ABSTRACT. Objective. To estimate the long-term cost-effectiveness of a hypothetical screening program for untreated amblyopia in 3-year-old children conducted by orthoptists in all German kindergartens in the year 2000.

Methods. A cost-utility analysis was performed for which a decision tree was combined with a Markov model. Incremental costs and effects during the children’s remaining lifetime were estimated. The model took into account the probability of treatment without screening, age-specific treatment success rates, costs of screening and treatment, as well as effects of unilateral and bilateral visual impairment caused by amblyopia and other eye diseases coming along later in life on quality of life (utility). Model parameter values were obtained from a field study of orthoptic screening in kindergarten, from the literature, and from expert interviews. Costs were estimated from a third-party payer perspective. Uncertainty was assessed by univariate and probabilistic sensitivity analysis (Monte Carlo simulation).

Results. The incremental cost-effectiveness ratio (ICER) of orthoptic screening was 7397 Euro (€) per quality-adjusted life year (QALY) when costs and effects were discounted at 5%. In univariate sensitivity analysis, the ICER was sensitive to the uncertainty regarding the utility of unilateral visual impairment and to the discount rate for effects; besides uncertainty regarding the prevalence of untreated amblyopia, the odds ratio of success of treatment when started late, and the probability of treatment without screening had a noticeable but much smaller effect. Monte Carlo simulation yielded a 90% uncertainty interval for the ICER of 3452 €/QALY to 72 637 €/QALY; the probability of an ICER <25 000 €/QALY was 84%.

Conclusions. The ICER of orthoptic screening seems to fall within a range that warrants careful consideration by decision-makers. Much of the uncertainty in results comes from the uncertainty regarding the effect of amblyopia on quality of life. To reduce this uncertainty, the impact of amblyopia on utility should be investigated.

PEDIATRICS 2004;113:e95–108. URL: http://www.pediatrics.org/cgi/content/full/113/2/e95; amblyopia, vision screening, children, preschool, cost-effectiveness, decision modeling, Markov process.

ABBREVIATIONS. VA, visual acuity; QALY, quality-adjusted life year; ICER, incremental cost-effectiveness ratio; CI, confidence interval; OR, odds ratio; BMES, Australian Blue Mountain Eye Study.

Preschool vision screening mainly aims at preventing amblyopia by early detection and subsequent treatment of amblyogenic factors such as strabismus and certain refractive errors. Recently, there has been concern about the lack of scientific data on the effectiveness of preschool vision-screening programs, and rigorous evaluation has been called for. Because of increasing health expenditure, there is a growing interest among decision-makers in the cost-effectiveness of medical interventions. Many argue that the cost-effectiveness of a service has to be demonstrated before it may be covered by social health insurance or national health service programs. However, little is known about the cost-effectiveness of childhood vision-screening programs.

In a recent field study, we analyzed the effectiveness and cost-effectiveness of vision screening performed by orthoptists in German kindergartens. Orthoptists are specialized medical aides with thorough training in infant and child vision assessment, ocular motility-disorder examinations, and amblyopia management. In this study, screening was performed at age 3 because, in Germany, children may attend kindergarten from age 3 onward and the fourth year of life was considered best for vision screening. From age 3 onward, in most children monocular visual acuity (VA) can be assessed reliably by simple screening methods, and there is evidence that treatment is most effective when administered as early as possible.

In the field study, the clinical endpoint of analysis was the ophthalmologic diagnosis, with the number of newly detected cases of amblyopia being the measure of effects. However, diagnosis of amblyopia is only an intermediate endpoint, especially with respect to the cost-effectiveness of screening. A more comprehensive analysis should also take into account the costs and effects of subsequent treatment and the effect of a visual deficit on quality of life throughout the remaining lifetime. Such an analysis would require either many years of follow up or the use of economic modeling. Modeling allows one to combine different sources of evidence, to transfer study results to different contexts and extrapolate them to the longer term, and to analyze uncertainty of results in detail.
Building on the results of the field study, the purpose of this analysis was to model the long-term cost-effectiveness of a hypothetical orthoptic screening program for children at age 3 conducted in all German kindergartens in the year 2000. Incremental costs and effects that would occur during the children's remaining lifetime were estimated. The uncertainty inherent in the model was analyzed in detail to assess the precision of results and to identify those variables with the greatest potential impact on cost-effectiveness.

**METHODS**

This section is organized as follows: First, the study design is described in general and then the decision-analytic model is explained in detail, followed by the description of the model parameters used.

**Study Design**

Based on a decision-analytic model, a cost-utility analysis was performed in which quality-adjusted life years (QALYs) were used as the measure of effects. QALYs are calculated by weighting the duration of health states by a preference-based score of health-related quality of life (utility), measured on a scale from 0 (dead) to 1 (perfect health).\(^{22}\)

In the model, the strategy "orthoptic screening at age 3 in kindergarten" ("screening") was compared with the strategy "usual care." Although usual care involves no orthoptic screening in kindergarten, amblyopia still may be detected when children see an ophthalmologist with or without referral (eg, by their pediatrician). The main outcome measure was the incremental cost-effectiveness ratio (ICER), ie, the ratio of the difference in mean costs \( \frac{C_{\text{screening}}}{C_{\text{usual care}}} \) and the difference in mean effects \( \frac{E_{\text{screening}}}{E_{\text{usual care}}} \) between the screening and usual-care strategies:

\[
\text{ICER} = \frac{C_{\text{screening}} - C_{\text{usual care}}}{E_{\text{screening}} - E_{\text{usual care}}} \tag{1}
\]

Hence, for the comparison, a purely incremental approach was chosen, ie, only the differences in costs and QALYs between the 2 strategies were considered. This means that, when calculating QALYs, only health states were considered with utility losses set off by

- unilateral visual impairment caused by a not successfully treated target disease;
- visual impairment caused by any other unilateral eye disease coming along later in life. In a proportion of patients already affected by unilateral visual impairment caused by a target disease, this would cause bilateral visual impairment; in all other persons, this would result in unilateral visual impairment.

