

University of Warwick institutional repository

This paper is made available online in accordance with publisher policies. Please scroll down to view the document itself. Please refer to the repository record for this item and our policy information available from the repository home page for further information.

To see the final version of this paper please visit the publisher's website. Access to the published version may require a subscription.

Authors:	C. Nendel, U. Schmutz, A. Venezia, F. Piro and C. R. Rahn
Article title:	Converting simulated total dry matter to fresh marketable yield for field vegetables at a range of nitrogen supply levels
Year of publication:	2009
Link to published version:	http://dx.doi.org/10.1007/s11104-009-0015-0
Publisher statement:	The original publication is available at www.springerlink.com

1 Converting simulated total dry matter to fresh marketable yield for field vegetables at a range
2 of nitrogen supply levels

3

4 C. Nendel · U. Schmutz · A. Venezia · F. Piro · C.R. Rahn

5

6 C. Nendel (✉)

7 Institute for Vegetable and Ornamental Crops,

8 Theodor-Echtermeyer-Weg 1, 14979 Großbeeren, Germany

9 Present address: Leibniz-Centre for Agricultural Landscape Research, Institute for Landscape

10 System Analysis,

11 Eberswalder Straße 84, 15374 Müncheberg, Germany

12 e-mail: nendel@zalf.de

13

14 U. Schmutz

15 Henry Doubleday Research Association,

16 Ryton Organic Gardens, Coventry, CV8 3LG, UK

17

18 A. Venezia · F. Piro

19 Centro di Ricerca per l'Orticoltura,

20 Via dei Cavalleggeri 25, Casella Postale 48,

21 84098 Pontecagnano, Italy

22

23 C.R. Rahn

24 Warwick HRI,

25 Wellesbourne, CV35 9EF, UK

26

27 Abstract

28 Simultaneous analysis of economic and environmental performance of horticultural crop
29 production requires qualified assumptions on the effect of management options, and
30 particularly of nitrogen (N) fertilisation, on the net returns of the farm. Dynamic soil-plant-
31 environment simulation models for agro-ecosystems are frequently applied to predict crop
32 yield, generally as dry matter per area, and the environmental impact of production. Economic
33 analysis requires conversion of yields to fresh marketable weight, which is not easy to
34 calculate for vegetables, since different species have different properties and special market
35 requirements. Furthermore, the marketable part of many vegetables is dependent on N
36 availability during growth, which may lead to complete crop failure under sub-optimal N
37 supply in tightly calculated N fertiliser regimes or low-input systems. In this paper we present
38 two methods for converting simulated total dry matter to marketable fresh matter yield for
39 various vegetables and European growth conditions, taking into consideration the effect of N
40 supply: (i) a regression based function for vegetables sold as bulk or bunching ware and (ii) a
41 population approach for piecewise sold row crops. For both methods, to be used in the context
42 of a dynamic simulation model, parameter values were compiled from a literature survey.
43 Implemented in such a model, both algorithms were tested against experimental field data,
44 yielding an Index of Agreement of 0.80 for the regression strategy and 0.90 for the population
45 strategy. Furthermore, the population strategy was capable of reflecting rather well the effect
46 of crop spacing on yield and the effect of N supply on product grading.

47

48 Key words

49 Dry matter yield; marketable yield; vegetables; model; nitrogen

50

51 Introduction

52 For the simultaneous analysis of economic and environmental performance of agricultural
53 production different paths have been followed, each having their own advantages and
54 drawbacks. Sectoral status analyses often ignore interactions on the process level or
55 summarise them using purely economical indicators (e.g. Galdeano-Gomez 2008). Farm level
56 approaches often use static models (e.g. Bateman et al. 1999; Cembali et al. 2007; Münier et
57 al. 2004), which in few cases are based on previously obtained results of dynamic ones (e.g.
58 Pacini et al. 2004). If environmental practices in agriculture are assessed for their profitability,
59 the farmer's management options in the field are often not considered (e.g. Managi and
60 Karemera 2005). This is applicable to evaluate the performance of the respective economic
61 sector. However, it does not provide feedback relations to (i) support the farmer's choice for
62 market channels, crops and fertiliser management or (ii) to indicate directions for the
63 development of codes for Good Agricultural Practice. Performance evaluation of vegetable
64 production inevitably requires knowledge on the effect of N fertiliser on yield and the farm's
65 net return. As the system is complex, dynamic soil-plant-environment simulation models for
66 agro-ecosystems are often applied to predict crop yield and the environmental impact of
67 production (Kersebaum et al. 2007). However, only simple models with a limited
68 performance range have been available for field vegetable production systems up until now
69 (Fink and Scharpf 1993; Greenwood 2001; Rahn et al. 1996).

70 Decisions on estimating N fertiliser applications are mainly governed by two factors: at policy
71 level, environmental issues often determine legislative acts to control N fertiliser use in
72 agriculture, whereas at the farm level decisions are governed by economic considerations.
73 Both factors are included in a variety of models in different ways, depending on the scale of
74 the simulation. Agricultural sector models have been combined with dynamic nutrient
75 leaching models and used at national (Lehtonen et al. 2007) or watershed level (Faeth et al.
76 1991; Ribaud et al. 2001; Schou et al. 2000). However, at field level, more details have to be
77 considered and for this reason, soil-plant-environment models are often equipped with

78 additional modules which allow simultaneous evaluations of the economical and
79 environmental impact of crop management decisions (e.g. Hughes et al. 1995; Lindgren and
80 Elmquist 2005; Rejesus and Hornbaker 1999; Vatn et al. 1999). In most cases, the simulated
81 total above-ground crop dry matter (TDM) is used to calculate economic returns, which is
82 feasible for many agricultural systems.

83 Crop growth models that produce TDM as an output can generally be categorised as process-
84 based models driven by photosynthesis and empirical models that often use alternative drivers
85 (Marcelis et al. 1998). The objective behind functional process models is to explain crop
86 growth using the most important feedback regulation mechanisms. Most commonly, the leaf
87 area index development of the crop is simulated according to its ontogenesis, providing a
88 framework for photosynthesis to be calculated from radiation (Wang et al. 2002). Assimilates
89 produced from this process are then distributed – either actively or driven by sink demand – to
90 the different plant organs. This approach enables us to explicitly calculate the mass of the
91 crop's marketable part, and renders a conversion step obsolete if the dry matter content is
92 known or calculable. Numerous process models have been developed for all kinds of
93 vegetable crops (Marcelis et al. 1998). However, a generic, process-based crop growth model
94 for a large number of field-grown vegetables has not yet been presented. Such a task is
95 hampered by the fact that field vegetables have a broad range of morphologies, including
96 those for which a leaf area index in terms of a photosynthetic active surface is difficult to
97 determine (cabbages, lettuces, leek, etc.). A canopy approach has previously been applied for
98 such crops (Gutierrez et al. 1994; van Henten 1994). The EU-Rotate_N model (Rahn et al.
99 2007) requires a target yield to describe the growth of field vegetables. This approach restricts
100 the simulation of the crop's response to radiation and heat, limiting the model application to
101 situations in which a rough idea of the expected yield exists. However, using this approach in
102 combination with the critical N concept (Greenwood et al. 1986; Plenet and Lemaire 1999),
103 nitrogen uptake and fate can be simulated generically for most vegetables.

