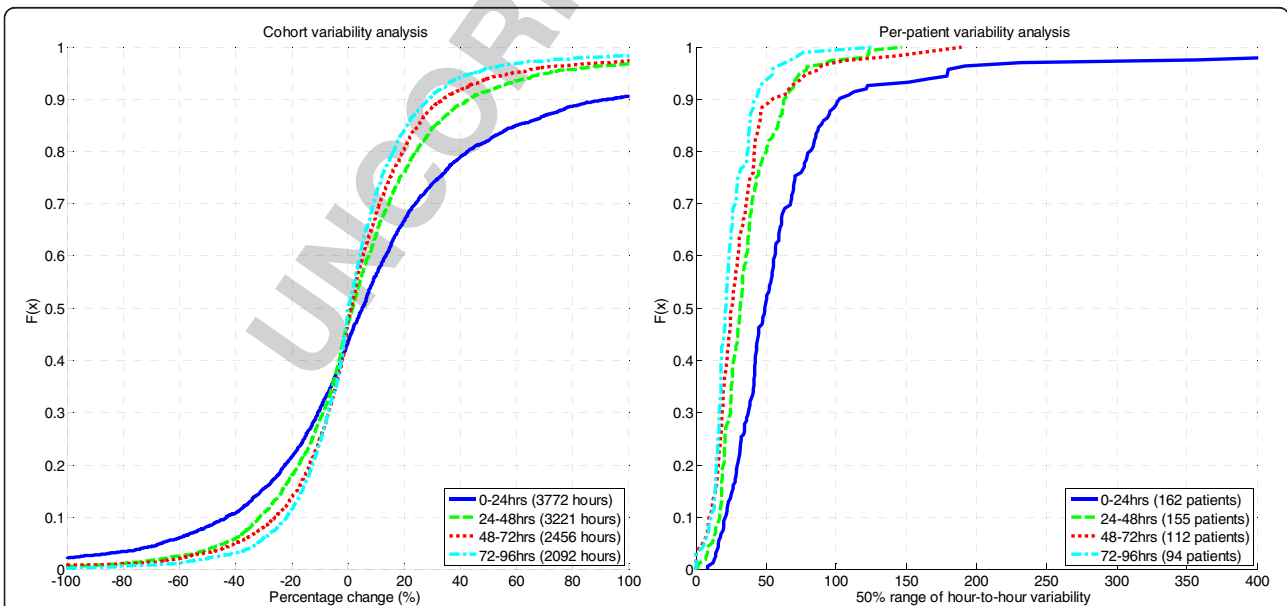


Figure 3 Insulin sensitivity variability distributions by cohort (hour-to-hour percentage change) and per-patient interquartile-range using 24 hr blocks of data.