Visual impairment caused by any bilateral eye disease was not considered in the model, because it would cause bilateral visual impairment in both patients with the target disease as well as persons without and hence would not set off differences in QALYs between the strategies compared.

For the model, a threshold for the presence of visual impairment had to be defined. In accordance with various therapeutic,\(^{16,17}\) epidemiologic,\(^{23,24}\) and disability studies,\(^{25}\) this threshold was set at a corrected VA of \(<0.5 (20/40): Monocular visual impairment was defined as VA \(<0.5 in the worse eye and \(\geq 0.5 in the better eye, and bilateral visual impairment was defined as VA \(<0.5 in both eyes.

The target disease was defined as untreated unilateral amblyopia, which, without treatment, would cause a lasting visual impairment.

**Decision-Analytic Model**

A decision tree was combined with a Markov model. Analysis was performed by using the DATA software package (version 4.0, TreeAge Software, Inc, Williamstown, MA).

**Decision Tree**

The decision tree (Fig 1) has 1 decision node that distinguishes the strategies screening and usual care. In the screening strategy, the model takes into account, by respective chance nodes, that only a proportion of children attend kindergarten at age 3 (labeled...
"kindergarten +/-") and would be present and participate in screening on the scheduled examination day (participation +/-). It also takes into account that a proportion of participating children would not cooperate sufficiently with the screening examination (cooperation +/-). Children with a positive screening result as well as noncooperative children would be referred to an ophthalmologist. At this point, the model takes into account that a proportion of parents would not comply with the referral (ophthalmologist +/-). Children in whom amblyopia was newly diagnosed by the ophthalmologist would be treated and, by a certain probability, successfully achieve VA >0.5 in the amblyopic eye. Therefore, at the end nodes of the screening strategy, 3 different health states can be reached:

1. “Healthy*: if a target disease was not present or was treated successfully.
2. “Unilateral visual impairment (amblyopia)**: if a target disease was detected by the screening but treatment was not successful.
3. “Untreated amblyopia": if a target disease was not treated yet.

At the end nodes of the usual-care strategy, only the healthy (if no target disease was present) or untreated-amblyopia (if a target disease was present) health states can be reached.

Markov Model

At each end node of the decision tree, a Markov process starts, ie, the end node of the decision tree corresponds to the initial health state of a Markov process. In a Markov process the course of a disease is divided into distinct states, and transition probabilities are assigned for movement between them over a discrete time period called the Markov cycle.26 By attaching estimates of resource use and health effects to the states and transitions in the model and then running the model over a large number of cycles, long-term costs and effects can be estimated.

According to the 3 different initial health states, 3 Markov processes (A–C) were distinguished. Markov process A starts with the initial untreated-amblyopia health state (Fig 2), which corresponds, eg, to the third end node from the top of the decision tree displayed in Fig 1. Besides, there are 5 additional health states defined in this Markov process. Transition to the healthy state takes place when the child is seen by an ophthalmologist and is treated successfully before age 11; transition to the unilateral-visual-impairment (amblyopia) health state occurs if the child is not treated successfully before age 11. From the healthy state, transition to the unilateral-visual-impairment health state takes place when a unilateral nonamblyopic eye disease comes along and causes unilateral visual impairment. From the unilateral-visual-impairment (amblyopia) health state, transition to the bilateral-visual-impairment (with amblyopia) health state occurs when the nonamblyopic eye of an amblyopic patient is affected by a unilateral eye disease, resulting in bilateral visual impairment. The state “death” can be reached from all other health states.

Markov processes B and C are subsets of Markov process A. Markov process B starts with the initial healthy state, and Markov process C starts with the initial unilateral-visual-impairment (amblyopia) health state. Possible transitions correspond to those described within Markov process A.

In the model presented here, the cycle length was 1 year. Simulation started in cycle 0, which corresponds to the 4th year of life (age 3), and was completed in cycle 86, which corresponds to the 90th year of life. To account for gender differences in mortality and incidence of visual impairment, all Markov processes were made gender specific.