104 The use of TDM as a basis for economic calculations leads to considerable problems for field
105 vegetable production systems. First, the relationship between TDM and marketable fresh
106 weight yield (MFY) is highly variable due to a large number of species (Marcelis et al. 1998)
107 and secondly, some vegetables can produce large amounts of unmarketable material that is
108 not harvested and is reincorporated as crop residues into the field. In addition, the TDM-MFY
109 relationship is often dependent on the N supply to the crop but unfortunately it does not
110 always follow simple optimum curves as i.e. for cereals. This is due to the fact that some
111 crops do not produce any marketable parts when the N supply drops below a certain threshold
112 (Rather and Schenk 2005; Thompson et al. 2000). Excessive N levels can also lead to the
113 marketable part of some vegetable crops becoming unmarketable, developing hollow stems,
114 fuzzy curds, black midribs or other disorders (Berard 1990; Scaife and Wurr 1990).

115 The implementation of the EU nitrates directive (The Council of the European Communities
116 1991) has reduced the amount of N being applied to agriculture and resulted in N
117 management in crop production becoming increasingly complex. Eventually this may result in
118 vegetables receiving sub-optimal nutrient supply under unfavourable weather conditions.
119 Consequently, decision support systems will need to support the possible effects of sub-
120 optimal N supply on crop growth and thus on farm net returns. This is especially important
121 when modelling low-input production systems such as organic vegetable farming, where sub-
122 optimal N conditions occur frequently. In these systems, crop spacing is used to adjust for N
123 utilization and yield.

124 This paper presents two strategies to overcome the specific problems of yield conversion for
125 field vegetable production, with some results of their use in a model aimed at supporting
126 environmentally sound vegetable production planning across Europe. Each strategy has its
127 own advantages and disadvantages, which will be discussed.

128

129

130 Material and Methods

131 Yield data

132 Marketable and total dry matter yield data obtained from experiments with different N
133 fertiliser levels were extracted from previously unpublished field vegetable experiments
134 undertaken as part of the EU-Rotate_N project (www.warwick.ac.uk/go/eurotaten) or during
135 the default work programme of regional research stations. The research stations located at
136 Großbeeren (Germany), Hanover (Germany), Kise (Norway), Landvik (Norway), Piikkiö
137 (Finland), Pontecagnano (Italy), Ryton (UK), Valencia (Spain), Wannweil (Switzerland),
138 Wellesbourne (UK) and Årslev (Denmark) provided experimental data on various field
139 vegetables, including a range of currently grown cultivars. Additional monitoring data from
140 managed farm sites were collected at Bobenheim, Böbingen, Kleinniedesheim, Lustadt and
141 Zeiskam (all Germany). All of the experiments include measurements of the soil mineral
142 nitrogen content at sowing or planting time, which is used to roughly estimate the total N
143 supply to the crop by adding it to the fertiliser amount applied during the course of the
144 experiment. For this method, it is assumed that no leaching of N out of the rooting zone
145 occurs. Furthermore, all of the experiments include measurements of crop fresh marketable
146 yield and above-ground dry matter at harvest. The fertiliser treatments cover a range of N
147 supply in which profitable vegetable production commonly occurs. Almost all of the
148 experiments include a zero fertiliser treatment, which enables crop production to be assessed
149 in low fertiliser regimes that meet the requirements of EU legislation. All in all, the levels of
150 detail for the various crops differ quite considerably. For many crops, detailed field
151 experiments from various countries, including multiple N treatments, are available. For some
152 crops, only single field experiments or published field research designed for other research
153 questions where the raw data could be used for our purpose are available. For the remainder,
154 expert knowledge was used to obtain a preliminary value for the model. Here, further research
155 and experimental data is required to back up the estimate. In those cases, crop parameters

156 were defined based on experimental results for similar species (parsnip, broad beans, spring
157 onion). A summary of the data used for algorithm development and parameterisation is given
158 in Table 1.

159

160

161 Strategy I: Regression approach

162 A feasible method of converting simulated above-ground total dry matter (TDM) into
163 marketable fresh weight yield (MFY), for which prices are available, is the use of a
164 conversion factor. This method follows the idea of the harvest index, which is used to
165 estimate grain yield from total crop biomass (Hay 1995). For most field vegetables, the
166 relationship between TDM and MFY is a function of the N supply to the crop.

$$167 \hat{m}_{\text{MFY}}(N_{\text{av}}) = f(N_{\text{av}}) \cdot \hat{m}_{\text{TDM}}(N_{\text{av}}) \quad (\text{Equation 1})$$

168 with

169 $\hat{m}_{\text{MFY}}(N_{\text{av}})$ = marketable fresh weight yield estimator for specific level of available N

170 $\hat{m}_{\text{TDM}}(N_{\text{av}})$ = simulated above-ground dry matter for specific level of available N

171 $f(N_{\text{av}})$ = conversion function

172 N_{av} = plant-available nitrogen in soil

173

174 From the empirical data three different types of N supply dependent TDM-MFY relationships
175 were classified according to best-fit of various trend and regression types (Table 2): Type A
176 includes all vegetable crops for which the conversion function is linear (Figure 1A). Type B
177 describes vegetable crops that have a linear ratio only at an optimum region but different
178 ratios under sub- and supra-optimum conditions (Figure 1B). Finally, type C covers all
179 vegetables which show no TDM-MFY relation (Figure 1C).

180

181 ((Table 2))

182

183 ((Figure 1))

184

185 For all type A and B vegetables the conversion function can be described with a polynomial
186 function of the type

$$187 f(N_{av}) = r_0 + r_1 \cdot N_{av} + r_2 \cdot N_{av}^2 \quad (\text{Equation 2})$$

188 with r_0 , r_1 and r_2 being crop specific parameters. Parameter values were estimated for all
189 vegetables for which sufficient data was available from the literature survey or from own
190 experiments (data not shown), using SigmaPlot 5.0 (SPSS Inc., Chicago, IL, USA). Parameter
191 sets for those vegetables to be converted following Strategy I, are compiled in Table 3. In
192 these particular cases, r_2 equals 0.

193 ((Table 3))

194

195 Strategy II: Population approach

196 The regression approach has a number of drawbacks: it does not consider that (i) different
197 crop spacing can affect the marketable yield as the absolute number of plants changes, (ii)
198 some vegetables do not produce any marketable yield when N supply is restricted and (iii)
199 very small or very large produce may be rejected by the consumer and will not be offered for
200 sale. Consequently a more complex approach is presented as an alternative strategy for crops
201 with a single marketable product: this population approach is based on the concept that crop
202 yield follows a normal distribution in respect of the size of the marketable part. A given
203 coefficient of variation describes the size distribution within the population. Minimum and
204 maximum size boundaries determine the fraction of the population's marketable produce that
205 is suitable for marketing. Identifying non-marketable produce allows the simulation model to
206 treat this biomass as crop residue to be recycled in the system.