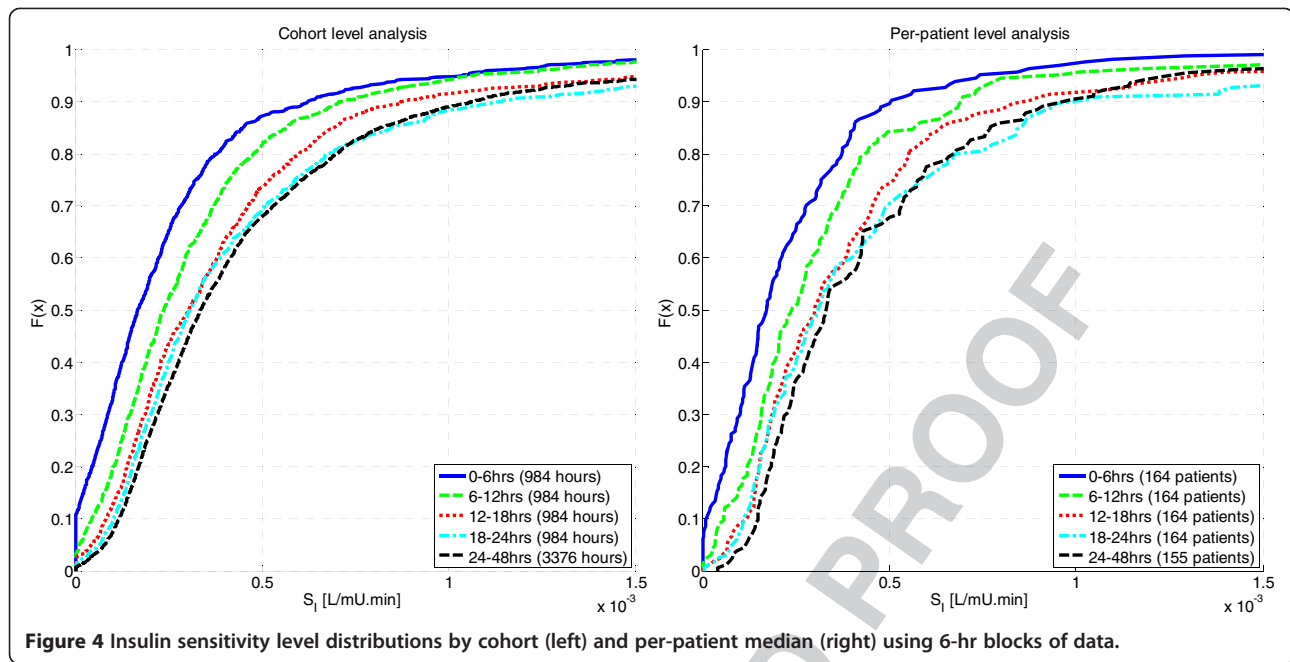


Figure 4 Insulin sensitivity level distributions by cohort (left) and per-patient median (right) using 6-hr blocks of data.

259 Further analysis shows that this day 1 result is primarily
 260 influenced by the first 12–18 hours of ICU stay. Over this
 261 time, rapid improvements in insulin sensitivity level and
 262 variability occur so that there is no statistically significant
 263 difference between 18–24 hours and day 2. From day 2
 264 onwards, changes in *SI* level and variability are not as large
 265 and of limited clinical and statistical significance.

266 Within the analyses, there are some differences in
 267 significance between cohort and per-patient results for
 268 comparisons after day 2. The overall findings noted in the
 269 preceding paragraph are the only clear, consistent trends
 270 across both analyses.

271 The counter-regulatory hormones: cortisol, glucagon,
 272 the catecholamines, as well as growth hormone are signifi-
 273 cantly elevated almost immediately after critical-insult,
 274 but decline rapidly over the first 12–48 hours [21–24].
 275 These hormones are known to cause increased hepatic

glucose production, inhibition of insulin release, and 276
 peripheral insulin resistance [22], all of which cause a 277
 decrease in the model-based *SI* metric used in this study. 278
 Hence, the low but rapidly increasing insulin sensitivity 279
 seen during the first 12–18 hours of ICU stay is likely due 280
 to the acute counter-regulatory response to critical illness. 281

Time in this study was referenced from the commencing 282
 of SPRINT, rather than ICU admission. 283
 However, the difference between admission time and 284
 commencing SPRINT was generally very short, with a 285
 median delay for this cohort of 1.9 hours. Within 6 286
 hours of admission, 81% of the cohort had commenced 287
 SPRINT. Hence, these results are applicable to the first 288
 few hours and days of ICU stay. 289

The insulin sensitivity parameter 290

The model-based parameter used in this study repre- 291
 sents a whole-body insulin sensitivity capturing overall 292
 metabolic response to exogenous insulin. *SI* captures the 293
 relative net effect of altered hepatic glucose production, 294
 peripheral and hepatic insulin-mediated glucose uptake, 295
 and endogenous insulin secretion. All of these effects 296

t5.1 **Table 5** Increasing cohort and per-patient median insulin
 t5.2 sensitivity over time (6-hr blocks)

Level analysis	Cohort analysis		Per-patient analysis	
	% Increase at median	<i>p</i> value	% Increase at median	<i>p</i> value
t5.6 Block 1–2 t5.7 (0–6 vs. 6–12 hr)	42	<0.0001	40	0.0007
t5.8 Block 2–3 t5.9 (6–12 vs. 12–18 hr)	28	<0.0001	26	0.0123
t5.10 Block 3–4 t5.11 (12–18 vs. 18–24 hr)	1	0.0335	3	0.4829
t5.12 Block 4–5 t5.13 (18–24 vs. 24–48 hr)	9	0.0452	7	0.3776

t5.14 *P* values calculated using Wilcoxon rank-sum test.

t6.1 **Table 6** Proportion of patients for whom median insulin
 t6.2 sensitivity increases between the blocks indicated in the
 t6.3 rows and columns