![Fig 2. Markov process A for course of disease with an initial untreated-amblyopia health state. Health states are represented by ovals, and possible transitions between those states are shown by arrows. Variable names adjacent to the arrows are the transition probabilities of the model (see Table 1). Variable names refer to men; for women tp.mono.m, tp.vis.m, and tpDeathm are replaced by tp.mono.f, tp.vis.f, and tpDeathf, respectively. *, Initial health state of Markov process B (shaded in bright gray); **, initial health state of Markov process C (shaded in dark gray).](http://www.pediatrics.org/cgi/content/full/113/2/e95)
<table>
<thead>
<tr>
<th>Model Parameter</th>
<th>Source*</th>
<th>Name Used in Decision Tree or Markov Model</th>
<th>Parameter Value</th>
<th>Range Used for Univariate Sensitivity Analysis</th>
<th>Distribution Used in Monte Carlo Simulation</th>
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<tbody>
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<td>Beta (27.115)†</td>
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<td>0.935–0.990</td>
<td>Beta (107; 4)†</td>
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<td>Test Characteristics of Orthoptic Screening</td>
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<td>sens_ort</td>
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<td>Unilateral visual impairment</td>
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<td>0.920–1.000</td>
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<td>Organization per screening examination, €</td>
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<td>c_org</td>
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<td>2.62–7.86</td>
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<td>Orthoptic screening examination excluding organization, €</td>
<td>Fixed costs, €</td>
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<td>c_fix</td>
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<td>3.28–5.08</td>
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<td>Variable costs, €</td>
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<td>Ophthalmologic examination, €</td>
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<td>c_op</td>
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<td>27.30–45.50</td>
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<td>Costs of treatment (€), by age at beginning of treatment</td>
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<td>425–1929</td>
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<td>9 years</td>
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<td>10 years</td>
<td>960</td>
<td>225–1805</td>
<td>Student’s t (6; 960; 225)¶</td>
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<td>Transition Probabilities</td>
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<td>t_deathm</td>
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<td>Mortality for women</td>
<td>L</td>
<td>t_deathf</td>
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<tr>
<td>Probability of visit to ophthalmologist per year, by age</td>
<td>L</td>
<td>op_visit</td>
<td>0.230</td>
<td>0.115–0.345</td>
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<td>Probability of visit × op_visit, distribution of p_visit: Uniform (0.5; 1.5)‡</td>
<td>3 years</td>
<td>0.270</td>
<td>0.135–0.406</td>
<td>Uniform (0.5; 1.5)‡</td>
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<tr>
<td>4 years</td>
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<td>0.175–0.524</td>
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<td>5 years</td>
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<td>0.139–0.417</td>
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<td>9 years</td>
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<td>Probability of treatment success when started late:</td>
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<td>eff_late</td>
<td>OR_k for treatment success, by age class (k)</td>
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<td>OR = exp(b); distribution of b: Normal (−0.76; 0.23)§</td>
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<td>4–5 years</td>
<td>OR = 0.468</td>
<td>0.187–0.536</td>
<td>OR = exp(b); distribution of b: Normal (−1.15; 0.27)§</td>
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<tr>
<td>6–10 years</td>
<td>OR = 0.317</td>
<td>0.187–0.536</td>
<td>OR = exp(b); distribution of b: Normal (−1.15; 0.27)§</td>
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<tr>
<td>Model Parameter</td>
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<td>Name Used in Decision Tree or Markov Model</td>
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<td>Transition probability per year from unilateral visual impairment (amblyopia) to bilateral visual impairment, by gender and age</td>
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<td>tp_vis_m</td>
<td>0.00166</td>
<td>0.00083-0.00248</td>
<td>p_trans × tp_vis_m, distribution of p_trans: Uniform [0.5; 1.5]‡</td>
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<tr>
<td>Men</td>
<td>L</td>
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<td>60-69 years</td>
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<tr>
<td>70-79 years</td>
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<tr>
<td>80-89 years</td>
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<td>Women</td>
<td>L</td>
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<td>0.00012-0.00048</td>
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<td>Transition probability per year from healthy to unilateral visual impairment (no amblyopia), by gender and age</td>
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<td>Men</td>
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<td>80-89 years</td>
<td>0.00862</td>
<td>0.00431-0.01293</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Discount Rate</td>
<td>**</td>
<td>cDR</td>
<td>0.050</td>
<td>0.000-0.050</td>
<td>(None)</td>
</tr>
<tr>
<td>Discount rate for costs</td>
<td>L</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Discount rate for effects</td>
<td>L</td>
<td>eDR</td>
<td>0.050</td>
<td>0.000-0.050</td>
<td>(None)</td>
</tr>
</tbody>
</table>

* Source of parameter value: L, literature; F, field study; E, expert opinion (see text for details).
† Beta distribution (α; β): values of α and β with decimal digits stand for fitted Beta distributions.
‡ Uniform distribution (minimum; maximum).
§ Normal distribution (mean; standard deviation).
¶ Discounted by 5% discount rate to year when treatment is started.
|| Student's t distribution (degrees of freedom: mean; observed standard deviation).
# Mortality rate for men in Germany.
** Mortality rate for women in Germany.
Model Parameters

Model parameter values were obtained from a field study of orthoptic screening in kindergarten, from the literature, and from expert interviews (Table 1). For each parameter, a (mean) value, which was used for the base analysis, was defined as well as a possible value range, according to which parameter values were varied in 1-way sensitivity analysis. Furthermore, for parameters that, in principle, could be sampled, distributions were specified for probabilistic sensitivity analysis (Monte Carlo simulation) as follows: for proportions of the form \( p = a / (a + b) \), such as test characteristics and utilities, which are bound to a 0–1 interval, Beta distributions were specified \([Beta(a, b)]\) with \( a = a + 1 \) and \( b = b + 1 \).26–27 Normal or Student’s \( t \) distributions were specified for the mean of continuous parameter values. If there were only data on the mean and the likely range of the 95% confidence interval (CI) of a continuous parameter but not the variance, the standard error was estimated by the formula

\[
s = \frac{u - l}{2 \times 1.96},
\]

where \( u \) and \( l \) are the upper and lower limits of the range, respectively.27 If there was only information on the range but not on the mean of a parameter, then a uniform distribution was used.