207 In a first step, dry matter yield (DMY) is calculated from simulated TDM using the crop
 208 specific harvest index HI as

$$209 \quad \hat{m}_{DMY}(N_{av}) = \hat{m}_{TDM}(N_{av}) \cdot HI(N_{av}) \quad (\text{Equation 3}),$$

210 describing the fraction of TDM that is commonly harvested and thus not left in the field as
 211 crop residue. DMY is now transferred into a single-plant fresh matter yield (PFY) by

$$212 \quad \hat{m}_{PFY}(N_{av}) = \frac{\hat{m}_{DMY}(N_{av})}{n \cdot c_{DM}} \quad (\text{Equation 4}),$$

213 with

214 n = number of plants per area unit, calculated from row and planting distances

215 c_{DM} = typical dry matter concentration of the harvested fraction

216

217 The size of the harvested part (i.e. curds, heads, etc.) is normally distributed and can thus be

218 described using $\hat{m}_{PFY}(N_{av})$ as mean fresh weight of a single plant product at harvest with a

219 given coefficient of variation (CV , standard deviation related to mean). A lower market limit

220 (L_{low}), describes the minimum weight of a product that can be sold at market and subsequently

221 a grade-out probability p_G can be calculated for the population.

$$222 \quad p_G(N_{av}) = \frac{1}{\sqrt{2\pi\bar{s}}} \cdot e^{-\left(\frac{L_{low} - \hat{m}_{PFY}(N_{av})}{2\bar{s}^2}\right)} \quad (\text{Equation 5})$$

223 with

224 $\bar{s} = \hat{m}_{PFY}(N_{av}) \cdot CV$ being the standard deviation of the sizes within the population of

225 harvested plant parts.

226

227 In the same way an upper market limit (L_{up}) is established which is used to grade out

228 harvested produce that is too big to be marketable. MFY is then calculated from PFY,

229 corrected by the grade-out probability.

$$230 \quad \hat{m}_{FMY}(N_{av}) = n \cdot \hat{m}_{PFY}(N_{av}) \cdot (1 - p_G(N_{av})) \quad (\text{Equation 6})$$

231 This approach requires the following crop specific data: row spacing, plant spacing, c_{DM} , L_{low} ,
232 L_{up} , CV , and HI , all of which need to be determined experimentally or acquired from expert
233 knowledge. Since only very limited empirical data on CV values were found on most crops,
234 the average CV of 0.3 was used for all crops, except for carrots. Parameter sets for those
235 vegetables to be converted following Strategy II, are compiled in Table 4.

236

237 ((Table 4))

238 The EU-Rotate_N simulation model

239 The EU-Rotate_N model (Rahn et al. 2007) is based on a dynamic process-based simulation
240 of the crop-soil-environment interaction in field vegetable production systems. N movement
241 in soil is driven by water balance and transport according to a capacity approach (Ritchie
242 1998), where the water content at saturation, field capacity and wilting point define the
243 hydraulic soil properties. Crop growth simulation follows a simple target yield approach, with
244 the daily growth increment being linearly reduced in case of water or N deficiency. Nitrogen
245 mineralisation from organic matter is based on the routines used in the DAISY model
246 (Hansen et al. 1991). Crop residues are assigned a dynamic C to N ratio, which reflects the
247 growth conditions of the crop with respect to N supply (Jensen et al. 2005).

248 Root growth is calculated using a heat sum approach and distributed spatially in a 2D soil cell
249 grid, allowing for the simulation of spacing effects in row crops. Crop and soil-specific
250 rooting depth enables the simulation of deep and shallow rooted crops, and their characteristic
251 N exploitation from the soil. N uptake is calculated as a function of crop N demand, which in
252 turn follows the critical N concept of Greenwood (2001). Fertility-building crops are
253 simulated using fixed growth rates and parameters for litter loss, N fixation processes and
254 winter kill.

255 An economic analysis of the simulation results is based on costs related (planting, base
256 fertilisers, irrigation, crop protection, weed control, hail insurance) and not related (packing,

257 drying, transport, commission, labour, fuel) to yield, costs for N fertilisers and applications
258 and on prices for marketable parts of the crop, which are stored in the model's database. In
259 order to link the TDM model output to the MFY prices, the algorithms for both conversion
260 strategies are embedded in the EU-Rotate_N model. An example economic analysis was
261 presented by Nendel (2009). The simulations presented below are conducted with the EU-
262 Rotate_N simulation model.

263

264 Testing the algorithms

265 *Simulation of marketable yield in a crop rotation experiment in Italy*

266 The two conversion strategies were tested against a crop rotation experiment carried out on
267 the CRA-ORT experimental station at Pontecagnano, Italy (40°38' N, 14°52' O). The
268 experiment has been conducted with four two-yearly rotations of four vegetable crops, each
269 grown at three nitrogen levels and two times per year (spring-summer versus autumn-winter
270 seasons) until completion of four cropping cycles. The crops included broccoli, cabbage,
271 spinach, lettuce and fennel. Broccoli did not prove sufficiently robust to withstand cold
272 season winds during the first crop series and was thereafter substituted with cabbage, which
273 was used from the second to the fourth series. The nitrogen fertilizer levels were based on
274 average farmer's practice and applied as 100% and 130%, respectively. A zero fertilizer
275 treatment completed the experimental design.

276 A split-plot field layout was used, with the rotations in main plots and the N rates in subplots,
277 with two replicates in adjacent blocks. Among others, observations included above-ground
278 TDM and MFY. The experiment started in November 2003 and was completed by July 2005.
279 The EU-Rotate_N model was applied to reproduce the rotations, using the observed yields set
280 as target yields. By default, the model displays MFY as the results of both conversion
281 strategies.

282 Both conversions were compared against the observed MFY data for one replicate of the
283 experiment, including 16 crops at three N levels, by computing the limits of agreement, a
284 predictive interval for the difference between simulated and observed yield for a new crop
285 among the tested species, following Bland and Altman (1986). The outcome p_i for a new crop
286 i yielded by two alternative estimation methods m (1, 2) can be modelled as:

$$287 \quad p_{mi} = \alpha_m + \mu_i + e_{mi}, \quad e_{mi} \sim N(0, \sigma_m^2) \quad (\text{Equation 7})$$

288 The differences $d_i = p_{1i} - p_{2i}$ are identically distributed with mean $\alpha_1 - \alpha_2$ and variance $\sigma_1^2 +$
289 σ_2^2 , so the 95% prediction interval for a new difference is $\alpha_1 - \alpha_2 \pm \left(1.96 + \sqrt{(\sigma_1^2 + \sigma_2^2)}\right)$.

290 Besides this numerical summary, plotting the differences d_i versus the averages \bar{p}_i evidences
291 the extent of the agreement between the two methods and the types of departure from it.

292 Results are displayed by plotting means versus differences (Bland-Altman plot).

293

294 *Simulation of marketable yield in a plant spacing experiment in Germany*

295 The Strategy II algorithm's ability to predict the effects of different plant spacings on
296 individual product weight was tested against data from a spacing experiment with white
297 cabbage (Variety "Quisto"), carried out at Großbeeren, Germany (52°20' N, 13°19' O), in

298 1995. Applying the EU-Rotate_N model, fertilisation and irrigation were triggered

299 automatically to ensure that simulated growth was not restricted by either water or N

300 deficiency. Local weather data from Potsdam (52°23' N, 13°03' O) was used for the

301 simulation. The target yield parameter which drives dry matter accumulation in the EU-

302 Rotate_N model was calibrated against the total dry matter measurements from the

303 experimental data using the least root mean squared error (RMSE) as a fitting criterion.