	6–12 hr	12–18 hr	18–24 hr	24–48 hr
t6.4 0–6 hr	0.74	0.78	0.77	0.79
t6.5 6–12 hr		0.76	0.7	0.72
t6.6 12–18 hr			0.55	0.64
t6.7 18–24 hr				0.58
t6.8				

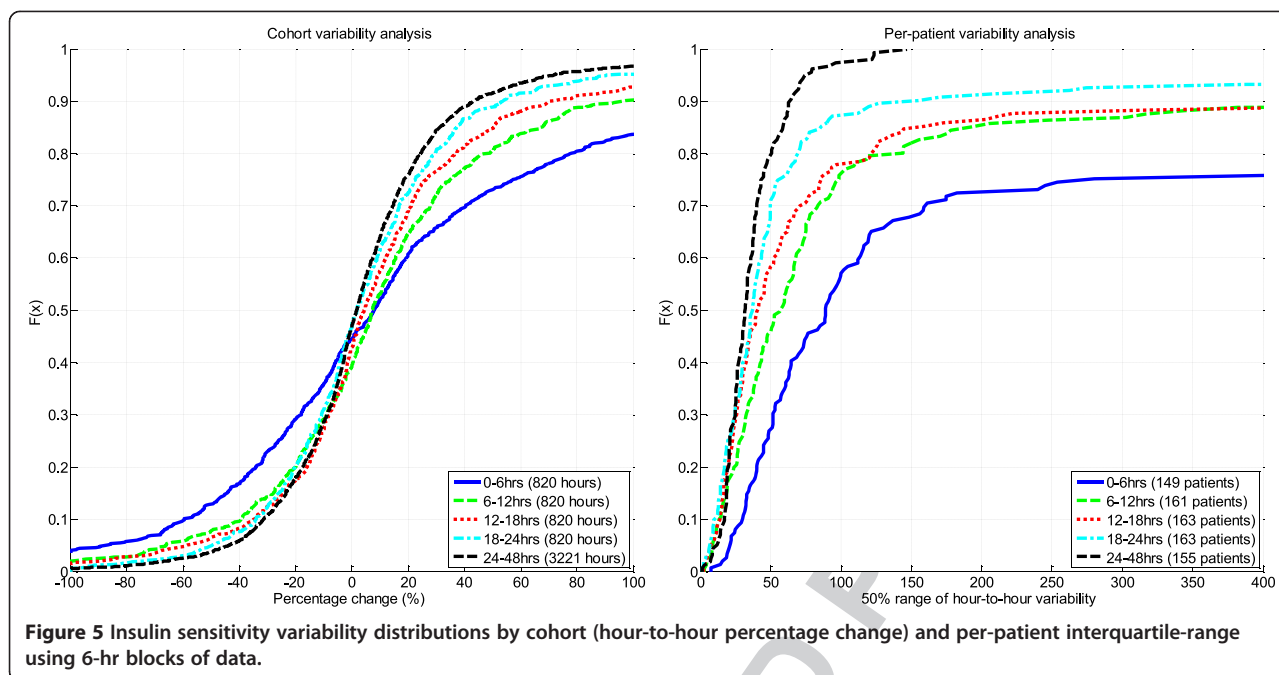


Figure 5 Insulin sensitivity variability distributions by cohort (hour-to-hour percentage change) and per-patient interquartile-range using 6-hr blocks of data.

297 are altered significantly in critical illness due to the
 298 stress response [25-27]. Hence, the metabolic balance
 299 that this parameter represents is an important consider-
 300 ation in TGC, because it determines a body's glycemic
 301 response to exogenous insulin and nutrition.

302 As an identified parameter, *SI* contains unmodeled
 303 physiological effects and measurement device noise. How-
 304 ever, Lotz et al. [17] indicated that this form of insulin
 305 sensitivity correlated very well ($r > 0.9$) with the "gold
 306 standard" euglycemic clamp and its change in a lifestyle
 307 intervention study on 73 normoglycemic healthy and
 308 obese subjects (146 clamp procedures before/after inter-
 309 vention). In the critical care setting, a similar version of
 310 the model and *SI* parameter has been cross-validated

311 against independent, matched patient data from a single
 312 center of the Glucontrol randomized, clinical trial [28].