Screening Population

In 2000, the number of children aged 3 in Germany was 814,700; 51.3% were male. Of these children, 435,700 (56.3%) were attending 1 of 43,728 kindergartens.29 Thus, the mean number of children aged 3 per kindergarten was 10.49. According to other field studies of vision screening in German kindergartens in which the participation rate ranged between 88%29 and 92%,30 the participation rate was assumed to be 90%, which would yield a mean of 9.44 participating children per kindergarten. For univariate sensitivity analysis, the range of the participation rate was set at 80% to 100%. This range was assumed to represent the 95% CI, and the standard error was estimated according to formula 2. Based on the estimated mean and standard error, a \( \beta \) distribution was fitted by using the method of moments estimation.27–31

Prevalence of Target Disease and Test Characteristics of Orthoptic Screening

Data on the prevalence of untreated amblyopia, on the screening test characteristics (sensitivity, specificity, and proportion of inconclusive results), and on the compliance with ophthalmologic examinations were obtained from a field study in which 1180 3-year-old children were examined by orthoptists in 121 German kindergartens.10

In the field study, the orthoptic screening examination consisted of cover tests, examination of eye motility and head posture, and uncorrected monocular VA testing using the Lea single optotype test (required test distance 10 ft [3 m], Precision Vision, Villa Park, IL) with pass threshold set at \( \geq 0.8 / 20 / 25 \) monocular VA in both eyes; or \( \geq 0.5 / 20 / 40 \) in both eyes and \( \geq 1 \) line difference between the VA of the right and left eyes (L. Hyvärinen, written communication, 1998). The ophthalmologic criteria for a target disease (positive gold standard) were: any newly administered patching therapy; or any newly administered spectacle therapy, if the corrected VA was \( \geq 0.4 / 20 / 50 \) in either eye or if the difference of VA between the right and left eyes was \( \geq 3 \) lines.10

For the base analysis it was assumed that, without treatment, children with a positive gold standard (2.2%) would be affected by a lasting unilateral visual impairment. For univariate sensitivity analysis, the range of prevalence was set at 1.0% to 5.0%, which includes the range of prevalence of amblyogenic factors stated in a recent review (2.7%–4.4%)3 and takes into account that more than half of all cases may already be detected at the time of screening. For the Monte Carlo simulation, a Beta distribution was defined based on the proportion of gold-standard-positive children found in the field study, \( a / (a + b) \), where \( a \) equals the number of children with positive gold standard, and \( b \) equals the number of children without positive gold standard.

For the test characteristics and compliance rates, values measured in the field study were used as base values. The limits of their 95% CIs were used for univariate sensitivity analysis, and Beta distributions were specified for the Monte Carlo simulation.

Because data on the compliance with direct referral to an ophthalmologist after inconclusive screening could not be obtained from the field study due to its design,24 this value had to be estimated based on other studies14,25 and was set at 90%. In univariate sensitivity analysis, this value was varied from 80% to 100%, and for Monte Carlo simulation a uniform distribution on this interval was used.

Success Rate of Treatment

In studies that used a VA \( \geq 0.5 \) as the threshold for treatment success for amblyopia, reported success rates range from 50% to 70%.24,26–28 In a meta-analysis based on 23 studies with 689 patients included, Flynn et al26 found an overall success rate of 74.3%. The authors developed a multiple logistic regression model to analyze the effect of age at the beginning of treatment on the success rate, controlling for type of amblyopia and VA at beginning of treatment. In a second study,17 in which 589 patients were included and the overall success rate was 59.9%, Flynn et al validated the regression model and again found a significant influence of age on treatment success: Compared with age group 0 to 3, the odds ratio (OR) for the success of treatment started at age group 4 to 5 was 0.47, and at age group 6 to 10 it was 0.32. They also reported the estimated regression coefficients \( b_k \) for the age groups \( k \) and their standard errors \( s_k \), from which the 95% CIs of OR\(_k\) could be calculated by using the formula27

\[
exp(\beta_k + z \times s_k),
\]

which yielded [0.30; 0.73] for age group 4 to 5 and [0.19; 0.54] for age group 6 to 10.

In the model, the mean treatment success rate at age 3 (p\(_{eff}\)) was set at 75% with the range of 60% to 90% used for univariate sensitivity analysis. For the Monte Carlo simulation, a Beta distribution was fitted for (p\(_{eff}\)), assuming that the mentioned range corresponds to the 95% CI and deriving the standard error according to formula 2.

Success rates of treatment started after age 3 (eff\(_{late}\)) were calculated based on the mentioned OR\(_k\) for older age groups \( k \) using the formula

\[
eff_{late, k} = \frac{p_{eff}/(1 - p_{eff}) \times OR_k}{1 + p_{eff}/(1 - p_{eff}) \times OR_k},
\]

Table 2 shows the resulting success rates of late treatment (eff\(_{late}\)) for various success rates of treatment started at age 3. In univariate sensitivity analysis, OR\(_k\) values were jointly varied according to their 95% CI, mentioned above. In the Monte Carlo simulation, OR\(_k\) values were calculated by using the coefficients for age groups \( b_k \), which were estimated in the logistic regression model, in the formula27

\[
OR_k = \exp(b_k),
\]

assuming a normal distribution for \( b_k \), with the estimated \( b_k \) as the mean and the standard deviation \( s_k \). Thus, \( b_k \) was calculated according to the formula27

\[
b_k = \ln OR_k - z \times s_k,
\]

where \( z \) follows the standard normal distribution.