304 Within the simulation the effect of crop spacing on crop dry matter was controlled by a 2D

305 root model algorithm, which takes competing water and nutrient requirements between plants

306 into consideration. The calculation of MFY (head fresh weight) and crop residues (leaf fresh
307 weight) occurs at harvest time (September 27, 1995) in the modelling procedure.

308

309 *Simulation of product grading in an N fertiliser experiment in Germany*

310 Individual products being too big or too small to meet market requirements are commonly
311 graded out and not offered to the consumer (gradeout procedure). To demonstrate the ability
312 of the Strategy II algorithms to reproduce product gradeout, an N fertiliser experiment with
313 cauliflower (Variety “Fremont”) carried out at Großbeeren, Germany in 1996, was simulated
314 using the EU-Rotate_N model. Simulated N treatments and irrigation matched the original
315 field experiment and marketable yield gradeout was calculated at harvest time (August 1,
316 1996). Gradeout parameters were $\hat{m}_{PFY}(N_{av})$, a default CV of 0.3 and a default HI of 0.45.

317 L_{low} for gradeout in simulation and experiment was set to an individual head weight of 600g;

318 L_{up} was not set.

319

320

321 Results

322 Simulation of marketable yield in a crop rotation experiment in Italy

323 MFY as converted by the model is compared to field data for the regression (Figure 2A) and
324 the population approach (Figure 2B). While the simulations for most vegetables produce
325 almost identical results for both methods, MFY of fennel was widely underestimated with the
326 regression method, indicating that the regression method is not applicable for fennel (Figure
327 3).

328

329

330 ((Figure 2))

331 ((Figure 3))

332

333 Simulation of marketable yield in a plant spacing experiment in Germany

334 The results of the simulation for head and leaf fresh matter weight, total N content and TDM
335 are shown in Figure 4. Since the simulated crop was well supplied, both spacing variants were
336 simulated with little variation in TDM development or N content (Figure 4, C and D). The
337 differences in yield simulation as displayed in Figures 4 A and B are based on the alteration
338 of the model's row distance variable only.

339

340 ((Figure 4))

341

342 Simulation of product grading in a N fertiliser experiment in Germany

343 The simulation results for three fertiliser scenarios (Figure 5, D – F) were compared to the
344 gradeout rates of the harvested heads in the experiment (Figure 5, A – C). The model was able
345 to reflect the general trend observed in the experimental data, in which lower N supply
346 yielded smaller heads and a larger fraction of heads graded too small. The large difference
347 between observed and simulated gradeout within the 120 kg N ha⁻¹ treatment is caused by
348 three individuals being only marginally lighter than L_{low} .

349

350 ((Figure 5))

351

352

353 Discussion

354 The comparison of both conversion methods against field data from Italy reveals that for most
355 of the tested crops both methods yield similar results. However, for fennel the regression
356 approach did not work sufficiently well, spoiling the over-all performance evaluation of this
357 method (Figure 2A). Here, the pattern (Figure 3) indicates that in this example a different

358 parameterisation would lead to a better result. As demonstrated for fennel, the two alternative
359 approaches can not be applied simultaneously to all vegetables. From Figure 1 C, it is obvious
360 that type C classified vegetables can not be described using a polynomial conversion
361 function, since the relationship is poor. Type C crops almost solely produce marketable parts
362 which are sold piecewise to the consumer. A crop that is sold by fruit, head, bulb or curd is
363 strongly dependent on quality parameters like head weight or diameter, colour, shape, and
364 appearance. Judging their market value by fresh weight alone is inappropriate. Consequently,
365 the regression approach is not applicable for type C vegetables, especially for crops that
366 require multiple harvests (i.e. aubergine, courgette). Also in the A and B categories we find
367 crops that are often sold piecewise to the consumer, (i.e. cauliflower, red and white cabbage,
368 celeriac). Although the regression approach would most likely give reasonable results for
369 those crops, the application of the population approach would be preferred for its added
370 advantages.

371 Most vegetable crops are grown in rows with set distances between the individual plants, so
372 altering the planting distance within or between the rows can affect the growth of the
373 individual plant considerably (Csizinszky 1996; Falzari et al. 2006; Ferrari et al. 2008; Jett et
374 al. 1995). The main reason for lowering plant density is to increase the supply of nutrients,
375 water and radiation to the individual plants (Francescangeli et al. 2006; Hussaini et al. 2000).

376 Applying the population approach means that the simulated TDM gives greater individual
377 head weights if the number of individuals per area is small (i.e. increased row or planting
378 distance), which is consistent with field observations. For industry production (i.e. canning,
379 processing, frozen foods) the crops are mostly grown at higher spacing to achieve large units,
380 while for direct or whole sale market channels smaller units are desired so tighter spacing is
381 used (e.g. Jett et al. 1995). How this can be implemented in the context of a crop growth
382 simulation model was shown with white cabbage as an example (Figure 2).

383 The major advantage of the population approach is the ability to consider product gradings. In
384 practice, produce that is too light or too small is graded out and left in the field along with the
385 crop residues. The same fate often awaits produce that is too heavy or large. In extreme
386 situations a whole crop could be graded unmarketable, due to lack of N, even though heads,
387 curds, bulbs or fruits have been produced (Figure 5 C). Equally, over-supply of N can cause
388 problems such as bolting, resulting in unsaleable products (i.e. lettuce, cauliflower, fennel).
389 Figure 5 illustrates the ability of the algorithms to reflect these relations using cauliflower as
390 an example, which could be even better if a higher number of individuals was available.
391 In the EU-Rotate_N model, all vegetables are allocated to one of the three categories which
392 describe the relationship between TDM and MFY. In general, two categories of crops will be
393 converted via the regression approach (Table 3): (i) crops that produce more than one
394 marketable part per plant (i.e. beans, tomatoes, cereals, maize) and (ii) small crops whose
395 planting density is rarely varied and that are normally sold as bulk or bunching ware (i.e.
396 small radish, beetroot, spinach). All other crops share some common production techniques
397 which can vary according to the desired market channel (i.e. carrot: narrow space → bunching
398 carrots for direct or pre-pack sale; wide space → large carrots grown for storage or
399 processing). These crops will be converted using the population approach (Table 4). Crops for
400 which no data were available to support the direct conversion approach will also be converted
401 this way (bell/sweet pepper, courgette/zucchini and eggplant/aubergine). In the case of some
402 root crops (beetroot, turnip, swede, sugar beet and radish), Strategy I proved the better
403 method, since here spacing variations are of minor practical importance and parameters to
404 support Strategy II are not well founded from literature and experiments.

405

406

407 Conclusion

408 The vast number of species used for vegetable production and the broad range of their
409 properties makes it difficult to apply a generic approach for economic analysis of scenarios
410 obtained from process simulations for vegetable growth and yield. The algorithms presented
411 here constitute a key link between process simulations of water and nutrient dynamics in
412 agro-ecosystems and the calculation of net returns on marketable produce, which in turn form
413 the basis of any economic evaluation of management strategies in vegetable production.
414 Integration of the algorithms into the dynamic, process-based simulation model for
415 horticultural crop rotations for European horticulture, EU-Rotate_N, enables the model to
416 simultaneously assess both ecological and economic consequences of different nitrogen
417 management practices in field vegetable production, providing valuable information for farm
418 managers and policy makers on different scales. Along with existing decision support
419 approaches for horticulture: NDICEA (Koopmans and Bokhorst 2002; van der Burgt et al.
420 2006), ORGPLAN (Padel 2002), and FBC (Cuttle 2006), it presents a major contribution to
421 the best available practise in modelling horticultural production systems and offers a range of
422 possibilities beyond the power of economic status analyses.