313 The analytic inaccuracy of bedside glucometers or any
 314 other sensor used to gather BG measurements influence
 315 individual values of *SI*. However, this study examines
 316 distributions of *SI* consisting of thousands of values
 317 identified from a wide range of BG values, thus both the
 318 random and bias components of error cancel out within
 319 each distribution. This effect was confirmed by Monte
 320 Carlo analysis (results not shown) using an error model
 321 for the glucometer derived from data supplied by the
 322 manufacturer [29].

Implications for tight glycemic control

323 With low and variable insulin sensitivity, glycemic levels
 324 may appear unresponsive and/or difficult to control effec-
 325 tively with exogenous insulin. This situation may provoke
 326 larger insulin doses from many protocols that have no
 327 explicit upper limits on insulin dose [6,30-32]. High levels
 328 of circulating insulin coupled with the observed variability
 329 in insulin sensitivity result in increased glycemic variability
 330 and an increased risk of hypoglycemia during the first 24
 331 hours of ICU stay.

332 Not only does glycemic variability pose a risk through
 333 hypoglycemia, it also is detrimental in its own right. Several
 334 studies [9-11,33] have shown that glycemic variability is
 335 independently associated with mortality in critically ill
 336 patients. More specifically, Bagshaw [12] showed that
 337 hypoglycemia and variability within the first 24 hours of
 338 ICU stay are each associated with increased mortality. In
 339 vitro, high glycemic variability was shown to increase
 340

t7.1 **Table 7** Reductions of the interquartile range (IQR) and
 t7.2 median per-patient range of hour-to-hour percentage
 t7.3 insulin sensitivity change over time

t7.4 Variability t7.5 analysis	Cohort analysis		Per-patient analysis	
	% Reduction t7.6 of IQR	p value	% Decrease t7.7 at median	p value
t7.8 Block 1-2 t7.9 (0-6 vs. 6-12 hr)	40	0.0017	36	<0.0001
t7.10 Block 2-3 t7.11 (6-12 vs. 12-18 hr)	24	0.0628	28	0.0673
t7.12 Block 3-4 t7.13 (12-18 vs. 18-24 hr)	0	0.0931	9	0.1032
t7.14 Block 4-5 t7.15 (18-24 vs. 24-48 hr)	18	0.1682	14	0.1075

t7.16 P values calculated using Kolmogorov-Smirnov test for cohort comparisons
 t7.17 and Wilcoxon rank-sum test for per-patient comparisons.

341 oxidative stress [34] and apoptosis [35], thereby suggesting
342 a rationale to explain the clinical association with poor
343 outcome.

344 Evidence from other studies [10,12] indicates an associ-
345 ation between hypoglycemia, glycemic variability, and mortal-
346 ity. However, the question remains: Is low and variable
347 glycemia the cause of increased morbidity and mortality?
348 Or is it just a symptom in very ill patients? Until this ques-
349 tion can be answered conclusively, it is perhaps best to form-
350 ulate TGC protocols not to exacerbate the situation,
351 which requires the ability to differentiate more and less
352 metabolically variable patients.

353 Another significant finding in this study is the range of
354 variability seen across patients, as well as over time
355 (Figures 3 and 5). Less variable patients, if identified, may
356 be treated more aggressively with insulin without com-
357 promising glycemic variability. Hence, model-based meth-
358 ods have been mooted as a means of better managing this
359 inter- and intra-patient variability [30,36].

360 Limitations

361 Only patients on the SPRINT TGC protocol were consid-
362 ered for this analysis as they had sufficient data density to
363 identify *SI* hourly. Patients were put on the SPRINT
364 protocol because they were hyperglycemic and thus were
365 likely to be biased towards lower insulin sensitivity com-
366 pared with other ICU patients. However, in the context of
367 investigating the implications of *SI* variability on TGC,
368 this cohort is appropriate.

369 Another limitation is the use of a model-based insulin
370 sensitivity parameter, as it is not measured directly and
371 may be influenced by modelling errors or un-modelled
372 effects. As an identified parameter, *SI* contains unmo-
373 delled physiological effects and measurement device
374 noise. However, as noted previously, this form of *SI* has
375 been shown to correlate very well with the “gold stand-
376 ard” euglycemic clamp [17,37] and has been shown to
377 be an independent marker of metabolic condition [28].
378 Finally, this method of analysis is robust to BG sensor
379 error.