<table>
<thead>
<tr>
<th>Age 3, %</th>
<th>Age 4–5, %</th>
<th>Age 6–10, %</th>
</tr>
</thead>
<tbody>
<tr>
<td>Success Rate (p(_{eff}))</td>
<td>Success Rate (OR = 0.47(^*))</td>
<td>Success Rate (OR = 0.32(^*))</td>
</tr>
<tr>
<td>90</td>
<td>81</td>
<td>74</td>
</tr>
<tr>
<td>85</td>
<td>73</td>
<td>64</td>
</tr>
<tr>
<td>80</td>
<td>65</td>
<td>56</td>
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<tr>
<td>75</td>
<td>58</td>
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<tr>
<td>70</td>
<td>52</td>
<td>42</td>
</tr>
<tr>
<td>65</td>
<td>45</td>
<td>37</td>
</tr>
<tr>
<td>60</td>
<td>41</td>
<td>32</td>
</tr>
</tbody>
</table>

\(^*\) OR for success rate of treatment started after age 3 (eff\(_{late}\)) calculated based on success rate of treatment started at age 3; the success rate (eff\(_{late}\)) is calculated according to the formula: eff\(_{late}\) = \[p_{eff}/(1 - p_{eff}) \times OR_k]/[1 + p_{eff}/(1 - p_{eff}) \times OR_k].
Utilities

The utility of the bilateral-visual-impairment health state was derived from a recent study of a large series of patients with visual loss from various ocular diseases using the time trade-off method.\textsuperscript{39} In this study, a formula for converting VA of the better eye to a mean utility value was derived via regression analysis:

\[ U(\text{VA}) = \frac{0.374 \times \text{VA} + 0.514}{\text{VA}}. \] (7)

For the incremental approach, only the reduction of utility caused by visual impairment was relevant. Therefore, the utility of VA = 1.0 was set at \( U = 1.00 \), ie, it was increased by 0.11 compared with the utility predicted by formula 7. For a conservative estimate of the utility of the bilateral-visual-impairment health state, the utility of VA = 0.4 was calculated by using formula 7 and then also increased by 0.11, resulting in \( U = 0.78 \). For the health states without any visual impairment, the utility was constantly set at \( U = 1.00 \). Thus, compared with no visual impairment, utility of bilateral visual impairment was reduced by 0.22. Based on utilities reported by the same working group,\textsuperscript{39} this reduction of utility was varied by \( \pm 1/3 \), ie, from 0.15 to 0.29 for univariate sensitivity analysis, corresponding to utilities from 0.71 to 0.85. This interval was considered to reflect the 95% CI, and a Beta distribution was fitted for the Monte Carlo simulation.

The utility associated with unilateral visual impairment caused by amblyopia has not been investigated specifically thus far. In a recent study of the same working group, unilateral impairment (defined as VA \( \leq 0.5 \) in the worse eye and VA \( \geq 0.8 \) in the better eye) from various ocular diseases was found to cause a mean reduction of utility of 0.08, using the time trade-off method.\textsuperscript{40} Because individuals with only 1 sound eye due to amblyopia since childhood may develop compensatory visual mechanisms,\textsuperscript{9} the reported utility reduction was considered a maximum. It thus was varied from 0.00 to 0.08 in univariate sensitivity analysis, ie, from no disutility at all to the disutility stated in the empirical study,\textsuperscript{40} which corresponds to utilities ranging from 0.92 to 1.00. Having no additional information, for the base analysis a utility of 0.96 was used, ie, the middle of the interval. For the Monte Carlo simulation, a uniform distribution on the interval [0.92,1.00] was used.

Transition Probabilities of Markov Model

Probability of Successful Treatment Under Usual Care

Data on the age-specific probability for an amblyopic child to receive treatment under usual care (ie, without orthoptic screening) were not available. It was assumed that untreated amblyopia was definitely detected and treated when an affected child was seen by an ophthalmologist for whatever reason. The probability for a child to be seen by an ophthalmologist (op.visit) was estimated based on a representative survey of ophthalmologic patients in Germany,\textsuperscript{41} from which the age-specific number of ophthalmologic patients in 1997 was estimated.\textsuperscript{42} The age-specific probability (op.visit) was derived by dividing the age-specific number of patients by the age-specific number of persons in the German population in 1997.\textsuperscript{43} For the model it was assumed that the age-specific probability (op.visit) was independent of former visits and of the presence of amblyopia. Table 1 shows the calculated probabilities. For example, without screening, at age 3, 23% of children with untreated amblyopia would be seen by an ophthalmologist and receive treatment; before age 8, a accumulated 94% would have received treatment, and 95% would have received treatment before age 11. In univariate sensitivity analysis, the age-specific probabilities were reduced and increased by the factors 0.5 and 1.5, respectively. For the Monte Carlo simulation, a uniform distribution on the interval [0.5; 1.5] was used for this factor.