423

424

425 Acknowledgements

426 The authors gratefully acknowledge the provision of white cabbage and cauliflower data by
427 Carmen Feller of IGZ Großbeeren, Germany. The work received EU funding within the
428 project QLRT-2002-01100 - Development of a model based decision support system to
429 optimise nitrogen use in horticultural crop rotations across Europe (EU-ROTATE_N), co-
430 ordinated by Warwick HRI, UK (C.R. Rahn).

431

432 References

- 433 Andersen PC, Rhoads FM, Olson SM, Hill KD (1999) Carbon and nitrogen budgets in spring
434 and fall tomato crops. *Hortscience* 34, 648-652.
- 435 Bateman IJ, Ennew C, Lovett AA, Rayner AJ (1999) Modelling and mapping agricultural
436 output values using farm specific details and environmental databases. *J. Agric. Econ.* 50,
437 488-511.
- 438 Berard LS (1990) Effects of nitrogen-fertilization on stored cabbage .1. Development of
439 physiological disorders on tolerant and susceptible cultivars. *J. Hort. Sci.* 65, 289-296.
- 440 Bianco VV, Elia G, De Palma E (1996) Dosi di azoto, scarducciatura, epoca di raccolta,
441 produzione e qualità del carciofo. *Atti III Giornate scientifiche SOI* 481-482.
- 442 Bland JM and Altman DG (1986) Statistical methods for assessing agreement between two
443 methods of clinical measurement. *Lancet* 1, 307-310.
- 444 Cembali T, Folwell RJ, Huffaker RG, McCluskey JJ, Wandschneider PR (2007) Economics
445 of alternative simulated manual asparagus harvesting strategies. *Agric. Syst.* 92, 266-294.
- 446 Colauzzi M, Calzolari P, Cuter M, Bonomi L, Schiavi M (2003) Optimisation of water and
447 nitrogen in potato cultivation. *L' Informatore Agrario* 59, 37-42.
- 448 Csizinszky AA (1996) Optimum planting time, plant spacing, and nitrogen and potassium
449 rates to maximize yield of green cauliflower. *Hortscience* 31, 930-933.
- 450 Cuttle SP (2006) Development of the FBC model to estimate the nitrogen available from
451 fertility-building crops in organic rotations. *Asp. Appl. Biol.* 79, 259-262.
- 452 Damato G, Manolio G, Bianco VV (1998) Sowing dates, nitrogen rates, pruning and yield of
453 *Lagenaria siceraria* (Molina) Standl. in southern Italy. *Acta Hort.* 467, 295-303.
- 454 Djurovka M, Markovic V, Ilin Z (1997) The effect of nitrogen fertilizer on the dry matter
455 content and mineral elements in radish. *Acta Hort.* 462, 139-144.
- 456 Elia A, Paolicelli F, Bianco VV (1991) Effect of sowing date, plant density and nitrogen
457 fertilizer on artichoke (*Cynara scolymus* L.): preliminary results. *Advances in Horticultural*
458 *Science* 5, 119-122.
- 459 Evers AM, Ketoja E, Hagg M, Plaami S, Hakkinen U, Pessala R (1997) Decreased nitrogen
460 rates and irrigation effect on celery yield and internal quality. *Plant Foods for Human*
461 *Nutrition* 51, 173-186.
- 462 Faeth P, Repetto R, Kroll K, Dai Q, Helmets G (1991) Paying the farm bill: U.S. Agricultural
463 policy and the transition to sustainable agriculture. World Resources Institute, Washington
464 D.C. pp. 71.
- 465 Falzari LM, Menary RC, Dragar VA (2006) Optimum stand density for maximum essential
466 oil yield in commercial fennel crops. *Hortscience* 41, 646-650.
- 467 Ferrari S, Furlani E, Ferrari JV, Santos ML, dos Santos DMA (2008) Development and yield
468 of the cotton plant under different row spacings and growth regulator application. *Acta*
469 *Scientiarum-Agronomy* 30, 365-371.

- 470 Fink M and Scharpf H (1993) N-Expert - a decision support system for vegetable fertilization
471 in the field. *Acta Hort.* 67-74.
- 472 Foti S, Mauromicale G, Ierna A (2005) Response of seed-grown globe artichoke to different
473 levels of nitrogen fertilization and water supplies. *Acta Hort.* 681, 237-242.
- 474 Francescangeli N, Sangiacomo MA, Marti H (2006) Effects of plant density in broccoli on
475 yield and radiation use efficiency. *Sci. Hort.* 110, 135-143.
- 476 Galdeano-Gomez E (2008) Does an endogenous relationship exist between environmental and
477 economic performance? A resource-based view on the horticultural sector. *Environ. Resource*
478 *Econ.* 40, 73-89.
- 479 Gaviola S, Lipinski V, Nijensohn L (1998) Response of onions for drying to fertilization.
480 *Ciencia del Suelo* 16, 119-121.
- 481 Greenwood DJ (2001) Modeling N-response of field vegetable crops grown under diverse
482 conditions with N_ABLE: A review. *J. Plant Nutr.* 24, 1799-1815.
- 483 Greenwood DJ, Neeteson JJ, Draycott A (1986) Quantitative relationships for the dependence
484 of growth-rate of arable crops on their nitrogen-content, dry-weight and aerial environment.
485 *Plant Soil* 91, 281-301.
- 486 Gutierrez AP, Mariot EJ, Cure JR, Riddle CSW, Ellis CK, Villacorta AM (1994) A model of
487 bean (*Phaseolus vulgaris* L.) growth types I-III - Factors affecting yield. *Agric. Syst.* 44, 35-
488 63.
- 489 Hansen S, Jensen HE, Nielsen NE, Svendsen H (1991) Simulation of nitrogen dynamics and
490 biomass production in winter-wheat using the Danish simulation-model DAISY. *Fert. Res.*
491 27, 245-259.
- 492 Hay RKM (1995) Harvest Index - A review of its use in plant breeding and crop physiology.
493 *Ann. Apl. Biol.* 126, 197-216.
- 494 Hughes D, Butcher W, Jaradat A, Penaranda W (1995) Economic analysis of the long-term
495 consequences of farming practices in the barley cropping area of Jordan. *Agric. Syst.* 47, 39-
496 58.
- 497 Hussaini MA, Amans EB, Ramalan AA (2000) Yield, bulb size distribution, and storability of
498 onion (*Allium cepa* L.) under different levels of N fertilization and irrigation regime. *Trop.*
499 *Agric.* 77, 145-149.
- 500 Inam A (2002) Effect of nitrogen application on the marketable yield of turnip. *Adv. Plant*
501 *Sci.* 15, 641-643.
- 502 Jensen LS, Salo T, Palmason F, Breland TA, Henriksen TM, Stenberg B, Pedersen A,
503 Lundström C, Esala M (2005) Influence of biochemical quality on C and N mineralisation
504 from a broad variety of plant materials in soil. *Plant Soil* 273, 307-326.
- 505 Jett LW, Morse RD, Odell CR (1995) Plant-density effects on single-head broccoli
506 production. *Hortscience* 30, 50-52.