380 A further limitation is the relatively small cohort size
381 available for analysis. The demands of manually tran-
382 scribing written clinical data into electronic form and
383 the specific inclusion criteria have restricted the number
384 of patients for whom complete glycemic control data are
385 currently available for analysis. The size of this cohort
386 has precluded subgroup analyses, such as diabetic and
387 cardiovascular surgery patients, because these subgroups
388 only contain 20–40 patients. With relatively few
389 patients, the subgroup analyses fail to demonstrate stat-
390 istical significance, despite effect sizes and trends very
391 similar to that seen in this overall analysis. Thus, these
392 comparisons will be completed in the future, when more
393 patient data become available.

The findings of this study should be equally valid in 394
other ICUs where attention to TGC and blood glucose 395
measurement frequency may be a lower priority. Al- 396
though the data density might not be present to allow 397
such units to explicitly identify *SI* hourly, these results 398
indicate that patients will still have lower and more vari- 399
able insulin sensitivity on day 1 than later in their ICU 400
stay. Thus, suggestions of higher glycemic targets, con- 401
servative insulin dosing, and modulation of carbohydrate 402
nutrition are especially pertinent. 403

Without the ability to identify patient-specific metabolic 404
states, a protocol should be less aggressive over the first 405
few days, and particularly the first 24 hours, to minimize 406
variability. It may be important for protocols to consider 407
higher glycemic targets on the first days of ICU stay (com- 408
pared with later days) to ensure safety. Perhaps a glycemic 409
target similar to the current guidelines of 7.8–11 mmol/L 410
[38–40] is most appropriate for the first 24 hours with the 411
target range, reducing over days 2 and 3 to more normo- 412
glycemic levels as *SI* level and variability improve. 413

Greater blood glucose measurement frequency and 414
conservative insulin dosing can mitigate the impact of *SI* 415
variability on risk [41] and also should be considered for 416
the first few days of stay. Modulation of carbohydrate 417
nutrition, within limits [42], can reduce the need for ex- 418
ogenous insulin to better manage glycemia [43]. 419

420 Conclusions

The results of this study indicate that critically ill patients 421
have significantly lower and more variable insulin sensitivity 422
on day 1 than later in their ICU stay, particularly during the 423
first 12–18 hours. This effect is likely due to the acute 424
counter-regulatory response to critical illness. Greater vari- 425
ability with lower *SI* early in a patient’s stay greatly increases 426
the insulin required, potential glucose flux due to variation 427
in *SI*, and thus the risk of greater glycemic variability and 428
hypoglycemia. Both glycemic variability and hypoglycemia 429
have been associated with poor outcomes in the ICU. 430

Clinically, these results suggest that TGC patients re- 431
quire greater care over the first few days of ICU stay to 432
minimize safely the outcome glycemic variability. It may 433
be important for protocols to consider higher glycemic 434
targets on the first days of ICU stay to ensure safety. 435
Equally, greater measurement frequency, conservative 436
insulin dosing, and modulation of carbohydrate nutrition 437
can mitigate the impact of variability on risk and should 438
be considered for the first few days of stay. 439

440 Additional files

Additional file 1: Detailed description of the SPRINT protocol,
listing unique features and differences to other TGC protocols.

Additional file 2: Detailed description of the glucose-insulin system
model and the *SI* parameter [19,43–74].

441
442
443
444

445 Abbreviations

446 ICU: Intensive care unit; SPRINT: Specialized relative insulin and nutrition
447 titration; TGC: Tight glycemic control; *S_I*: Insulin sensitivity metric
448 (model-based); $\Delta\%S_I$: Hour-to-hour percentage changes in insulin sensitivity;
449 CDF: Cumulative distribution function; IQR: Interquartile range;
450 APACHE: Acute physiology and chronic health evaluation; KS:
451 Kolmogorov-Smirnov (test).

452 Competing interests

453 The authors declare that they have no competing interests.

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470 Authors' contributions

471 JGC, GS, and ALC conceived and developed the SPRINT protocol. GS
472 implemented the protocol with staff at Christchurch Hospital. CGP, ALC, JGC,
473 GS, TD, J-CP, and SP assisted with the data analysis, idea generation, some
474 (or all) data collection, and/or the analysis and interpretation of the data
475 and/or statistical analysis. CGP, JGC, and ALC drafted the manuscript
476 primarily, although all of the authors made contributions. All authors read
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