In Markov process A, before age 11 (cycles 0–7) the probability of transition from the untreated-amblyopia health state to the healthy state (ie, the probability of successful treatment under usual care) was obtained by multiplying the age-specific probability (op.visit) with the age-group-specific success rate of treatment. From age 11 onward, the probability of this transition was assumed to be 0.

Incidence of Unilateral Visual Impairment Other Than Amblyopia

Because data on age- and gender-specific incidence of unilateral visual impairment were not available in the literature, incidence of children with unilateral amblyopia was divided into fixed costs (eg, for traveling) and variable costs, which depended on the number of children examined per kindergarten (eg, labor.
TABLE 3. Costs of Treatment of Amblyopia by Age at Beginning of Treatment

<table>
<thead>
<tr>
<th>Age at Beginning of Treatment, ( y )</th>
<th>Costs Without Discounting</th>
<th>Costs Discounted at 5% Discount Rate*</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean ( m_j ), €</td>
<td>Standard Error ( s_j ), €</td>
</tr>
<tr>
<td>3</td>
<td>1844 €</td>
<td>316 €</td>
</tr>
<tr>
<td>4</td>
<td>1734 €</td>
<td>318 €</td>
</tr>
<tr>
<td>5</td>
<td>1619 €</td>
<td>319 €</td>
</tr>
<tr>
<td>6</td>
<td>1496 €</td>
<td>320 €</td>
</tr>
<tr>
<td>7</td>
<td>1365 €</td>
<td>327 €</td>
</tr>
<tr>
<td>8</td>
<td>1156 €</td>
<td>278 €</td>
</tr>
<tr>
<td>9</td>
<td>953 €</td>
<td>243 €</td>
</tr>
<tr>
<td>10</td>
<td>680 €</td>
<td>180 €</td>
</tr>
</tbody>
</table>

* Discounted to year at beginning of treatment.

costs for examination time). Vice versa, when calculating the costs per single examination, the share of fixed costs depended on the number of children per kindergarten. Because in the field study the average number of children examined per kindergarten was 9.88, fixed costs per screening examination were multiplied by the factor 9.88/9.44 to adjust for the smaller average number of children (9.44) to be examined per kindergarten in the model.

For the fixed and variable costs per single examination in kindergarten, standard errors of the mean could be estimated in the field study. Thus, these 2 cost categories were varied according to their 95% CI in univariate sensitivity analysis, and normal distribution was used for Monte Carlo simulation. With respect to organization, costs were varied by ±25% in univariate sensitivity analysis. This interval was considered to reflect the 95% CI, and for Monte Carlo simulation, a normal distribution was fitted by estimating the standard error according to formula 2.

To account for changes in participation rates that would lead to changes in the average number of children examined per kindergarten, the mean and the standard error of fixed costs were adapted accordingly.

Costs of Diagnostic Ophthalmologic Examination

Costs of a standard ophthalmologic examination of children referred from screening were based on the German social health insurance’s relative value scale for outpatient physician services, which defines individual physician services and states point volumes for them. For the conversion factor (point value), which varies by region, a German average of 0.041 € was used. For univariate sensitivity analysis, these costs were varied by ±25%, which corresponded to the range that the conversion factor was likely to vary by region. This interval was considered to reflect the 95% CI. For Monte Carlo simulation, a normal distribution was fitted by estimating the standard error according to formula 2.

Treatment Costs

Because there were no published data on the cost of amblyopia treatment, costs were estimated based on expert opinion. Seven experts of amblyopia treatment (4 ophthalmologists and 3 orthoptists) from different German treatment centers filled in a standardized questionnaire. In the questionnaire, medical services and items (eg, glasses, patches, etc) possibly used for amblyopia treatment were listed. Respondents were asked to estimate the mean number of services and items used per year during up to 9 years of treatment. Services and items were valued monetarily by using “administrative” prices paid by the German statutory health insurance. The methodology of estimating treatment costs has been described in detail elsewhere.49

It was assumed that treatment was completed before age 12. To estimate the costs of treatment started later than age 3, treatment costs were reduced by excluding the costs of years that would correspond to age >12.

Costs of treatment started at age 3 to 10 were calculated based on each expert’s estimate of resource utilization. From these figures, mean estimated costs \( m_j \) of treatment started at age \( j \) were calculated (and used for the base analysis) as well as the standard errors \( s_j \), which reflect the uncertainty of the expert with regard to average costs (Table 3). In univariate sensitivity analysis, the treatment costs were jointly varied according to their 95% CIs. For the Monte Carlo simulation, age-specific treatment costs \( k_j \) were calculated by using the formula

\[
k_j = m_j + t \times s_j, \tag{10}
\]

where \( t \) follows a standardized Student’s \( t \) distribution with 6 degrees of freedom.

Discounting

Costs and effects were discounted at 5%. In sensitivity analysis, discount rates of 0% and 3% were also used.

Sensitivity Analysis

One-Way Sensitivity Analysis

To analyze the effect of uncertainty in single model parameters on the ICER, univariate sensitivity analysis was performed by varying single-parameter values according to the ranges described above and repeating the analytical solution of the model.