- 507 Kersebaum KC, Hecker J-M, Mirschel W, Wegehenkel M (2007) Modelling water and
508 nutrient dynamics in soil-crop systems: a comparison of simulation models applied on
509 common data sets. In *Modelling water and nutrient dynamics in soil crop systems*. Eds. K C
510 Kersebaum, J-M Hecker, W Mirschel and M Wegehenkel. pp 1-17. Springer, Stuttgart.
- 511 Kirnak H, Kaya C, Higgs D, Tas I (2003) Responses of drip irrigated bell pepper to water
512 stress and different nitrogen levels with or without mulch cover. *J. Plant Nutr.* 26, 263-277.
- 513 Kirnak H, Tas I, Kaya C, Higgs D (2002) Effects of deficit irrigation on growth, yield, and
514 fruit quality of eggplant under semi-arid conditions. *Austr. J. Agric. Res.* 53, 1367-1373.
- 515 Koopmans CJ and Bokhorst J (2002) Nitrogen mineralisation in organic farming systems: a
516 test of the NDICEA model. *Agronomie* 22, 855-862.
- 517 Lehtonen H, Barlund I, Tattari S, Hilden M (2007) Combining dynamic economic analysis
518 and environmental impact modelling: Addressing uncertainty and complexity of agricultural
519 development. *Environ. Mod. Software* 22, 710-718.
- 520 Lindgren U and Elmquist H (2005) Environmental and economic impacts of decision-making
521 at an arable farm: An integrative modeling approach. *Ambio* 34, 393-401.
- 522 Managi S and Karemera D (2005) Trade and environmental damage in US agriculture. *World*
523 *Rev. Sci. Technol. Sust. Dev.* 2, 168-190.
- 524 Marcelis LFM, Heuvelink E, Goudriaan J (1998) Modelling biomass production and yield of
525 horticultural crops: a review. *Sci. Hort.* 74, 83-111.
- 526 Münier B, Birr-Pedersen K, Schou JS (2004) Combined ecological and economic modelling
527 in agricultural land use scenarios. *Ecol. Mod.* 174, 5-18.
- 528 Nendel C (2009) Evaluation of Best Management Practises for N fertilisation in regional field
529 vegetable production with a small scale simulation model. *Eur. J. Agron.* 30, 110-118.
- 530 Pacini C, Wossink A, Giesen G, Huirne R (2004) Ecological-economic modelling to support
531 multi-objective policy making: a farming systems approach implemented for Tuscany. *Agr.*
532 *Ecosyst. Environ.* 102, 349-364.
- 533 Padel S (2002) Development of software to plan conversion to organic production (OrgPlan).
534 In *UK Organic Research 2002 - Proceedings of the COR Conference*. Ed. J Powell. pp 169-
535 172. Aberystwyth.
- 536 Parisi M, Giordano I, Pentangelo A, Villari G (2006) Effects of different levels of nitrogen
537 fertilization on yield and fruit quality in processing tomato. *Acta Hort.* 700, 129-132.
- 538 Pimpini F, Filippini MF, Sambo P, Gianquinto G, Lazzarin R (2002) Effect of fertilisation on
539 yield of red chicory "Rosso di Chioggia" and "Rosso di Treviso" grown in two different
540 environments. *Riv. Agron.* 36, 89-97.
- 541 Pimpini F, Filippini MF, Sambo P, Gianquinto G, Lazzarin R (2000) Fertilization effects on
542 nitrate content in two types of red chicory. *Riv. Agron.* 34, 406-418.

- 543 Plenet D and Lemaire G (1999) Relationships between dynamics of nitrogen uptake and dry
544 matter accumulation in maize crops. Determination of critical N concentration. *Plant Soil* 216,
545 65-82.
- 546 Rahn CR, Greenwood DJ, Draycott A (1996) Prediction of nitrogen fertiliser requirement
547 with HRI WELL_N computer model. In *Progress in Nitrogen Cycling (Proceedings of the 8th*
548 *Nitrogen Fixation Workshop, Ghent, 5-8 September 1994)*. Eds. O van Cleemput, G Hofman
549 and A Vermoesen. pp 255-258. Kluwer, Dordrecht.
- 550 Rahn CR, Zang K, Lillywhite RD, Ramos C, De Paz JM, Doltra J, Riley H, Fink M, Nendel
551 C, Thorup-Kristensen K, Pedersen A, Piro F, Venezia A, Firth C, Schmutz U, Rayns F,
552 Strohmeyer K (2007) Using the EU-Rotate_N model to forecast the effects of nitrate
553 legislation on the economic output and environmental benefits in crop rotations. In *Mineral*
554 *versus organic fertilization - Conflict or synergism?* Eds. S De Neve, J Salomez, A van den
555 Bossche, S Haneklaus, O van Cleemput, G Hofman and E Schnug. pp 433-439. International
556 Scientific Centre of Fertilizers, Braunschweig, Budapest, Vienna.
- 557 Rather K and Schenk M (2005) Nitrogen and curd compactness of cauliflower (*Brassica*
558 *oleracea* var. *botrytis*) F-1-hybrid. *Eur. J. Hort. Sci.* 70, 60-66.
- 559 Rejesus RM and Hornbaker RH (1999) Economic and environmental evaluation of alternative
560 pollution-reducing nitrogen management practices in central Illinois. *Agr. Ecosyst. Environ.*
561 75, 41-53.
- 562 Resende FV, Oliveira P, Souza R (2000) Growth, yield and nitrogen uptake in garlic
563 produced by tissue culture, cultivated under high nitrogen levels. *Hort. Bras.* 18, 31-36.
- 564 Ribaudo MO, Heimlich R, Claassen R, Peters M (2001) Least-cost management of nonpoint
565 source pollution: source reduction versus interception strategies for controlling nitrogen loss
566 in the Mississippi Basin. *Ecol. Econ.* 37, 183-197.
- 567 Ritchie JT (1998) Soil water balance and plant water stress. Eds. G Y Tsuji, G Hoogenboom
568 and P K Thornton. pp 41-54. Kluwer Academic Publisher, Dordrecht.
- 569 Rosati A, Escobar-Gutierrez A, Burns I (2002) First attempt to simulate the response of
570 aubergine crops to N supply: a means to optimise N fertilisation. *Acta Hort.* 571, 137-142.
- 571 Rosen CJ and Tong CBS (2001) Yield, dry matter partitioning, and storage quality of
572 hardneck garlic as affected by soil amendments and scape removal. *Hortscience* 36, 1235-
573 1239.
- 574 Rumpel J (1998) Effect of long-term organic, mineral, and combined organic-mineral
575 fertilization on yield of onions (*Allium cepa* L.) grown from seeds. *Biuletyn Warzywniczy* 48,
576 5-15.
- 577 Rumpel J and Kaniszewski S (1994) Influence of nitrogen fertilization on yield and nitrate
578 nitrogen content of turnip-rooted parsley. *Acta Hort.* 371, 413-419.
- 579 Santos BM, Morales-Payan JP, Stall WM, Bewick TA (1998) Influence of purple nutsedge
580 (*Cyperus rotundus*) density and nitrogen rate on radish (*Raphanus sativus*) yield. *Weed Sci.*
581 46, 661-664.

- 582 Scaife A and Wurr DCE (1990) Effects of nitrogen and irrigation on hollow stem of
583 cauliflower (*Brassica oleracea* var. *botrytis*). *J. Hort. Sci.* 65, 25-29.
- 584 Scaziota B, Marco G, Palchetti E, Rocca F, Vecchio V (2002) How to distribute N in out-of-
585 season potato crops. *L' Informatore Agrario* 58, 63-65.
- 586 Schou JS, Skop E, Jensen JD (2000) Integrated agri-environmental modelling: A cost-
587 effectiveness analysis of two nitrogen tax instruments in the Vejle Fjord watershed, Denmark.
588 *J. Environ. Manage.* 58, 199-212.
- 589 Temperini O, Colla G, Saccardo F, Brancaleone M (2000) Fertilizer application and choice of
590 variety as the basis for improving celery yield. *Informatore Agrario* 56, 95-98.
- 591 The Council of the European Communities . Council directive of 12 December 1991
592 concerning the protection of waters against pollution caused by nitrates from agricultural
593 sources (91/676/EEC). L 375. (1991). Brussels. Official Journal.
- 594 Thompson TL, Doerge TA, Godin RE (2000) Nitrogen and water interactions in subsurface
595 drip-irrigated cauliflower: 1. Plant response. *Soil Sci. Soc. Am. J.* 64, 406-411.
- 596 Ugur A, Bozokalfa M, Esiyok D (2004) Effects of harvest stage and nitrogen doses on yield
597 and quality of endive (*Cichorium endivia* L.). *Ege Universitesi Ziraat Fakultesi Dergisi* 41, 1-
598 8.
- 599 van der Burgt GJHM, Oomen GJM, Habets ASJ, Rossing WAH (2006) The NDICEA model,
600 a tool to improve nitrogen use efficiency in cropping systems. *Nutr. Cycl. Agroecosys.* 74,
601 275-294.
- 602 van Henten EJ (1994) Validation of a dynamic lettuce growth model for greenhouse climate
603 control. *Agric. Syst.* 45, 55-72.
- 604 Vatn A, Bakken L, Botterweg P, Romstad E (1999) ECECMOD: an interdisciplinary
605 modelling system for analyzing nutrient and soil losses from agriculture. *Ecol. Econ.* 30, 189-
606 205.
- 607 Wang E, Robertson MJ, Hammer GL, Carberry PS, Holzworth D, Meinke H, Chapman SC,
608 Hargreaves JNG, Huth NI, McLean G (2002) Development of a generic crop model template
609 in the cropping system model APSIM. *Eur. J. Agron.* 18, 121-140.
610
611

612 Table 1: Summary of field experiments on the total aboveground and marketable yields of vegetables over a range of nitrogen supply rates.

613 Literature resources are printed in *italic*, the remainder are unpublished field experiments. FMY = farm yard manure.

Crop name	Number of experiments	Countries covered	Number of nitrogen treatments	Range of rootzone nitrogen supply rates	Resources
Artichoke	3	Italy	13	49 – 651	<i>Bianco et al. 1996; Elia et al. 1991; Foti et al. 2005</i>
Beetroot	9	Germany	44	102 – 403	Hanover 1984-98; Großbeeren 1998
Bell/Sweet pepper	1	Turkey	3	70 – 210	<i>Kirnak et al. 2003</i>
Broccoli	15	Germany, Norway	57	60 – 462	Hanover 1981-98; Großbeeren 1996, 2001; Kise 1999-2001
Brussels Sprouts	8	Germany	31	225 – 508	Hannover 1981-83; Großbeeren 1999
Carrot	13	Denmark, Germany, UK	80	45 – 204	Årslev 1988-90; Hanover 1986-91; Wellesbourne 1996
Carrot (Industry)	4	Germany	16	63 – 353	Großbeeren 1993
Cauliflower	15	Germany, Norway, UK	76	41 – 456	Hanover 1979-94; Großbeeren 1996, 2001; Wellesbourne 1993; Kise 1995-2001
Celeriac	2	Germany	4	108 – 260	Hanover 1988; Lustadt 2004
Celery	7	Finland, Germany, Italy	37	61 – 341	Hanover 1980-96; <i>Evers et al. 1997; Temperini et al. 2000</i>
Chinese cabbage	9	Germany	41	130 – 408	Hanover 1985-97
Courgette/Zucchini	1	UK	1	1.3 t ha ⁻¹ FMY	Ryton 2003
Cucumber (Pickling)	3	Finland	3	140 – 151	Piikkiö 2001-03
Eggplant/Aubergine	2	Italy, Turkey	6	95 – 450	<i>Kirnak et al. 2002; Rosati et al. 2002</i>
Endive	1	Turkey	4	0 – 180	<i>Ugur et al. 2004</i>
Fennel	4	Italy	12	73 – 847	Pontecagnano 2003-05
French Beans	3	Germany	18	45 – 300	Hanover 1979-94
Garlic	1	Brazil	5	0 – 250	<i>Resende et al. 2000; Rosen and Tong 2001</i>
Kale	3	Germany	10	50 – 300	Hanover 1993-98

Kohlrabi	16	Germany	63	37 – 320	Hanover 1987-94; Großbeeren 1994-95;
Lamb's lettuce	5	Germany	26	37 – 581	Großbeeren 2004-05, Zeiskam 2003; Kleinniedesheim 2003
Leek	18	Denmark, Germany, UK	99	44 – 451	Hanover 1979-98; Wellesbourne 1996; Årslev 1988-90
Lettuce (Butterhead)	11	Germany	50	27 – 232	Hanover 1980-88; Großbeeren 1994; 1999
Lettuce (Crisp)	9	Germany	38	24 – 263	Hanover 1982-97
Onion	8	Argentina, Germany, Norway, Poland, Spain	42	11 – 200	Hanover 2002-03; Großbeeren 2000; Kise 1990-96, Valencia 1999; <i>Gaviola et al. 1998; Rumpel 1998</i>
Parsley	3	Germany, Poland, Switzerland	10	28 – 229	Kleinniedesheim 2004, Wannweil 1987, <i>Rumpel and Kaniszewski 1994</i>
Pea	2	Germany	2	58 – 100	Großbeeren 1996; Kleinniedesheim 2003
Potato (Early)	3	Germany, Norway	33	30 – 210	Kise 1993-94, Bobenheim 2004
Potato	4	Italy, Spain	10	276 – 449	Valencia 1999; <i>Colauzzi et al. 2003; Scaziota et al. 2002</i>
Radicchio	3	Germany, Italy	17	49 – 405	Böbingen 2003; <i>Pimpini et al. 2000; Pimpini et al. 2002</i>
Radish	3	Germany, Serbia, USA	8	0 – 274	Lustadt 2003; <i>Djurovka et al. 1997; Santos et al. 1998</i>
Red cabbage	3	Germany	8	46 – 428	Hannover 1996-98
Savoy cabbage	5	Germany	29	110 – 447	Hannover 1988-90
Small radish (Spring)	8	Germany	8	120 – 394	Großbeeren 1998-2001; Zeiskam 2003
Small radish (Summer)	13	Germany	13	99 – 636	Großbeeren 1998-2001; Zeiskam 2003-04
Spinach	4	Germany	17	41 – 429	Hanover 1993-94; Großbeeren 1996
Squash	1	Italy	6	0 – 300	<i>Damato et al. 1998</i>
Swede	1	Norway	5	30 – 190	Kise 1995-96
Tomato (Processing)	2	Italy, USA	10	0 – 269	<i>Andersen et al. 1999; Parisi et al. 2006</i>
Turnip	4	Germany, India	22	6 – 400	Hanover 2002; <i>Inam 2002</i>
White cabbage	9	Denmark, Germany, UK	48	50 – 467	Årslev 1988-89; Hanover 1984-87; Wellesbourne 1996
White cabbage (Summer)	4	Norway	44	13 – 393	Landvik 1990-91
White cabbage (Winter)	4	Norway	44	13 – 393	Landvik 1990-91
White cabbage (Industry)	3	Germany, UK	9	165 – 455	Großbeeren 1994