Probabilistic Sensitivity Analysis (Monte Carlo Simulation)

In the Monte Carlo simulation, values for all model parameters were sampled from their respective specified distributions, and the analytical solution of the model was repeated. \( ^{27} \) This process of resampling from each of the distributions and recalculating the incremental costs and effects from the model was repeated 10,000 times to generate a distribution of the estimated ICER. Uncertainty intervals were estimated from the simulated data by taking the 2.5th and 97.5th percentile values to represent the endpoints for a 95% interval,27 and cost-effectiveness acceptability curves \( ^{51} – ^{53} \) were constructed.

RESULTS

Base Result

The screening strategy was associated with incremental costs of 13.34 € per child and incremental effects of 1.803 × 10⁻³ QALYs per child. Thus, for the total population of 814,700 3-year-old children, of whom 50.7% (90% × 56.3%) would participate in screening, incremental costs were 10.87 million €, and incremental effects were 1469 QALYs. The ICER was 7397 €/QALY.

Univariate Sensitivity Analysis

Figure 3 shows the effect of varying single-parameter values on the ICER. Uncertainty with respect to the utility of unilateral visual impairment had the strongest potential impact on the ICER. If this utility was 0.92, the ICER decreased to 3706 €/QALY; if it was 1.00 (ie, if only bilateral visual impairment but not unilateral visual impairment affected utility), the ICER increased to 1.904 million €/QALY. Figure 4 shows the association between the ICER and the utility of unilateral impairment for the range of 0.92 to 1.00: If the utility was >0.988, the ICER increased to >25 000 €/QALY; if it was >0.994, the ICER increased to >50 000 €/QALY.
Besides uncertainty with respect to the prevalence of the target disease, the OR of treatment success when started later and the probability for a child to be seen by an ophthalmologist (opH14061 visit) had a marked, but much smaller impact on the ICER. All other parameters had only little impact, among those being the treatment costs. Almost no impact had the uncertainty with respect to the incidence of visual impairment caused by other eye disease and the utility of bilateral visual impairment.

If only costs but not effects (QALYs) were discounted, the ICER was 2219 €/QALY. If neither costs nor effects were discounted, the ICER was 2163 €/QALY. If costs and effects were both discounted at 3%, the ICER was 5138 €/QALY.

Monte Carlo Simulation

Uncertainty Intervals

Figure 5 shows the joint distribution of incremental costs and effects generated in 10 000 simulations.
on the cost-effectiveness plane. The 95% uncertainty interval for the incremental costs per child ranged from 10.14 to 18.63 €, and for incremental effects from $8.607 \times 10^{-5}$ to $5.005 \times 10^{-3}$ QALYs. For the total population of 814,700 3-year-old children, 95% of incremental costs lay between 8.26 million and 15.18 million €, and 95% of the effects lay between between 70 and 4078 QALYs.

The 95% uncertainty interval of the ICER was 3138 to 148,228 €/QALY, which corresponds to the slope
of the lines representing the interval endpoints in Fig 5. The 90% uncertainty interval was 3452 to 72,637 €/QALY. Without discounting effects, the 95% uncertainty interval was 903 to 30,053 €/QALY.

Cost-Effectiveness Acceptability Curves

When costs and effects were discounted at 5%, Monte Carlo simulation yielded an ICER <25,000 €/QALY in 84% and an ICER <50,000 €/QALY in 92%. Without discounting effects, an ICER <25,000 was yielded in 97%, and an ICER <50,000 in 99%. This is displayed by the cost-effectiveness acceptability curves in Fig 6, which give the proportion of observed simulation results lying below varying threshold values for the ICER.

Because much of the variability of the results was due to the uncertainty with respect to the utility of unilateral visual impairment, Monte Carlo simulations were also performed keeping this utility constant at different levels and varying all other parameters according to their distributions (Fig 7). If this utility was 0.92, 0.94, 0.96, 0.98, or 0.99, in 95% of the simulations the ICER was below 6375, 8491, 12,561, 25,296, and 50,789 €, respectively.

DISCUSSION

The Markov model aimed at estimating the lifelong gain in health-related quality of life due to prevention of amblyopia through screening and subsequent treatment, as well as the associated costs. The model took into account that children with untreated amblyopia may be detected and successfully treated without screening.

The ICER of orthoptic screening at age 3 in German kindergarten was found to be ~7400 €/QALY when costs and effects were discounted at 5%. Thus, the ICER was more favorable than that of many routinely provided health care interventions54 and similar to the ICER found for diabetic retinopathy screening.55 Because the analysis included effects occurring during the remaining lifetime of up to 86 years, discounting effects had substantial impact on the ICER. Without discounting, the ICER was just about one third of the base result. Discounting costs had only little impact on the ICER, because all considered costs occurred during the first 9 years after screening.

Univariate sensitivity analysis showed that uncertainty regarding the utility of unilateral visual impairment had great potential impact on the ICER. If unilateral visual impairment due to amblyopia was not associated with any loss in utility, then orthoptic screening would very likely not be cost-effective. In other words: Just the risk of developing bilateral visual impairment later in life would most likely not justify preschool vision screening. Yet, if unilateral visual impairment was associated with a utility loss of only 0.01, then the ICER fell below 30,000 €/QALY. ICERs obtained in Monte Carlo simulation showed a wide range but were mostly within limits likely to be acceptable to decision-makers. For example, the probability for the ICER being below 25,000 €/QALY was 84%.