614 Table 2: Classification of vegetable crops into different types of relation between total dry
 615 matter content and marketable fresh matter yield.

Type A crops (linear)	
Apiaceae	Carrots, Parsley, Parsnips, Celeriac
Brassicaceae	Broccoli, Cauliflower, Brussels sprouts, Turnip, Radish
Chenopodiaceae	Beetroot
Fabaceae	Broad and French bean, Cowpea
Liliaceae	Garlic
Poaceae	Sweet corn
Solanaceae	Main and Early potatoes
Type B crops (curvilinear or linear only at optimum region)	
Apiaceae	Carrots (Industry), Celery
Brassicaceae	Kohlrabi, Kale, Red cabbage, Savoy cabbage, White cabbage, Swedes
Chenopodiaceae	Spinach
Liliaceae	Leek, Onions
Type C crops (not related, multi harvest, perennial)	
Asteraceae	Butterhead and Iceberg lettuce, Artichoke
Brassicaceae	Chinese cabbage
Cucurbitaceae	Courgettes, Cucumber, Melon, Squash, Pumpkin
Fabaceae	Runner bean
Solanaceae	Aubergine, Tomato,

616

617

618 Table 3: Empirical parameters for vegetables and main agricultural crops which total above-
619 ground dry matter is to be converted to marketable fresh matter yield using the regression
620 approach (Strategy I).

Crop name	r_0	r_1		r_0	r_1
Artichoke	2.1	0	Parsley	8.3	0
Barley	0.61	0	Peas	1.4	0
Beetroot	5.45	0	Potato (Early)	2.36	-0.0014
Broccoli	3.5	0	Potato	10	0
Broccoli	2.5508	-0.0026	Radish	9	0
Brussels sprouts	1.9516	-0.0015	Rye/Triticale	0.6	0
Cucumber (Pickling)	13.7	0	Spinach	10.189	0.0172
French beans	4	0	Spring onion	9	0
Kale	5	0	Sugar beet	3.04	0
Lamb's lettuce	10.6	0	Swede	10.88	-0.026
Maize (CCM)	1.2	0	Sweet corn	0.7	0
Maize (Silage)	3	0	Tomato (Processing)	9.429	0.0115
Maize (Grain)	1	0	Turnip	5.84	0
Oat	0.61	0	Wheat	0.61	0
Oil seed rape	0.38	0			

621

622 Table 4: Empirical parameters for vegetables and main agricultural crops which total above-
623 ground dry matter is to be converted to marketable fresh matter yield using the population
624 approach (Strategy II).

Crop name	Dry matter content (c_{DM}) %	Harvest index (HI) %	Minimum fresh weight (L_{low}) g	Maximum fresh weight (L_{up}) g	Population coefficient of variance (CV)
Bell/Sweet pepper	10.0	70	90		0.3
Carrot	12.0	83	50	250	0.5
Cauliflower	10.0	45	600		0.3
Celeriac	15.0	71	100	1000	0.3
Celery	10.0	70	150		0.3
Chinese cabbage	3.6	70	350		0.3
Courgette/Zucchini	10.0	70	50	450	0.3
Eggplant/Aubergine	10.0	70	100		0.3
Endive	4.0	80	150		0.3
Fennel	10.0	80	80		0.3
Garlic	40.0	80	60		0.3
Kohlrabi	10.0	70	80	500	0.3
Leek	12.2	68	35		0.3
Lettuce (Butterhead)	4.0	80	150		0.3
Lettuce (Crisp)	4.0	80	300		0.3
Melon	7.0	80	1500		0.3
Onion	10.4	75	20		0.3
Parsnip	17.0	95	50		0.3
Radicchio	5.0	40	200		0.3
Red cabbage	9.3	65	350		0.3
Savoy cabbage	9.3	65	350		0.3
Small radish (Spring)	4.8	84	4	20	0.3
Small radish (Summer)	4.2	85	4	20	0.3
Squash	10.0	80	0		0.3
Water melon	7.0	80	1500		0.3
White cabbage	9.3	65	350		0.3
White cabbage (Summer)	7.8	75	350		0.3
White cabbage (Winter)	12.0	54	350		0.3
White cabbage (Industry)	9.3	65	350		0.3

625

626

627 Figure 1: Examples of the relation between total above-ground dry matter (TDM) and
628 marketable fresh matter yield (MFY) for different levels of plant available N as defined as N
629 in soil and plant at harvest, showing linear (A), curvilinear (B), or no (C) relationships.

630

631 Figure 2: Bland-Altman plots for differences between marketable yields converted from
632 simulated total dry matter using the regression approach (A) and the population approach (B)
633 versus observed marketable yields from different experimental crop rotations.

634

635 Figure 3: Marketable yields converted from simulated total dry matter: Plotting the regression
636 approach results versus the population approach results for simulations of different
637 experimental crop rotations.

638

639 Figure 4: Simulation of a plant spacing experiment with white cabbage “Quisto”: Mean
640 individual marketable yield (A) and crop residues (B) are calculated by the model first at
641 harvest time. Above-ground crop N content (C) and dry matter development (D) are
642 continuously simulated. Data show measured values at 30×50 cm (■) and 50×50 cm (□)
643 spacing, calculated final fresh matter yield at 30×50 cm (◆) and 50×50 cm (◇) spacing
644 and dynamic simulations for 30×50 cm (—) and 50×50 cm (---) spacing.

645

646

647 Figure 5: Simulation of an N fertiliser experiment with cauliflower “Fremont”. Measured
648 individual head weight distribution for a 240 kg N ha^{-1} (A), 120 kg N ha^{-1} (B) and 0 kg N ha^{-1}
649 (C) treatment (n = 24). Normal distribution based on simulated mean head weight and a
650 cauliflower default coefficient of variation (0.3) for a 240 kg N ha^{-1} (D), 120 kg N ha^{-1} (E)
651 and 0 kg N ha^{-1} (F) fertiliser scenario. Lower boundary for gradeout was set to an individual
652 head weight of 600g.