Model Structure

For modeling, the course of disease had to be divided into distinct states, and hence thresholds for the presence or absence of visual impairment had to be defined. Because in reality no such thresholds

---

Fig 6. Cost-effectiveness acceptability curves of the screening strategy for various discount rates
exist, this necessarily entails some simplification. Theoretically, more health states could have been defined, representing different degrees of visual impairment but also making the model more complicated. However, for this purpose, respective data were not available. Uncertainty with respect to the definition of thresholds was addressed by varying model parameters widely in sensitivity analyses.

The model did not distinguish between different causes of amblyopia. This simplification had to be made, because data on age-specific treatment success rates were not available in more detail. However, in 2 studies based on large patient samples, no significant effect of the cause of amblyopia on treatment success was identified after controlling for age and VA at the beginning of treatment, which both had a significant effect. Although the effect of age was explicitly modeled, the effect of VA on treatment success was accounted for by varying treatment success rate at age 3 widely in sensitivity analysis.

The model did not allow any return to preceding health states, eg, from unilateral visual impairment (without amblyopia) to healthy (eg, after treatment of other eye diseases) states. Although such a return may be possible in reality (eg, after treatment), the error introduced by this restriction is considered small because most transition probabilities were estimated from prevalence data. For improvement of unilateral visual impairment caused by amblyopia without treatment, no evidence was found in the literature.

The Markov process was restricted to 86 cycles. Thus, possible QALY gains accumulated from age 90 onward were not considered. When discounted, these QALY gains would be very small. Without discounting, however, they would make the ICER more favorable, as would a future increase in life expectancy.

In the model, only direct medical costs for detection and treatment of the target disease were considered from the perspective of a third-party payer. Thus, administrative prices (eg, fees) were used where applicable. Because these prices are likely to be comparable to opportunity costs, and no cost sharing by patients was assumed, the calculated costs are probably close to direct medical costs from the societal perspective. Direct nonmedical costs (eg, for traveling to an ophthalmologist) were not considered but probably were small. Direct costs caused by visual impairment coming along later in life were not considered either, but it would have had only little impact on the results due to discounting: If, eg, bilateral visual impairment caused direct costs of 10 000 € per year, the ICER would drop by only 2% to 7220 €/QALY. Indirect costs, caused by losses in productivity due to visual impairment, were not considered either. Including indirect costs would most likely make the ICER more favorable. However, according to the Panel on Cost-Effectiveness in Health and Medicine, effects of disease on productivity or income may already be reflected by the preference-based valuation of health-related quality of life (utility), and hence there would be no need to include indirect costs in cost-utility analysis.

**Critical Model Parameter Values**

Values for model parameters were obtained from a field study, the literature, and a survey of experts.
Uncertainty in parameter values was addressed by defining a base model and conducting extensive univariate sensitivity analysis. To assess uncertainty in various model parameters simultaneously, probabilistic sensitivity analysis (Monte Carlo simulation) was performed.

Regarding the prevalence of the target disease, it was assumed that all children detected in the field study would have been affected by lasting visual impairment without treatment. By this assumption, prevalence may be overestimated. However, in the field study, the combined prevalence of the target disease and of amblyopia already treated was ~4%, which falls within the prevalence range for unscreened populations given in the literature.

The age-specific probability that a child with untreated amblyopia is seen by an ophthalmologist under usual care was derived from a representative cross-sectional survey of ophthalmologic patients in Germany. This procedure entailed considerable uncertainty, which was addressed by varying these parameter values widely in sensitivity analysis and assuming uniform distribution in Monte Carlo simulation. Under the extreme assumption that, without screening, none of the target diseases would be detected, the ICER would drop to 4528 €/QALY.

The incidence of unilateral visual impairment not caused by amblyopia was estimated based on a large epidemiologic study from Australia (BMES). Because of comparable ethnicity and level of medical care, comparability with the German population was assumed. Yet, in the BMES, only adults aged ≥49 were included. Thus, visual impairment before age 49 was not considered. However, its prevalence in the youngest age group (49–54 years) of the BMES was rather low. If the prevalence rate in the age group 49–54 years resulted from a constant incidence from birth onward, the ICER would be reduced only slightly to 7392 €/QALY.

For valuation of health-related quality of life, utilities were used that had been elicited for visual impairment caused by various ocular diseases. In the underlying studies, neither the cause of visual impairment nor age, gender, education, ethnicity, comorbidity, or duration had a significant impact on the utility. Thus, the use of nondisease-specific utilities seems to be justified, because specific data on the utility associated with unilateral visual impairment caused by amblyopia were not available in the literature. However, the utility loss caused by amblyopia starting very early in life could be comparatively smaller due to adaptation. Hence, in sensitivity analysis, the respective utility loss was varied widely, even to no loss at all.

The best estimate available for the costs of amblyopia treatment was based on expert opinion, which showed only little variability. For treatment started after age 3, lower treatment costs were assumed because of shorter duration. However, it may be possible that late treatment would be more intense and thus more costly, which would make the ICER more favorable.

CONCLUSIONS

The ICER of orthoptic screening seems to fall within a range that warrants careful consideration by decision-makers. The presented uncertainty intervals and cost-effectiveness acceptability curves should enable decision-makers to appraise the results based on their own risk aversion. Much of the uncertainty in results comes from the uncertainty regarding the effect of amblyopia on health-related quality of life. To reduce this uncertainty, additional studies should investigate the impact of amblyopia on utility in more detail.

ACKNOWLEDGMENTS

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